# **Identification of cambium stem cell factors and their positioning mechanism**

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## **Abstract**

 Wood constitutes the largest reservoir of terrestrial biomass. Composed of xylem, it arises from one side of the vascular cambium, a bifacial stem cell niche that also produces phloem on the opposing side. It is currently unknown which molecular factors endow cambium stem cell identity. Here we show that TDIF ligand-activated PXY receptors promote the expression of CAMBIUM- EXPRESSED AINTEGUMENTA-LIKE (CAIL) transcription factors to define cambium stem cell identity in the *Arabidopsis* root. By sequestrating the phloem-originated TDIF, xylem- expressed PXY confines the TDIF signaling front, resulting in the activation of CAIL expression and stem cell identity in only a narrow domain. Our findings show how signals emanating from cells on opposing sides ensure robust yet dynamically adjustable positioning of a bifacial stem cell layer.

 **One-Sentence Summary:** The TDIF-PXY ligand-receptor pair defines cambium stem cells by controlling the expression of CAIL transcription factors.

## **Main Text**

 In seed plants, stem cell populations that drive apical-basal growth are formed in the embryo. However, the vascular cambium (hereafter cambium), which promotes radial growth, and thus the majority of plant biomass, is formed following germination (*1*, *2*). In the *Arabidopsis thaliana* root, this occurs when cells with xylem identity promote stem cell function in their neighbors. This xylem identity cell layer is thus considered to be the cambium organizer (*3*) (**Fig. 1A**). The cambium is dynamic in size, ranging from a few to multiple undifferentiated cells, depending on the level of proliferation, yet contains only a single bifacial stem cell layer (*3*–*5*). Due to the rarity of transit amplifying divisions within the xylem or phloem lineages (*3*), the vast majority of observed cell divisions in the *Arabidopsis* cambium are stem cell divisions. The remaining undifferentiated cells in the cambium have xylem or phloem identity (**Fig. 1A**). How the organizer cells can exert their exquisite control over stem cells at variable distance is still unknown. A prerequisite to addressing this question is identification of regulators that define stem cell identity within the cambium, which have not been determined either. In the root cambium, the stem cell organizer is defined by a local signaling maximum of auxin, which contributes to stem cell positioning (*3*, *6*). Auxin promotes the expression of CLASS III HOMEODOMAIN-LEUCINE ZIPPER (HD-ZIP III) transcription factors defining xylem identity (*3*, *7*, *8*), as well as a receptor kinase, PHLOEM INTERCALATED WITH XYLEM (PXY) (*3*). TRACHEARY ELEMENT DIFFERENTIATION INHIBITORY FACTOR (TDIF), which is derived from phloem-expressed CLAVATA3/ESR-RELATED 41 (CLE41) and CLE44, is the cognate ligand of PXY. Disruption of TDIF-PXY signaling causes major patterning and stem cell maintenance defects (*9*–*14*). These defects suggest that the elusive regulators of cambium stem cell identity are likely to be TDIF- PXY regulated. Here we define a set of *AINTEGUMENTA-like* genes (CAMBIUM-EXPRESSED AILs; CAILs) as performing this function, and show that their positioning is defined by opposing

- gradients of TDIF and auxin.
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## **Identification of CAILs as cambium stem cell factors downstream of TDIF-PXY**

 To identify genes involved in specifying cambium stem cells, we compared transcriptomes of wild type *Arabidopsis* seedlings undergoing cambium initiation to an overexpressor of TDIF (*p35S:CLE41*) and *pxy* mutant, characterized by enhanced or reduced TDIF-PXY signaling, respectively (*9*, *12*). In line with these phenotypes, GO terms under-represented in *pxy* included meristem maintenance; those over-represented in *p35S:CLE41* included meristem growth (**Fig. S1**). Transcripts under-represented in *pxy* and over-represented in *p35S:CLE41* included *PLETHORA 3* and *5* (*PLT3* and *PLT5*), members of the *AINTEGUMENTA-LIKE/PLT* (*AIL/PLT*) family (**fig. S1; and data S1**). Previously, different members of the AIL/PLT transcription factor family have been associated with promoting stemness and/or an undifferentiated state in apical meristems (*15*–*18*). In the cambium, *AINTEGUMENTA* (*ANT*) is specifically expressed in stem cells (*3*, *19*) and its absence leads to reduced radial growth (*19*, *20*). Thus, AIL/PLTs represent strong candidates for cambial stem cell regulators. Ectopic cell proliferation in *p35S:CLE41* occurs where xylem parenchyma cells reside in wild type. This phenotype was suppressed by *plt3plt5*, demonstrating that these genes are required for ectopic cambium proliferation in *p35S:CLE41* 

 (**Fig. 1, B and C**). Nevertheless, the *plt3plt5* line demonstrated no obvious cambium phenotype (**Fig. 1, B and C**). Thus, we investigated the AIL/PLT family for further redundancy. The AIL/PLT family consists of 8 members (*21*, *22*). Expression analysis of *AIL*/*PLT* fluorescent reporters showed that *ANT* (*3*), *PLT3*, *PLT5* and *PLT7* are expressed in the cambium, while *PLT1*, *PLT2* and *PLT4* expression appeared to be absent (**Fig. 1D; and fig. S2**). *ANT* was downregulated in the *pxy* mutant and *PLT7* upregulated in *p35S:CLE41* (**fig. S1C; and data S1**). Thus, we hypothesize that PLT3, PLT5, PLT7 and ANT act redundantly in cambium development downstream of TDIF- PXY signaling. Supporting this hypothesis, induction of *ANT*-*YFP* or *PLT5-YFP* under the *PXY* promoter restored cambium activity in a *pxy* mutant (**Fig. 1E**). While *ant* mutants are characterized by reductions in cambium activity (*19*, *20*), *PLT3*, *PLT5* and *PLT7* mutant combinations failed to show cambial phenotypes (**Fig. 1, F and G; and fig. S3C**). To determine potential redundancy between *ant* and *plt*s, several mutant combinations were analyzed. While *antplt5* roots demonstrated slight but significant reductions of secondary growth, quadruple mutant (*plt3plt5plt7-cr;ant-GK*) roots showed major reductions in secondary growth (**fig. S3, A and C**). Similarly, in *plt3plt5plt7-cr* (*23*) lines transformed with a gene editing construct to target *ANT*, secondary growth was strongly reduced in the majority of T1 individuals. However, *plt3plt5plt7- cr;ant-GK* and *plt3plt5plt7-cr;ant-cr* lines were unable to maintain a shoot apical meristem (**fig. S3B and fig. S4A**), as previously shown for *antplt3plt7* (*24*)*.* Thus, both to assess loss of the four *AIL*/*PLT* genes during cambium development, and to avoid secondary effects caused by shoot apical meristem loss, we generated an inducible genome editing (IGE) (*25*) construct targeting *ANT* (*ANT-IGE*). *ANT-IGE* was introduced to both the null *plt3plt5plt7-cr* (*23*) and *plt3plt5plt7- tdna* (*22*) backgrounds**.** Primary growth appeared normal in both conditional quadruple mutants, albeit *plt3plt5plt7-cr;IGE-ant* showed slightly reduced root length (**fig. S4A**). *plt3plt5plt7- tdna;IGE-ant* displayed radial sectors without differentiated secondary xylem vessels and reduced phloem sieve elements suggesting loss of cambium identity (**fig. S4C**). In the *plt3plt5plt7-cr;IGE- ant* null background, secondary growth was significantly reduced, and this was associated with a reduction in cambial cells per radial cell file, or occasionally, cell files with a complete loss of cambial cells (**Fig. 1, F and G; and fig. S4B**). These data demonstrate that the four CAILs - PLT3, PLT5, PLT7, and ANT - are critical in maintaining cambium identity.

 The expression of *CAIL*s is typically present in both daughter cells following recent divisions, which are identifiable by the presence of a thin cell wall (white arrowheads in **Fig. 2A**). As such, *CAIL* expression marks cambial stem cells and their daughters. In these daughter cells partial overlap between *CAIL*s and the neighboring early xylem and phloem identity reporters occurs (*3*) (**Fig. 2A; and fig. S5, A and B**). To investigate the role of CAILs in shaping these cell identities, we focused on PLT5 as a representative factor. To obtain a genome-wide view of PLT5 action in the cambium we performed RNA-seq on root tissues undergoing radial growth after 8 hours and 24 hours of induced overexpression of *PLT5*. We generated a 17-β-estradiol inducible (*26*) line, *35S:XVE>>PLT5-TagRFP,* for that purpose. Remarkably, 37% of xylem identity (*27*) and 53% phloem identity (*28*) genes were downregulated after 8 hours of *PLT5* induction, in comparison to 26% of all genes downregulated. Among core cell cycle genes (*29*) 55% were upregulated after 24 hours of PLT5 overexpression in comparison to 36% of all upregulated genes (**Fig. 2B; fig. S**6**, A and B; and data S1**). Thus, PLT5 regulates a large set of genes associated with cell proliferation and xylem and phloem formation, upregulating the former while downregulating the latter two.

 Next, we investigated the consequences of *PLT5* induction on morphogenesis. Short-term *PLT5* induction promoted ectopic DNA replication observed with EdU staining, accompanied by ectopic

cell divisions within xylem and phloem (**Fig. 2, D and E; and fig. S6C**). In support of our RNA-

 seq data (**Fig. 2B; fig. S6, A and B**), *PLT5* induction caused rapid down-regulation of xylem (*VND6*) and phloem (*PEAR1, APL*) reporter lines (**fig. S6D; and Fig. 2C**), leading to inhibition of xylem vessel and phloem sieve element formation (**Fig. 2, F and G**), and subsequently, inhibition of radial growth (**fig. S6E**). These data demonstrate that in the cambium, PLT5 maintains cell division capacity and the undifferentiated state of cambium cells. This occurs both through promoting cell division and through active opposition of differentiation to either xylem or phloem. Together, over-expression and loss-of-function analysis shows that CAILs are key cambium stem cell factors.

## **Computational model for cell fate determination in cambium**

 *CAIL* expression occurs only in a narrow stem cell domain of the cambium. This contrasts with the *PXY* receptor expression domain, which is strongest in the xylem/organizer domain tapering off towards weak expression in stem cells (*3*) **(Fig. 3A; and fig. S5C**). Even though cambium activity was restored in *pxy* by induction of *PLT5-YFP* or *ANT-YFP* in the *PXY* domain*,* xylem differentiation appeared perturbed (**Fig. 1E**). This demonstrates the importance of constrained CAIL expression in the stem cell, raising the question of how the downstream CAILs are constrained to the subdomain of low *PXY* expression. In any biological system, ligand binding to a receptor results in its sequestration from the pool of free ligands. *CLE41* expression and the subsequent TDIF peptide gradient extends from the phloem (*12*, *13*, *30*) (**fig. S5C**), thus the first PXY receptors that TDIF peptides encounter are those located at the lower end of the PXY gradient. Therefore, we hypothesized that sufficiently strong TDIF sequestration by PXY could abrogate further TDIF spread and thereby lead to a narrow active TDIF-PXY signaling domain and hence restrict *CAIL* expression to the stem cells. After a 24-hour application of synthetic TDIF peptide, the expression of *PLT5* and *ANT* expanded towards xylem parenchyma, coinciding with ectopic cell divisions (**Fig. 3A; and fig. S5D**), supporting the idea that excess TDIF prohibits sufficient sequestration. Long-term TDIF treatment led to further expansion of *PLT5* expression and cell proliferation in xylem parenchyma (**Fig. 3B**).

 To address whether the TDIF sequestration hypothesis could explain the above observations, we developed a computational model (codes are available on <https://tbb.bio.uu.nl/khwjtuss/cambium> as well as (31)). This model combined the regulatory interactions discovered here and those published previously (**Fig. 3C, top panel; and fig. S**7**)**, to determine the spatial patterning of the cambium and cells differentiating towards xylem and phloem fate **(Modelling Methods**). Previous models investigated cambium patterning dynamics while partly invoking hypothetical regulatory factors (*31*–*34*). Instead, we focused on testing the TDIF sequestration hypothesis while also incorporating the role of the newly identified CAIL factors by using a simple 1D static tissue model. The model incorporates TDIF-PXY promotion of *PLT5* (as a representative of the PLT subclade) and *ANT* (**Fig. 3A; and fig. S5C**); repression of *PXY* by PLT5 which we observed upon induction of *PLT5* (**Fig. 2B; and Fig. 3D**); auxin-mediated promotion of HD-ZIP III (*3*, *7*, *8*), ANT (*3*, *35*), and PXY (*3*); and HD-ZIP III promotion of xylem differentiation (*3*, *36*–*38*) and inhibition of phloem differentiation (*3*, *39*). By using *HD-ZIP III*-targeting inducible *miR165a* (*3*), we noticed that HD-ZIP IIIs repress *ANT* in xylem identity cells (**Fig. 3E**). Prior to testing the influence of TDIF sequestration by PXY in a multicellular setting, we first explored the capacity of this network to correctly assign cell fate identity in a single cell given various auxin and TDIF levels (for details on parameter sweep see **Modelling Methods**). Specifically, (i) cells experiencing high auxin and low TDIF levels should express HD-ZIP III and acquire xylem fate;

 (ii) cells with high TDIF and low auxin levels should obtain phloem fate; and (iii) cells with intermediate TDIF and auxin levels should have high ANT and PLT5 expression and thus cambium stem cell identity. Parameter sweeps in which simulations were performed until a steady state was reached showed that these requirements were met for a wide range of parameter values independently of specific fate determination threshold levels (**fig. S8, A-C; and Modelling Methods).** However, in the presence of high TDIF and high auxin levels, a more robust formation of the xylem occurs in conditions when we maximize the repression of ANT by HD-ZIP III (**fig. S8D; and Fig. 3C, bottom panel**).

#### **TDIF sequestration mechanism explains observed cambium phenotypes**

 From the parameter sweep we derived a final set of parameter values providing robust single cell patterning (Supplementary Tables S1-S3) with maximum HD-ZIP III-mediated ANT repression. Arabidopsis root cambium radial cell files typically consist of 2 to 6 cells (**Fig. 1, A and G**)(*3*), and cell type patterning has to be robust to variations in cambium width. Therefore, we extended our model to a row of 3-5 cells on which oppositely oriented auxin and TDIF gradients were superimposed. To decipher the role of gene regulatory network (GRN) architecture versus TDIF sequestration on cambium patterning robustness, we decoupled in our model the TDIF-PXY complex formation that induces ANT/PLT5 expression from the TDIF sequestration that limits the amount of TDIF available for diffusion (for details see Supplemental Methods). We started with model settings in which TDIF sequestration was ignored and TDIF-PXY binding was set to medium strength (*Kd* of 5). In a 3-cell tissue these settings readily resulted in correct patterning of the three cell types for variable TDIF gradient settings (**fig. S9A**). However, under 5-cell settings, overlap between the TDIF and PXY gradients was insufficient for TDIF-PXY to promote PLT5 expression, even when TDIF production was further elevated. Furthermore, this same elevated TDIF level caused invasion of PLT5 expression into the xylem in 3-cell settings (**fig. S9B)**. Clearly, increasing or decreasing TDIF expression would improve one issue at the cost of the other indicating a lack of robustness in cell type patterning across variable tissue widths. As a next step, we compared the patterning capacity of this strong ANT repression with that of strong TDIF 202 sequestration  $(K_d \text{ of } 1)$  (**fig. S9C and D**), both with default TDIF gradient parameters. In the strong TDIF sequestration scenario we halved ANT repression, thus reducing the impact of GRN architecture to enable us to focus instead on the effect TDIF sequestration has on patterning. (**fig. S10**; **Modelling Methods**). In contrast to the patterning issues, we observed with strong ANT repression, strong TDIF-PXY binding and TDIF sequestration ensured sufficient PXY-TDIF overlap under the 5-cell settings to promote PLT5 expression, and under 3-cell settings PLT5 invasion into the xylem was largely prevented. Nevertheless, under those settings, lack of strong HD-ZIP III-mediated repression on ANT allowed auxin to induce ANT in the xylem in the 3-cell settings. Combined, this suggests that both regulatory network and sequestration mechanisms must be active *in planta* to some extent. To test which of these mechanisms dominates patterning of the narrow cambial domain observed *in planta*, we sought to investigate both hypothesized mechanisms in perturbed conditions simulating patterning in a 4-cell tissue. One prominent difference between the modelled "maximum HD-ZIP III ANT repression", and "strong TDIF sequestration" parameter regimes was the predicted behavior of ANT. Under maximum HD-ZIP III-mediated ANT repression, the model predicted that decreased *PXY* expression would reduce active TDIF-PXY complexes and consequently PLT5 expression in the cambium (**Fig. 4A**). By contrast, in regimes where TDIF sequestration was dominant, the model predicted that decreased

 *PXY* expression resulted in xylemward expansion of TDIF gradient, leading to activation of further PXY receptors and hence ANT expression towards the xylem (**Fig. 4B**). To compare these differential predictions against *in planta* behavior, we lowered *PXY* levels by inducible RNA interference (*RNAi-PXY*) in an *ANT* fluorescent reporter background in Arabidopsis roots. Upon *RNAi-PXY* induction, *PXY* levels dropped to 71% of wild type levels (**fig. S6F**) and a shift in *ANT*  expression towards the xylem was observed (**Fig. 4, C and D; fig. S9E**), as the model predicted for the TDIF sequestration scenario.

 Further corroboration of the dominance of TDIF sequestration came from the predicted differential responses of the two modelled mechanisms to increased TDIF levels emanating from the phloem. In the maximum HD-ZIP III ANT repression mechanism, a modelled increase in TDIF levels resulted in little change to the *ANT* expression domain (**Fig. 4E**). By contrast, for TDIF sequestration, expansion of the *ANT* expression domain towards the xylem was predicted (**Fig. 4F**). Supporting the sequestration model, TDIF overproduction *in planta* via phloem precursor- specific *pPEAR1:XVE>>CLE41* line resulted in expansion of *ANT* expression into the xylem (**Fig. 4, G and H; fig. S9F**). Thus, while both mechanisms may co-exist *in planta*, our modelling and experimentation suggest that the sequestration mechanism provides the dominant patterning constraint. Thus, the sequestration of TDIF by its PXY receptor effectively constrains TDIF mobility to the first few PXY expressing cells, enabling robust, spatially constrained patterning of the cambium stem cells. To confirm this behavior, wild type and two signaling-impaired versions of PXY-YFP, kinase-dead (K747E) (*40*) and truncated (ΔKD), were generated (**Fig. 5A**). As expected, only the wild type PXY-YFP version under its own promoter complemented *pxy,*  indicating that the PXY-YFP fusion was functional (**fig. S11A**)*.* Next, in wild type, we expressed each of the three PXY-YFP versions in phloem precursors, the source tissue of TDIF, using a *PEAR1* promoter (**fig. S11B**). In all three instances this resulted in a *pxy* phenotype (**Fig. 5B**). Since all three versions contain an intact TDIF binding domain, these results suggest TDIF was sequestered in the source tissue, prior to meeting the endogenous PXY receptors in the cambium stem cells. Our data indicate that sustained TDIF-PXY binding underlies this strong sequestration, even in the absence of a functional kinase domain. Translational reporters (*pPXY:gPXY-YFP* and *pPXY:cPXY-YFP*) displayed greater signal asymmetry between the stem cell daughters than the transcriptional version (*pPXY:erVenus*), with signal stronger in the xylem-side daughter (**Fig. 5, C and D**). As the phloem-side daughter cell experiences higher TDIF levels than the xylem-side daughter cell, one explanation for the asymmetry is that PXY-YFP with sequestered TDIF is subject to turnover. While the exact mechanism for sequestration requires further investigation, internalization and subsequent ligand and receptor degradation has been observed in other similar receptor kinases (*41*–*44*).

## **Discussion**

 Our combined experimental and modelling approach shows how PXY-mediated TDIF sequestration generates a robust patterning mechanism. By manipulating the auxin and TDIF gradients, we observed that the balance between the auxin gradient and the TDIF gradient determines the localization of the cambial stem cells as well as cambium size (*6*) **(Fig. 4G, fig. S12)**. We propose that this patterning mechanism flexibly enables adjustment of both phloem to xylem ratio and overall growth. A dominance of auxin localizes the cambium stem cell phloemward, leaving room for xylem cells to differentiate (*6*) **(fig. S12**). A TDIF dominated condition localizes the cambium stem cells xylemward, allowing phloem cells to differentiate,

 although the underlying connection to phloem differentiation remains to be studied. A strong combination of the two gradients allows for a larger cambium that sustains a larger total cell production (*45*) (**Fig. 4G; and fig. S12**). A WUSCHEL-related HOMEOBOX gene, *WOX4*, has been shown to act downstream of TDIF-PXY signaling (*46*, *47*). It remains to be studied how WOX4 is integrated with the signaling network described here, particularly because *WOX4* has a broader expression domain than *CAIL*s in the cambium (this paper and (*3*, *19*)). Also, it remains to be determined through which mechanism TDIF-PXY promotes *CAIL* transcription. Previously, we discovered a stem cell organizer at the xylem side of the cambium, defined by high levels of auxin signaling (*3*). Here, CAIL transcription factors were identified as the key stem cell factors operating downstream of the auxin-regulated PXY receptor. We elucidated how through sequestering TDIF ligands on the edge of the PXY gradient, the auxin-promoted organizer can induce CAILs and perform cambial stem cell patterning at a distance. Our findings suggest that like animals, plants use opposing morphogen gradients fine-tuned by sequestration-based feedback mechanisms to control precise positioning and cell fate decisions.

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**Competing interests:** The authors declare no competing interests.

 **Data and materials availability**: All lines involved in this study are available upon reasonable request from the corresponding authors. Gene accession numbers are as follows: *ANT*, AT4G37750; *APL,* AT1G79430; *AtHB8*, AT4G32880; *CLE41*, AT3G24770; *MIR165A*, AT1G01183; *PEAR1*, AT2G37590; *PLT1*, AT3G20840; *PLT2*, AT1G51190; *PLT3*, AT5G10510; *PLT4*, AT5G17430; *PLT5*, AT5G57390; *PLT7*, AT5G65510; *PXY/TDR*, AT5G61480; *VND6*, AT5G62380; *PEX4/UBC21* AT5G25760. The data of transcriptomes of *pxy* and *35S:CLE41* is available on GEO (accession number [GSE119872\)](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE119872). The data of transcriptome of *35S:XVE>>PLT5-tagRFP* is available on GEO (accession number [GSE264403\)](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE264403). The source codes for the different models described here is available on <https://tbb.bio.uu.nl/khwjtuss/cambium> as well as in GitHub (31). 

- **Supplementary Materials**
- **This pdf includes:**
- Experimental Materials and Methods
- Modelling Methods
- Tables S1 to S6
- Figs. S1 to S12
- References (48–73)
- 
- **Other Supplementary Materials for this manuscript includes the following:**
- Data S1 (Source and RNA-seq data)
- Data S2 (Primers, constructs, seeds)
- 

## **Figures and Figure legends**



**Fig. 1. CAILs operate downstream of TDIF-PXY to maintain cambial stem cells**

 **(A)** Schematic representation of 14-day-old wild-type (Col-0) cross section of *Arabidopsis* root. **(B)** Plastic cross-sections of 14-day-old Col-0, *plt3plt5*, *p35S:CLE41*, and *plt3plt5;p35S:CLE41* roots. **(C)** Ratio between numbers of xylem parenchyma cells and secondary vessels. **(D)** Confocal cross-section of 14-day-old *pPLT5:erRFP* root. (**E**) Confocal root cross-sections of 10-day-old Col-0, *pxy*, and 5-day-old *pPXY:XVE>>ANT-YFP* and *pPXY:XVE>>PLT5-YFP* lines induced for 5 days in *pxy* background. The numbers in the top right corner of subpanels represent the frequency of the observed phenotype. Cell walls were stained with SR2200 (grey), and lignified cell walls were stained with 0.1% basic fuchsin (magenta). **(F)** Confocal root cross-sections of 15-d-old Col- 0, *plt3plt5plt7-cr* and *plt3plt5plt7-cr;IGE-ant*. The false coloring in the inset highlights the cambial cells (brown) used for the quantification in panel (G). **(G)** Boxplot showing the cambial cell number per cell file between differentiated vessels and sieve elements, as visualized in the insets of panel (F). Letters in (C and G) indicate significant differences using Kruskal-Wallis test

 followed by Dunn's post-hoc multiple comparisons with Bonferroni adjusted *p-values* (*p*< 0.05). White arrowheads mark stem cell divisions, black arrowheads mark ectopic divisions, red arrows

 mark the primary xylem axis, and the black dashed line marks the position of the cambium. Phloem (p), sieve element (se), xylem vessels (v), and xylem parenchyma (xp). Scale bars, 10 µm (B, D,

E, and F).



 **Fig. 2. PLT5 promote cambial cell divisions and inhibit secondary xylem and phloem differentiation**

 **(A)** Confocal cross-sections of 12-day-old *pPLT5:erRFP pATHB8:erVenus* (left) and *pPLT5:erRFP pPEAR1:erVenus* (right) double marker lines. **(B)** Heatmap showing normalized log2FoldChange (FC) as determined by RNA-seq, upon induction of *PLT5* over-expression for 8 hours and 24 hours. **(C)** Lateral view of 8-day-old *35S:XVE>>gPLT5-TagRFP* roots with early (*pPEAR1:erVenus*) and late (*pAPL:3xYFP)* phloem markers after a 2-day induction. Box plots show quantification of relative fluorescent signal intensity. **(D)** Confocal cross-sections of  *35S:XVE>>gPLT5-TagRFP* after 2-day induction with 16 hours of EdU (green) incorporation (in 12-day-old roots) to detect S-phase nuclei. **(E)** Barplot of relative distribution (RD) of EdU-595 positive nuclei along the radial root axis in panel (D) (primary xylem axis = 0; root surface = 1). 596 Scale shown on green on the mock panel;  $n_c$  = number of cross-sections analyzed;  $n_n$  = total number of nuclei. **(F)** Cross-sections of *35S:XVE>>gPLT5-TagRFP* after 4-day induction (in 8-day-old plants). Red dots indicate the sieve elements and blue dots indicate vessels in the lower half of the root. **(G)** Quantification of secondary xylem vessel and sieve element numbers. In (E) error bars 600 show mean  $\pm$  standard error. Significant differences were calculated using two-tailed student t-test 601 for each relative distance class ( $ns = p > 0.05$ ). In (C) and (G) significant differences between mock and induced conditions were calculated using two-sided Wilcoxon-Mann-Whitney test. White arrowheads mark recent cell division, red arrows mark the primary xylem axis. Vessels (v), phloem 604 (p), sieve elements (se). Scale bars,  $10 \mu m(A)$ ,  $100 \mu m(C)$ , and  $50 \mu m(D)$  and F).



#### **Fig. 3. Regulatory network defining cambium, xylem and phloem**

 **(A)** Confocal cross-sections of roots carrying *pPLT5:erRFP* and *pPXY:erVenus* double markers in 609 15-day-old plants after 24 hours of P9A $_{(n=18)}$  (control) or TDIF $_{(n=16)}$  treatment. P9A is a mutant version of TDIF, incapable to bind PXY. While wild-type or P9A-treated cambia typically have a single cell in every radial cell file that has undergone a recent division (thin cell wall, marked with white arrowhead), TDIF-treated cambium have additional, ectopic divisions in the xylem domain (multiple arrowheads). Quantification of *pPLT5:erRFP* expression (right) shows increases and a shift towards xylem. (ns = *p*>0.05). **(B)** Confocal cross-sections of *pPLT5:erRFP* expression after 4-day TDIF treatment (in 13-day-old root). White dashed line marks the cambium. Yellow arrows mark the primary xylem axis. **(C)** Modelling cambial cell fate decision making under different combinations of auxin and TDIF concentrations. Top panel: Interaction network underlying cambial cell fate decision making incorporated in the model. PE (yellow) refers to 'Published Elsewhere', TS (orange) refers to 'This Study'. Supporting details with references are listed in (fig. S7). Bottom panel: Fate map of a cambial cell exposed to different auxin-TDIF combinations. For each Auxin-TDIF combination, the left square within the rectangle shows the HD-ZIP III level at steady state and the right square shows the summed PLT5+ANT level at steady state. The black arrowheads on the color bars in the top panel show the threshold levels. Applied cell fate thresholds 624 are as follows: If PLT5+ANT  $\ge$  75 cambial identity (grey); if PLT5+ANT < 75 and HD-ZIP III  $\ge$  = 30 xylem identity (blue); if HD-ZIP III < 30 phloem identity (green). Parameter settings are described in the Modelling Methods (maximum HD-ZIP III ANT repression settings). While in the parameter sweep, TDIF levels were varied between 0 and 100, here TDIF levels between 0 and 70 are shown to focus on the key region in terms of cell fate acquisition. **(D)** Lateral view of *35S:XVE>>gPLT5-TagRFP;pPXY:erVenus* 8-day-old plants after a 2-day induction with quantification of relative fluorescent signal intensity. Significant difference between mock and induced condition was calculated using two-sided Wilcoxon-Mann-Whitney test. **(E)** Confocal cross-sections of *35S:XVE>>miRNA165a;gANT-3xYFP* after 2-day induction (in 12-day-old plants). Quantification of positional expression of *gANT-3xYFP* in cambium. Barplot showing position of *gANT-3xYFP*-positive nuclei. In (E) categorical distribution was tested using a chi- square test. Data points represent the mean of each biological repeat. White arrowheads mark recent cell divisions. Vessels (v). Scale bars, 10 µm (A, B and E), 200 µm (D).



637<br>638 638 **Fig. 4. TDIF sequestration ensures a narrow** *ANT* **expression domain** 

 **(A, B, E** and **F)** Comparison between the maximum HD-ZIP III repression parameter regime (A and E) and strong sequestering effect of PXY on TDIF (B and F). Heatmap shows the gene expression level. Parameter settings are described in the Modelling Methods (maximum HD-ZIP III repression versus strong sequestration settings). (A and B) Effect of reducing *PXY* expression rate in the two regimes. In (A) this leads to overall reduction of *ANT* expression, while in (B) it leads to a xylemward shift of *ANT* expression. (E and F) Responses of the two regimes to incorporating various TDIF production rates, expressed as percentage of default phloem production rate, in all cells. Maximum HD-ZIP III repression safeguards the xylem from ANT expression, while in the strong sequestering regime *ANT* is expressed in the xylem. **(C)** Confocal cross-sections of *35S:XVE>>PXY-RNAi; pANT:erRFP* after 2-day induction (in 12-day-old plants). **(D)** Quantification of mean *pANT:erRFP* intensity position. **(G)** Confocal cross-sections *pPEAR1:XVE>>CLE41;pANT:erRFP* after 2-day induction (in 12-day-old plants). **(H)** Quantification of mean *pANT:erRFP* intensity position. In (D and H) error bars show ± standard 652 error. Significance was determined using two-sided Wilcoxon-Mann-Whitney test,  $(ns = p>0.05)$ . White arrowheads mark recent cell divisions, orange arrows mark the position of xylem identity cells before induction. Xylem identity cell was numbered as position 0, towards xylem (positions 655 -1, -2 and -3) and towards phloem (positions 1, 2..., and 8). Vessels (v). Scale bars 10  $\mu$ m (C and G).



### **Fig. 5 TDIF peptide is sequestered by PXY protein**

- (**A**) Schematic representation of PXY translational reporters. PXY protein consists of a Leucine-
- rich repeat domain containing the TDIF binding motif, a transmembrane domain (TMD) and
- kinase domain (KD). (**B**) Confocal root cross sections of *pxy* and 5-day-old
- *pPEAR1:XVE>>gPXY-YFP*, *pPEAR1:XVE>>gPXYK747E-YFP* and *pPEAR1:XVE>>gPXY<sup>∆</sup>KD -*
- *YFP* lines induced for 5 days each of which showed a *pxy*-like stem cell differentiation
- phenotype with vessels (v) adjacent to a sieve elements (se), by contrast to the mock. (**C**)
- Confocal root cross-sections of 14-day-old *pPXY:erVENUS*, *pPXY:gPXY-YFP* and *pPXY:cPXY-*
- *YFP.* Phloem (p), xylem vessels (v). (**D**) Boxplot showing the ratio of mean signal intensities
- between xylem side (ii) and phloem side (i) daughter cells of the cambial stem cell within radial
- cell files. Letters in (D) indicate significance using one-way ANOVA with a Tukey post hoc test.
- Red arrows mark the primary xylem axis. White arrowheads mark recent cell divisions. Scale
- 672 bars 10  $\mu$ m (B and C).



#### **Materials and Methods**

#### Plant material and cloning

 The Col-0 background was used throughout. *pxy* (*9*), *35S:CLE41* (*12*)*, plt3plt5* (*plt3-1, plt5-2*) (*48*)*, ant- GK* (GK-874H08) (*20*), *plt5* (*cho1*) (*49*), *plt3plt5plt7-cr* (*23*), *plt3plt5plt7-tdna* (*plt3-1*, SALK\_127417, *plt5-2*, SALK\_059254; *plt7-1*, SAIL\_1167\_C10) (*22*), *gPLT1-YFP* (*18*), *gPLT2-YFP* (*18*), *gPLT4-YFP* (*18*), *gPLT7-YFP* (*22*), *gVND6-GUS* (*50*), *pAPL:3xYFP* (*51*) have been described previously. These lines together with the new lines generated in this study are listed in **Data S2**.

All entry clones were generated by PCR amplification of the target sequence. PCR products were then

712 recombined into MultiSite Gateway compatible pDONR entry vectors using a BP clonase reaction or using<br>713 Golden Gate assembly cloning methods. Multisite Gateway technology was used to assemble the entry

Golden Gate assembly cloning methods. Multisite Gateway technology was used to assemble the entry

714 clones into Gateway compatible binary vectors using multisite gateway LR reactions. Primers for PCR<br>715 amplification, pDONR entry vectors, and expression vectors are listed in the **Data S2**. amplification, pDONR entry vectors, and expression vectors are listed in the **Data S2**.

 Upon plant transformation of expression vectors, putative single insertion lines were identified based on 717 Mendelian segregation of the selectable marker. Multiple single insertion lines were screened for each construct to observe the most consistent expression patterns or phenotypes.

construct to observe the most consistent expression patterns or phenotypes.

To obtain *plt3plt5plt7ant* quadruple mutants, *plt3plt5plt7-cr* (♀) was crossed with *ant-GK* (♂). *PLT5* and

*PLT7* are located in the long arm of chromosome 5. Thus, we expected linkage between *PLT5* and *PLT7*.<br>**721** 140 F2 seedlings were germinated for genetic analysis. Since *plt3plt7* double mutants are unable to form

721 140 F2 seedlings were germinated for genetic analysis. Since *plt3plt7* double mutants are unable to form<br>722 lateral roots (48), and *antolt3plt7* has defects in SAM maintenance (24), we genotyped only those F2 lateral roots (*48*), and *antplt3plt7* has defects in SAM maintenance (*24*), we genotyped only those F2

723 seedlings lacking lateral root formation and showing defects in SAM (altogether 3 individuals). These 3<br>724 seedlings were genotyped as *plt3plt5plt7-cr; ant-GK*. To analyze the root cambium phenotype of

seedlings were genotyped as *plt3plt5plt7-cr;ant-GK*. To analyze the root cambium phenotype of *plt3plt5plt7-cr;ant-GK* quadruple mutant, we germinated 618 F2 seedlings and identified additional 7

726 seedlings with defects in lateral root formation and SAM maintenance. These 7 seedlings underwent 727 anatomical analysis. anatomical analysis.

 To generate multiple independent quadruple mutants, we transferred a CRISPR construct targeting *ANT* (*ant-cr*) into *plt3plt5plt7-cr* and analyzed T1 seedlings. This CRISPR construct contained two sgRNAs designed to create a large deletion in *ANT*. The primers used to generate two fusions of a small nuclear RNA promoter and sgRNA (*pAtU3-sgRNA*) are listed in the Data S2. The resulting PCR products were 732 cloned into a *p2PR3-Bsa I-ccdB-Bsa I* entry vector using Golden Gate and Gibson assembly cloning<br>733 methods thus concatenating the two *pAtU3b-sgRNA1-ANT and pAtU3b-sgRNA2-ANT* fragments. The methods thus concatenating the two *pAtU3b-sgRNA1-ANT and pAtU3b-sgRNA2-ANT* fragments. The *zCas9i* (*52*) was cloned into *p221z*-*Bsa I-ccdB-Bsa I* using Golden Gate and the primers used to generate *zCas9i* PCR product are listed in the **Data S2**. The final binary vector was generated in a single MultiSite Gateway LR reaction by combining a RPS5A promoter, *221z-zCas9i*, *2R3z-2x-pAtU3b-sgRNA-ANT* and

destination vector pFRm43GW (*25*).

 To generate conditional *plt*/*ail* quadruple mutants, we took an advantage of an inducible genome editing system (IGE) (*25*). An inducible CRISPR construct targeting *ANT* was transformed into two different *plt3plt5plt7* mutant backgrounds, a null combination *plt3plt5plt7-cr* (*23*) and a weaker allele combination *plt3plt5plt7-tdna* (*22*). This IGE vector contained the same two sgRNAs as described above to create a

large deletion in *ANT* upon treatment with 17-β-estradiol. The IGE binary vector was generated in a single

MultiSite Gateway LR reaction by combining a 17-β-estradiol-inducible *WOODEN LEG* (*WOL*) promoter

744 (26), which drives expression in the primary vascular cylinder (53), giving rise to all the secondary tissue<br>745 in the root (3),  $Cas9p$ ,  $2R3z-2x-pAtU3b-sgRNA-ANT$  and destination vector pFRm43GW (25). in the root (*3*), *Cas9p*, *2R3z-2x-pAtU3b-sgRNA-ANT* and destination vector pFRm43GW (*25*).

746 To facilitate screening of transformed seeds, we used seed-specific RFP fluorescence provided by<br>747 pFRm43GW. The plt3plt5plt7-cr; IGE-ant and plt3plt5plt7-tdna; IGE-ant lines were germinated directly in

*pFRm43GW*. The *plt3plt5plt7-cr;IGE-ant* and *plt3plt5plt7-tdna;IGE-ant* lines were germinated directly in

1/2 GM plates supplemented with 5 µM 17-β-estradiol and grown for 15 days and 13 days respectively

alongside Col-0 and their respective *plt3plt5plt7-cr, plt3plt5plt7-tdna* controls.

- 750 To generate the translational reporter of PXY, full-length genomic DNA and cDNA of *PXY* without stop
- 751 codons were amplified with primers BsaI-PXY-F and BsaI-PXY-R. YFP with a stop codon was amplified
- 752 with primers BsaI-VenYFP-F and BsaI-VenYFP-R. The resulting genomic PXY (gPXY) or cDNA PXY
- 753 (cPXY) PCR products together with the YFP PCR product were cloned into the *221z*-*Bsa I-ccdB-Bsa I* by
- 754 Golden Gate cloning. The *pPXY:gPXY-YFP* and *pPXY:cPXY-YFP* binary vectors were generated by
- 755 combining 1R4z-pPXY, 221z-gPXY-YFP or 221z-cPXY-YFP, 2R3a-3AT and destination vector
- 756 pFRm43GW in a single MultiSite Gateway LR reaction.
- To generate the translational reporter of PXY with a K747E mutation, a  $PXY^{K747E}$  mutation was introduced<br>T58 by primers. 221z-gPXY-YFP was used as template and amplified with two pairs of primers Bsal-PXY-F +
- by primers. 221z-gPXY-YFP was used as template and amplified with two pairs of primers BsaI-PXY-F +
- 759 BsaI-PXY<sup>K747E</sup>-R and BsaI-PXY<sup>K747E</sup>-F + BsaI-VenYFP-R. These two PCR products were cloned into 760 221z-*Bsa I-ccdB-Bsa I* by Golden Gate cloning, resulting 221z-gPXY<sup>K747E</sup>-YFP. *pPXY:gPXY<sup>K747E</sup>-YFP*
- 761 binary vectors were generated by combining  $1R4z-pPXY$ ,  $221z-qPXY^{K747E}-YFP$ ,  $2R3a-3AT$  and
- 762 destination vector pFRm43GW in a single MultiSite Gateway LR reaction.
- 763 To generate the reporter for truncated PXY without kinase domain  $PXY^{\Delta KD}$ , a 2076 bp PXY fragment
- downstream of the PXY start codon was amplified with primers BsaI-PXY-F and BsaI- PXY<sup>ΔKD</sup>-R. The
- PCR product was inserted into 221z-*Bsa I-ccdB-Bsa I* by Golden Gate cloning, resulting 221z-PXY<sup>ΔKD</sup>.
- *POS pPXY: gPXY<sup>ΔKD</sup>-YFP* binary vector was generated in a single MultiSite Gateway LR reaction by combining
- 1R4z-pPXY, 221z-gPXY<sup>ΔKD</sup>, 2R3a-VenYFP-3AT and destination vector pFRm43GW.
- 768 For *pPEAR1:XVE>>gPXY-YFP* and *pPEAR1:XVE>>gPXY<sup>K747E</sup>-YFP*, entry clones 1R4a-pPEAR1:XVE,
- 769 221z-gPXY-YFP or 221z-gPXY<sup>K747E</sup>-YFP and 2R3a-3AT were incorporated into pCAM-kan-R4R3 by LR 770 reaction, respectively.
- For *pPEAR1:XVE>>gPXY<sup>∆KD</sup>-YFP*, entry clones 1R4a-pPEAR1:XVE, 221z-gPXY<sup>∆KD</sup>, 2R3a-VenYFP-
- 772 3AT were incorporated into pCAM-kan-R4R3 by LR reaction.
- 773

## 774 Plant growth and chemical treatments

- 775 All the plants were grown vertically in a plate in a  $23^{\circ}$ C growth chamber with 8 hours dark & 16 hours light 776 cycle. Seeds were surface sterilized using 70% ethanol with Tween-20 (µl/ml) solution for 5 min with 777 vortexing, followed by five washes in sterile Milli-Q (MQ) water. The sterilized seeds were stratified for 2 778 days at  $4^{\circ}$ C in darkness before plating them on  $1/2$  germination medium (GM) containing 0.5x Murashige 779 and Skoog (MS) media with vitamins (Duchefa), 0.8% plant agar, 1% sucrose and 0.5 g/l MES pH 5.8. 780 Alternatively, 0.5x MS (pH 5.8), 1% sucrose and 1% agar was used. The age of the plants was measured 781 from when the plates were vertically positioned in the growth cabinet.
- 782 Aqueous 10 mM stocks of P9A and TDIF peptides (GeneCust) were prepared and stored at the -80°C.<br>783 10 mM stocks of EdU (Thermo Fisher), dissolved in dimethyl sulfoxide (DMSO) was prepared and stored 10 mM stocks of EdU (Thermo Fisher), dissolved in dimethyl sulfoxide (DMSO) was prepared and stored 784 at -20°C. 17-β-estradiol (EST) (Sigma), was prepared as a 20 mM stock solution in DMSO and stored at - 785 20°C.
- 786 Short term (24 hours or less) TDIF and P9A treatments were performed in liquid 1/2GM containing the 787 respective peptide with the working concentration of 10 µM. Longer treatments were performed on 1/2GM 788 plates. 17-β-estradiol induced gene expression was achieved by transferring plants onto plates containing<br>789 5 μM 17-β-estradiol or an equal volume of DMSO as a mock treatment, except in Fig. 2F and fig. S6E and 789 5 µM 17-β-estradiol or an equal volume of DMSO as a mock treatment, except in **Fig. 2F and fig. S6E** and
- 790 *35S:XVE>>PLT5-TagRFP* RNA-Seq where 1 µM 17-β-estradiol was used. For EdU incorporation, plants
- 791 were placed in liquid 1/2GM containing 10 µM EdU for 16 hours prior to fixation.
- 792
- 793 RNA-Seq profiling and data analysis

794 Transcriptomes of *pxy* and 35S: CLE41 in comparison to wild type were determined from seedlings grown<br>795 on vertical plates for 7 days. Upon harvesting, seedlings were separated into root and shoot samples by on vertical plates for 7 days. Upon harvesting, seedlings were separated into root and shoot samples by 796 separating plants at the root-hypocotyl junction. RNA and library preparation in biological quadruplicate<br>797 was performed as described (54). 50 bp single end reads were obtained on the Illumina HiSeq 4000 was performed as described (*54*). 50 bp single end reads were obtained on the Illumina HiSeq 4000 platform. Sequencing was performed by the QB3 Genomics Facility, University of California, Berkeley. 799 Quality checking and trimming of the raw FASTQs was performed with Kraken (55) followed by alignment<br>800 to TAIR10 with Tophat2 (56). An average of 3.4M reads were obtained per treatment. Gene counting was to TAIR10 with Tophat2 (*56*). An average of 3.4M reads were obtained per treatment. Gene counting was performed with HTSeq (*57*), and differential gene expression was determined with DESeq2 (*58*). Cut-offs 802 for differential expression was an adjusted p value < 0.05. The data is available on GEO (accession number [GSE119872\)](https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE119872). DEGs are listed in **Data S1**. GO analysis was performed on genes differentially expressed in either 35S: CLE41 or *pxy* relative to wild type (Col-0). either 35S: CLE41 or *pxy* relative to wild type (Col-0).

For PLT5 transcriptome analysis, *35S:XVE>>PLT5-TagRFP* seeds were germinated on 1/2 GM plates for

 9 days or 9 days 16 hours, and then transferred to 1 µM 17-β-estradiol or DMSO plates for 24 hours or 8 hours, respectively. For each sample, 1.5 cm of root segments below the root-hypocotyl junction were

collected from 15 individuals. Visible lateral roots were removed. RNA isolation, library preparation and

data analysis were done as previously described (*59*) except single-end reads (86bp) and were mapped to

*Arabidopsis* reference genome (TAIR 10.39). Differential expression between the mock and inductions was

analyzed using the edgeR package (*60*). Subsequently, Pvalue < 0.05 was applied to identify differentially

expressed genes (DEGs). The data is available on GEO (accession number GSE264403). DEGs are listed

- in **Data S1**.
- 
- 815 Thin sections, GUS staining, and light microscopy

All the root samples were sectioned 5mm below the hypocotyl junction unless mentioned otherwise.

 Samples were fixed overnight in 1% glutaraldehyde, 4% formaldehyde in 0.05 M sodium phosphate buffer pH 7.2, followed by dehydration through an ethanol series, and embedding in plastic resin using either

819 Historesin (Leica) or JB4 (Polysciences). 3, 5 or 10  $\mu$ m sections were cut with either an RM2055 microtome<br>820 (Leica) using a microtome knife, or a Shandon Finesse E+ microtome (Thermo) using a glass blade.

- (Leica) using a microtome knife, or a Shandon Finesse E+ microtome (Thermo) using a glass blade.
- 821 Sections were stained with either double staining of 0.05% ruthenium red (Sigma-Aldrich) and toluidine

 blue (Sigma-Aldrich; 5s in each respectively, rinsed between staining's and afterwards with water), or 0.025% aqueus toluidine blue for 30s. Sections were mounted in either water or Histomount (National 824 Diagnostics) and visualized either with a Leica 2500 Microscope or Zeiss Axioskop using 20x or 40x<br>825 objectives.

objectives.

 For GUS-stained samples, Historesin was used. The GUS-staining protocol was adapted from (*61*). 827 Samples were held in GUS-staining solution at 37°C until the appropriate staining level was reached prior to fixation. to fixation.

- 
- 
- 830 Fluorescent marker analysis, vibratome sections and EdU detection
- Lateral view of the fluorescent samples were analyzed in plates using Leica MZ165FC microscope, Hamamatsu C11440 digital camera, and Leica LAS X program.
- 
- 833 For cross sections fluorescent samples were fixed as previously described (3), prior to embedding in 4% agarose. Agarose blocks were cut with the vibratome into 200  $\mu$ m sections for confocal analysis. Sections agarose. Agarose blocks were cut with the vibratome into 200 µm sections for confocal analysis. Sections
- were placed in PBS and stained with SR2200 (1:1000, Renaissance Chemicals) for cell wall staining.
- Lignified cell walls were stained with Basic Fuchsin, as previously described (*62*).
- To visualize the EdU positive nuclei after EdU incubation, The Click-iT EdU Alexa Fluor 488 Imaging Kit
- (Thermo Fisher) was used for detection with a modified EdU detection mix (*63*). Samples were incubated
- in the detection mix for 1h and then transferred into PBS with SR2200 (1:1000).

#### 841 Confocal microscopy and image processing

 Confocal imaging was performed on PBS-mounted samples with a Stellaris 8 confocal microscope, except *35S:XVE>>gPLT5-TagRFP* analysis was carried out with Leica SP5 (20x and 63x objectives; Leica). Images were obtained using Las AF software (Leica). Samples visualized with multiple channels were imaged in the sequential scan mode. Confocal settings vary between experiments but were constant within experiments. The exception to this was cell wall staining to aid visualization. Here, SR2200 (cell wall) signal was adjusted during imaging (but not the fluorophore of interest) and therefore SR2200 settings varies between the sample and respective control.

- 
- Image projections
- Circle unwrapping projections (**Fig. 3A**) were performed as previously described (*6*).
- 
- Image analysis

 Image analysis and quantification were performed using Leica AF Lite 2.6.x, LithoGraphX 1.2.2 with Builder 1.2.2.7, and FIJI ImageJ v1.52 (*64*). For **Fig. 1C**, Cell numbers were calculated within 35 µm diameter area (primary xylem was in the center of this area). For image quantification in **Fig. 2E**, the distances of EdU-labelled nuclei were measured from the central point of each cross-section, as previously described (*3*). Locations of EdU-positive nuclei were expressed as a relative position along the radii of 859 cross-sections. For each sample, the frequency distribution of EdU-labelled nuclei was then calculated and<br>860 assigned into one of ten classes based on their relative positions in the cross-section. For each independe assigned into one of ten classes based on their relative positions in the cross-section. For each independent repeat, the mean and standard error of the frequency distributions of different cross-sections were calculated and plotted according to the treatments applied. To identify significant differences in the frequencies per distance class between the mock and induced conditions, Student's t-tests were applied.

 Operational definition of cambial stem cells relied on morphological features of the tissue, supported by previous lineage tracing analysis which demonstrated that transit amplifying divisions occurred rarely in 866 this tissue (3). Therefore, the vast majority of divisions were derived from the cambium stem cell. Cambium stem cells were thus considered to be those with thin cell walls suggesting a very recent deposition of stem cells were thus considered to be those with thin cell walls suggesting a very recent deposition of phragmoplast. Subsequent experiments demonstrated that this feature co-occurred with CAIL expression (Fig. 1D for *PLT5*; fig. S2 for *ANT*, *PLT3* and *PLT7*). However, where *PLT5* was over-expressed or induced by TDIF (and in any other CAIL manipulation data e.g. **fig. 2F, fig. 3B, fig. 4G**), recent cell divisions, as judged by the presence of thin cell walls, was considered to be cambial cells, not stem cells. This is due to 872 inability to define stem cells in cases when stem cell regulators, CAILs, are manipulated. In future studies, characterization of CAIL-independent stem cell markers will enable the precise identification of the cell characterization of CAIL-independent stem cell markers will enable the precise identification of the cell 874 identities in the CAIL manipulation lines.

 Number of cambium cell layers in **Fig. 1F, 1G** refers to number of cells in a given radial cell files between recently differentiated sieve elements and vessels. We omit quantifying radial cell files which did not contain recently differentiated sieve elements or vessels, or poor image quality prevented us to make quantification. For quantification of **fig. S3A**, **fig. S3C**, **fig. S4B** and **fig. S4C**, the diameter indicates the distance between two phloem poles.

 For quantification of *plt3plt5plt7-tdna;IGE-ant* root cross-sections in comparison to wild type and *plt3plt5plt7-tdna* controls **(fig. S4C)**, sectors in the upper panel lacking secondary xylem vessel differentiation were considered as mutant sectors as such sectors were absent in controls. The lower panel considers secondary phloem differentiation around primary phloem pole. Controls developed secondary phloem around primary phloem pole, but sectors in *plt3plt5plt7-tdna;IGE-ant* plants with reduced

- 885 secondary xylem formation produced also less secondary phloem (quantified as number of sieve elements)<br>886 were considered to be mutant sectors. were considered to be mutant sectors.
- For quantification of *gANT-3xYFP* fluorescence in **Fig. 3E**, only non-xylem-pole pericycle cell lineages
- were considered. The first, second and third nearest-neighboring cambial cell to vessels were defined as
- position 1, 2 or 3, respectively and assigned as *gANT-3xYFP*-positive or negative depending on the presence or absence of signal. To determine the presence or absence of signal, nuclei with YFP signals were extracted
- 891 using watershed segmentation and thresholds for lower signal intensity, nuclear size (more than 2  $\mu$ m<sup>2</sup>),
- and circularity (0.5-1). Extracted nuclear images were compared with the original images to assess the
- correctness of the chosen parameters. Occasionally, watershed failed to separate neighboring nuclei or over-
- 894 segmented a single nucleus. In those cases, the extracted nuclei images were manually corrected.
- For quantification of *pANT:erRFP* fluorescence in **Fig. 4D**, the signal intensity in one cell file was measured 896 from the outermost xylem vessel phloemward. For **Fig. 4H**, a line was drawn along the tangential axis of the root cross section between the outer edges of vessels. Cells between these two vessels on this line were the root cross section between the outer edges of vessels. Cells between these two vessels on this line were considered to be at position zero. The radial cell file phloemward from this cell zero were marked as 899  $+1,+2,+3...+8$ ; and towards xylem as  $-1,-2,-3$ .
- For quantification of *pPXY:erVENUS*, *pPXY:gPXY-YFP* and *pPXY:cPXY-YFP* in **Fig. 5D**, the mean signal
- intensity of each cambium stem cell daughter cells in one radial cell file was measured with Fiji. In one
- radial cell file, the ratio is calculated with the mean signal intensity of xylem-side daughter cell (ii) divided
- with the mean signal intensity of phloem-side daughter cell of stem cell (i).
- 
- RT-qPCR
- 11-day old plants with inducible *PXY* RNAi expression in *ANT* fluorescent reporter backgrounds (*35S:XVE>>PXY-RNAi; pANT:erRFP*) were transferred to 1/2GM plates containing either 5 µM of 17-β- estradiol (induced) or DMSO (mock) for three days. For RNA purification, 2 cm of primary roots 0.5 cm below the root-hypocotyl junction were harvested from ≥10 individuals for induced- or mock-treated plants. Total RNA from root samples was purified with RNeasy Plant Mini Kit (QIAGEN) with an on-column 911 DNase I treatment (QIAGEN). cDNA was synthesized using the iScript™ cDNA Synthesis Kit (Bio-Rad) 912 following the manufacturer's instructions. qRT-PCR experiments were carried out in 10 µl reaction volume using the LightCycler 480 SYBR Green I Master Mix (Roche Life Science) in a CFX384 Touch Real-Time PCR instrument (Bio-Rad). The PCR programme included an initial denaturation step at 95°C for 5 min, 915 then 45 cycles of (95°C for 10s, 59°C for 10s, 72°C for 15s), followed by a melting curve analysis. Each sample was run three times. Expression levels were normalized using the comparative CT Method (ΔΔCT 917 method) against the UBC21 reference gene expression (65). All primers used in qRT–PCR are listed in 918 Data S2. Data S2.
- 

## General methodology and statistical analysis

921 All experiments were repeated at least two times. We excluded samples that germinated poorly, or showed<br>922 overall growth defects that were confirmed genetically not to be related to the genotype. The number of overall growth defects that were confirmed genetically not to be related to the genotype. The number of individual roots analyzed is shown as *n* in the figures or figure legends, except for **Fig. 1G, Fig. 4, D, H, Fig. 5D**, *n* represents radial cell file. The fraction in the corners of some figures indicates the frequency of the observed phenotype. Before assessing statistical analyses, normality of residues distribution and variances homoscedasticity were checked using Shapiro's and Levene's tests, respectively, to determine 927 the type of statistical analyses that can be used for each quantitative dataset. Accordingly, Wilcoxon-Mann-<br>928 Whitney test was used to assess mean comparison for the Fig. 2, C and G; Fig. 3, A and D; and Fig. 4, D Whitney test was used to assess mean comparison for the Fig. 2, C and G; Fig. 3, A and D; and Fig. 4, D **and H** while student t-tests were used for the **Fig. 2E and fig. S3A**. For multiple comparisons, Kruskal-Wallis and Dunn's post-hoc tests were used for the **Fig. 1, C and G**. For the **Fig. 3E**, categorical distribution

931 was tested using Chi-square test. One-way ANOVA with a Tukey multiple comparisons test was used for 932 Fig. 5D, fig. S3C and fig. S4B. All the p-values for the different statistical comparison are available in the **Fig. 5D, fig. S3C and fig. S4B**. All the p-values for the different statistical comparison are available in the 933 data supplemental information. Specific tests are detailed in the figure legends. All statistical analysis were<br>934 performed in the R studio version 2023.06.0-421 and GraphPad. For boxplots, the central line indicates performed in the R studio version 2023.06.0-421 and GraphPad. For boxplots, the central line indicates the median; the bounds of the box show the 25th and 75th percentiles; and the whiskers indicate maximum and 936 minimum values (the values out of the whiskers are outliers). For bar plots, the bar height and the error bar represent the mean and the standard error of the mean, respectively, except **fig.** S6F, the error bar indica represent the mean and the standard error of the mean, respectively, except **fig. S6F**, the error bar indicates 938 standard deviation. For plots showing quantitative data, every individual data point is shown on top of the 939 plots. For the Fig. 3A comparisons of the distance between P9A and TDIF conditions in both  $pPLT5:erRFP$  plots. For the **Fig. 3A** comparisons of the distance between P9A and TDIF conditions in both *pPLT5:erRFP* 940 and *pPXY:erVenus* were obtained by sub-setting all the negative distances values (xylem region) present around the mean ± se of the fluorescence peak in P9A samples (data available in the data S1). *p-values*  shown in the figure represent comparison of the mean distances between P9A and TDIF conditions using

two tailed Wilcoxon-Mann-Whitney test.

#### **Modelling Methods**

#### Model network architecture and dynamics

 In this paper we employ models of the radial patterning of the secondary tissue, using both a single cell and multicellular model setup. In the latter case, because of the rotational symmetry, we restrict ourselves to modeling vascular patterning in a single 1D cell file of the differentiating root vasculature. In our model all cells contain the same gene regulatory network describing the dynamics of auxin, TDIF, PXY, ANT, PLT5, 951 and HD-ZIP III concentrations. We took PLT5 as a representative of PLT3, PLT5 and PLT7, based on their 952 presence in the same PLT subclade and their highly redundant functionality. In contrast, since our data presence in the same PLT subclade and their highly redundant functionality. In contrast, since our data indicated a differential regulation of ANT as well as differential downstream effects of ANT, ANT was modeled separately. Note that we use [] to signify concentrations, and that concentrations in our model are in dimensionless units.

 Dynamics of the individual players making up this network are modeled using differential equations, for this we introduce variables *T* (TDIF), *X* (PXY), *C* (TDIF bound by PXY), *A* (ANT), *P* (PLT5), *H* (HD-ZIP III) and finally, *a* (auxin). Model cells can attain different vascular fates through experiencing different 959 combinations of auxin and TDIF levels and through the regulatory network translating this into different<br>960 gene expression patterns for PXY, ANT, PLT5 and HD-ZIP III. Cells with sufficient PLT5 and ANT gene expression patterns for PXY, ANT, PLT5 and HD-ZIP III. Cells with sufficient PLT5 and ANT expression adopt cambial identity. Cells lacking these factors differentiate into xylem if they have sufficient HD-ZIP III, or phloem if they lack HD-ZIP III (for more details on the applied threshold levels to determine cell fate see later sections). In addition to intracellular expression dynamics, the TDIF signaling peptide is excreted and diffuses in the apoplast, while the ANT and PLT5 proteins diffuse between cells via plasmodesmata.

- 
- Gene regulatory and signaling network architecture
- 

 The architecture of the model gene regulatory and signaling network is shown in **(fig. S7)**. Yellow PE (published elsewhere) and orange TS (this study) symbols refer to supporting data the interaction was based on and are summarized in the accompanying tables. In the below sections we describe the system of differential equations used to model the dynamics of the various network components.

- 
- Gene expression and signaling dynamics
- 
- PXY-TDIF expression and binding

 At the heart of the model is the interaction between the auxin induced receptor protein PXY (*X*) and the phloem produced mobile peptide ligand TDIF (*T*), that binds to it forming a PXY-TDIF complex (*C*). We capture the production, movement, association, disassociation, and degradation of PXY and TDIF with the following set of equations.

$$
\frac{dX}{dt} = p_X \frac{a^2}{a^2 + K_{a,X}^2} \left( \left( 1 - f_{P,X} \right) + f_{P,X} \frac{K_{P,X}^2}{P^2 + K_{P,X}^2} \right) - d_X X - K_{on} X T + K_{off} C \tag{1}
$$

$$
\frac{dT}{dt} = p_T - K_{on}XT + K_{off}C - d_T T \tag{2}
$$

$$
\frac{dC}{dt} = K_{on}XT - K_{off}C - d_C C
$$
\n(3),

982 where  $p_X$  is the maximum PXY production rate,  $K_{a,X}$  is the value for which the auxin-mediated induction of 983 PXY is at half maximum,  $f_{P,X}$  is the maximum fraction of PXY expression that can be repressed by PLT5 984 and  $K_{P,X}$  is PLT5 level at which this repression is at half maximum,  $d_X$  is the PXY degradation rate,  $K_{on}$  and 985 *K<sub>off</sub>* are the rates of PXY-TDIF association and disassociation,  $p_T$  and  $d_T$  are maximum production and 986 degradation rates for TDIF and  $d<sub>C</sub>$  is the degradation rate of PXY-TDIF complex. In the single cell layout, 987 through varying the level of  $p<sub>T</sub>$  we investigate the impact of TDIF level on cell fate.

988 The dissociation constant  $K_d = K_{off}/K_{on}$  determines the *binding strength* of TDIF (*T*) to PXY (*X*), and <br>989 hence influences the amount of complex (*C*) formed that may induce ANT (*A*) and PLT (*P*) expression (se hence influences the amount of complex (*C*) formed that may induce ANT (*A*) and PLT (*P*) expression (see equations below) and by necessity at the same time affects the amount of TDIF that is sequestered by PXY into complex and hence no longer available to freely diffuse, i.e. the *sequestration strength*. Since we hypothesize that sequestration plays an important role in cambium patterning, we need to compare model simulations with and without strong sequestration. If we were to do this using Eq. 1-3, to obtain weak 994 sequestration would imply using a very high  $K_d$ . However, in addition to resulting in weak TDIF sequestration this would also result in low complex levels and hence absence of ANT/PLT induction. To sequestration this would also result in low complex levels and hence absence of ANT/PLT induction. To artificially decouple these two effects that both influence cambium patterning, and investigate the importance of sequestration without affecting complex levels and ANT/PLT induction, we make use of an alternative set of equations:

999 
$$
\frac{dX}{dt} = p_X \frac{a^2}{a^2 + K_{a,X}^2} \left( (1 - f_{P,X}) + f_{P,X} \frac{K_{P,X}^2}{P^2 + K_{P,X}^2} \right) - d_X X \qquad (1*)
$$

$$
\frac{dT}{dt} = p_T - d_T T \tag{2*}
$$

1001 
$$
C = \frac{\left(X + T + \frac{Koff}{Kon}\right) - \sqrt{\left(X + T + \frac{Koff}{Kon}\right)^2 - 4 * (XT)}}{2}
$$
 (3\*)

 where X and T now stand for total instead of unbound PXY and TDIF levels, and steady state complex levels are computed from total PXY and TDIF levels following the earlier defined binding and unbinding rates yet ignoring complex degradation and without taking into account that complex formation results in sequestration, i.e we assume all T can diffuse freely.

1006

#### 1007 HD-ZIP III expression

1008 Aside from inducing PXY, auxin also induces the expression of the xylem identity transcription factor HD-1009 ZIP III (*H*). We use the following equation to describe HD-ZIP III expression dynamics:

$$
\frac{dH}{dt} = p_H \frac{a^4}{a^4 + K_{a,H}^4} - d_H H \tag{4}
$$

1010 where  $p_H$  is the maximum HD-ZIP III production rate,  $K_{a,H}$  is the auxin level at which HD-ZIP III production is half maximum and  $d_H$  is the degradation rate of HD-ZIP III. Note the use of a Hill-coefficient production is half maximum and  $d_H$  is the degradation rate of HD-ZIP III. Note the use of a Hill-coefficient of 4 as compared to the Hill-coefficient of 2 in our other equations. Importantly, here as in many other cases we have no detailed experimental data on the number of auxin response elements (AREs) in promotor or enhancer and whether or not auxin response factor (ARF) binding to these elements is cooperative, nor on 1015 other potential sources for non-linearity to support specific Hill coefficients. Still, non-linear dynamics are<br>1016 easonable to assume from the rationale that gene expression has a certain maximum (saturation) and t reasonable to assume from the rationale that gene expression has a certain maximum (saturation) and that 1017 particularly in development non-linear interactions are the rule rather than the exception and are essential<br>1018 for spatial patterning. Furthermore, the Hill-coefficients applied here are on the low side, and far fr for spatial patterning. Furthermore, the Hill-coefficients applied here are on the low side, and far from inducing steep on-off switch like behavior. The somewhat higher Hill-coefficient for auxin-driving HD- ZIP III activation was reverse engineered from the requirement that HD-ZIP III needs to become more highly expressed and hence dominant over PXY at higher auxin levels, while having a more spatially constrained domain and thus a higher auxin *Km*, thereby ensuring a stable xylem domain.

1023

#### 1024 ANT and PLT5 expression

1025 ANT expression is induced by both auxin and PXY-TDIF, while being repressed by HD-ZIP III. We assume

1026 that auxin and PXY-TDIF induction of ANT function additively and are independently antagonized by HD-<br>1027 ZIP III. Combined this results in the following equations:

ZIP III. Combined this results in the following equations:

1028

$$
\frac{dA}{dt} = p_A \left( \frac{a_{a,A}(H)a^2}{a^2 + K_{a,A}^2} + \frac{a_{c,A}(H)C^2}{C^2 + K_{c,A}^2} \right) - d_A A
$$

$$
a_{a,A}(H) = m_{a,A} \left( \left( 1 - r_{H,a} \right) + \frac{r_{H,a} K_{H,A}^2}{H^2 + K_{H,A}^2} \right)
$$
  
\n
$$
a_{C,A}(H) = m_{C,A} \left( \left( 1 - r_{H,C} \right) + \frac{r_{H,C} K_{H,A}^2}{H^2 + K_{H,A}^2} \right)
$$
\n
$$
(5),
$$

1029 where  $p_A$  is the maximum ANT production rate,  $a_{a,A}$  and  $a_{C,A}$  are the auxin and PXY-TDIF dependent ANT 1030 induction functions that depend on HDZIP-III repression,  $K_{a,A}$  and  $K_{C,A}$  are the values of auxin and PXY-1031 TDIF for which their activation of ANT expression are at half maximum,  $d_A$  is the degradation rate of 1032 ANT,  $m_{a,A}$  and  $m_{C,A}$  are the maximum fractions of auxin and TDIF signaling mediated ANT induction,  $r_{H,a}$ 1033 and  $r_{H,C}$  are the maximum fractions of these that can be repressed by HDZIP-III, and  $K_{H,A}$  is the HDZIP-<br>1034 III level at which this repression is half maximal. III level at which this repression is half maximal.

1035 Note that while auxin and TDIF-PXY complex are assumed to regulate ANT expression additively, auxin 1036 and TDIF-PXY also regulate ANT expression in a multiplicative manner due to the auxin dependent 1037 expression of PXY. Finally, HDZIPIII, which antagonizes ANT expression is also auxin-dependent. Since expression of PXY. Finally, HDZIPIII, which antagonizes ANT expression is also auxin-dependent. Since 1038 there is no experimental evidence to support either additive or multiplicative regulation of ANT repression 1039 by auxin and TDIF-PXY, the additive scenario was chosen as it represents a worst-case scenario as it 1040 enables ANT activation outside of the domain of high TDIF-PXY expression, making constrained 1041 ANT/PLT expression more challenging and rendering HDIZIPIII repression more important. 1041 ANT/PLT expression more challenging and rendering HDIZIPIII repression more important.

1042 In contrast to ANT, PLT5 expression is only induced by TDIF peptide signaling and is not repressed by 1043 HD-ZIP III, resulting in the following simpler equation:

$$
\frac{dP}{dt} = p_P \frac{C^2}{C^2 + K_{C,P}^2} - d_P P
$$
\n(6),

1044 where  $p<sub>P</sub>$  is the maximum PLT5 expression rate,  $K<sub>C,P</sub>$  is the PXY-TDIF level for which the expression of 1045 PLT5 is half maximum, and  $d_P$  is the degradation rate of PLT5.

#### 1047 Model parametrization

1048

1049 After defining network architecture based on experimental data and establishing the basic differential 1050 equations-based model describing the dynamical interactions resulting from this network architecture, we 1051 next determined the (range of) parameter values to be used for the model.

- 1052
- 1053 Defining a parameter range
- 1054
- 1055 Production and degradation rates

 As a first step, in absence of absolute quantitative data, we scaled the maximum level for the signaling molecules auxin and TDIF, as well as for the transcription factors HD-ZIP III, ANT, and PLT5 and the receptor protein PXY to a value of 100, which can be interpreted as meaning 100%. Note that such a "relativistic" approach is frequently applied in computational developmental modeling studies, both inside and outside of plant biology (examples are (*66*, *67*)) The maximum level of 100 implies that the ratio of 1061 production (p) over degradation (d) rates for each gene is set at 100, while the actual rates can be varied.<br>1062 We typically use a d of 0.0002s<sup>-1</sup> (half-life of approximately 1 hour) and hence a p of 0.02[]s<sup>-1</sup> ( We typically use a *d* of  $0.0002s^{-1}$  (half-life of approximately 1 hour) and hence a  $p$  of  $0.02$ []s<sup>-1</sup> (see table 1063 **S1**), which albeit somewhat arbitrary is chosen such that it is significantly faster than the ra **S1**), which albeit somewhat arbitrary is chosen such that it is significantly faster than the rate of cell 1064 division, ensuring that despite growth and division driven dilution of proteins, concentrations are alwa division, ensuring that despite growth and division driven dilution of proteins, concentrations are always near steady state. While in the current model there is no growth and division of cells, this allows seamless future incorporation of these processes without requiring reparameterization. For the ANT and PLT5 1067 proteins  $(p_A, d_A, p_p, d_p)$  we use 10 times slower dynamics (so half-life of approximately 10 hours). We 1068 based this on our previous work that identified a slow turnover of PLT2 proteins, assuming that the closely based this on our previous work that identified a slow turnover of PLT2 proteins, assuming that the closely related PLT3,5, and 7 and ANT modeled here have similar turnover dynamics (*18*).

1070

1071 As an exception to the above, for the receptor protein PXY we applied a higher production and degradation 1072 rate (typically  $p_X=0.12$ []s<sup>-1</sup> and  $d_X=0.0012$ s<sup>-1</sup> is used, see **table S1**). The rationale behind these higher rates is that these dynamics encompass both receptor internalization and turnover. Assuming that upon rates is that these dynamics encompass both receptor internalization and turnover. Assuming that upon 1074 binding of the TDIF ligand to the PXY receptor, the PXY-TDIF complex turns over at the same rate as 1075 isolated PXY  $(d_C = 0.0012s^{-1})$ , this enabled us to investigate the effect of receptor mediated TDIF signal negative sequestration and enhanced degradation on TDIF signaling gradient formation. In the parameter sweep, sequestration and enhanced degradation on TDIF signaling gradient formation. In the parameter sweep, 1077 performed on single cell simulations, so in absence of TDIF diffusing away to other cells, the  $p_T$  parameter 1078 is varied between 0 and  $0.02$ []s<sup>-1</sup> to vary total (free+bound) TDIF levels between 0 and 100. In the 1079 multicellular simulations, where TDIF is only produced in the phloem cell and diffuses xylemward we 1080 applied a higher production rate of  $0.03$ []s<sup>-1</sup> in simulations without actual PXY mediated TDIF 1081 sequestration, and an even higher production rate of  $0.13$ []s<sup>-1</sup> in simulations with sequestration to 1082 compensate for the faster turnover of bound TDIF. For a more detailed explanation of this absence/presence 1083 of TDIF sequestration see the parameter sweep section.

#### *Km* values and ANT expression regulation

 As a second step, having set the maximum level of 100, we were now able to deduce the relevant range for 1087 model  $K_m$  values, under the assumption that the incorporated players and regulatory interactions were correct and sufficient to explain cambium patterning. Put simply, *Km* values of above 100 would cause downstream activated regulatory factors to never reach their maximum levels or downstream repressed factors to never approach their minimum levels, whereas *Km* values of 5 or lower would nearly abolish the dependence on an upstream regulatory factor, setting a first broad range of relevant values. Based on the biological data, and the simulated auxin and TDIF profiles in our model, reasonable *Km* values can 1093 subsequently be further constrained.

 Experimental data show that at the phloem side, HD-ZIP III, ANT, PXY are all not expressed. Given that these 3 factors are induced by auxin, and that at the phloem side the superimposed shape of the auxin gradient results in auxin levels of around 8 for the 3 cell settings. Therefore, the *Km* at which auxin results in half maximum activation of HD-ZIP III, ANT and PXY factors should significantly exceed this auxin level in order to avoid activation of these factors at the phloem side and was therefore set to a value of at least 15. Additionally, since these factors are known to be highly expressed at the xylem side of the vasculature where auxin levels are highest (up to 100 in our model), the maximum value for these *Km* values 1102 should not be higher than 50 (70 for  $K_{a,H}$  because of its higher Hill coefficient). Taking into account the observation that HD-ZIP III is expressed exclusively in the xylem, where auxin levels are highest, while PXY is expressed in both xylem and cambium stem cells and ANT is only expressed in the cambium stem 1105 cells, where auxin levels are intermediate, we can further constrain and order the  $K_m$  values. (Note that ANT expression in the xylem is antagonized by HD-ZIP III, see **Fig. 3E**). Specifically, these data imply that *Ka,H* 1107 is larger than  $K_{a,X}$  and  $K_{a,A}$ . Finally we constrain  $K_{H,A}$  to only become half activated at an HD-ZIP III level of 30 to restrict HD-ZIP III activity to the high auxin domain. Thus, we have established a biologically plausible range for the 4 *Km* values in **table S2**.

1111 The last four parameters in **table S2** relate to the induction of ANT by auxin and PXY and its repression<br>1112 by HD-ZIP III. For these we again used biological data and practical considerations to constrain their val by HD-ZIP III. For these we again used biological data and practical considerations to constrain their values. Since auxin also induces PXY, the additive induction of ANT by auxin and PXY corresponds to a direct 1114 activation and indirect activation by auxin. To constrain maximum ANT expression and maintain its<br>1115 maximum level at 100 like for the other genes (discussed earlier), in our parameter sweep the parameters maximum level at 100 like for the other genes (discussed earlier), in our parameter sweep the parameters 1116 controlling the maximum contribution of these two fractions ( $m_{C,A}$  and  $m_{a,A}$ ) were varied such that their 1117 sum equals 1 (i.e.  $m_{C,A} = 1 - m_{a,A}$ ) causing their relative contribution to the maximum ANT expressio 1117 sum equals 1 (i.e. $m_{C,A} = 1 - m_{a,A}$ ) causing their relative contribution to the maximum ANT expression to 1118 be anti-correlated. Since the PXY-TDIF interaction is known to be critical for ANT expression, we set the be anti-correlated. Since the PXY-TDIF interaction is known to be critical for ANT expression, we set the 1119 maximum direct auxin fraction for *ANT* induction to 0.4 and hence the minimum for the indirect PXY-<br>1120 TDIF induction of ANT to 0.6. The last two parameters,  $r_{H,a}$  and  $r_{H,P}$ , represent the extent to which HD-Z 1120 TDIF induction of ANT to 0.6. The last two parameters,  $r_{H,a}$  and  $r_{H,P}$ , represent the extent to which HD-ZIP<br>1121 III represses the direct auxin induction of ANT, compared to the indirect PXY-TDIF induction of ANT 1121 III represses the direct auxin induction of ANT, compared to the indirect PXY-TDIF induction of ANT and<br>1122 are independently varied in our parameter sweep. A minimum of 0.4 was set as HD-ZIP III as the biological are independently varied in our parameter sweep. A minimum of 0.4 was set as HD-ZIP III as the biological data dictates that HD ZIP III at least partially represses ANT induction and we are not interested in the 1124 regime where this does not occur for the parameter sweep.

#### PXY-TDIF signaling related parameters

The parameters in **table S3** refer to the translation of PXY-TDIF signal to PLT5 expression as well as the

PLT5 repression of PXY expression. Since PXY-TDIF is formed at the region of overlap between opposing

PXY and TDIF gradients, where both PXY and TDIF levels are substantially submaximal we reasoned that

- 1130 as a minimum requirement parameter settings should result in maximum TDIF and PXY levels (100)<br>1131 translating into a high level of PXY-TDIF complex (e.g. 80) and little PXY and TDIF to remain unbound translating into a high level of PXY-TDIF complex (e.g. 80) and little PXY and TDIF to remain unbound 1132 (20). This gives us the following constraint for PXY-TDIF association  $(K_{on})$  and dissociation  $(K_{off})$  rates:
- 1133

$$
\frac{K_{off}}{K_{on}} = \frac{[P] * [T]}{[C]} = \frac{20 * 20}{80} = 5\tag{7}
$$

1135 Additionally, we assumed that association and dissociation rates are faster than the turnover of these 1136 proteins themselves. We settled on  $K_{on} = 0.02$  and  $K_{off} = 0.1$ , the latter being approximately 100 times 1137 faster than the degradation rate of PXY. faster than the degradation rate of PXY.

1138

1139 Note that since PXY is expressed as a gradient tapering off from the xylem side and TDIF diffusion results 1140 in a gradient tapering off from the phloem side, at the point where PXY and TDIF meet PXY and TDIF levels are considerably lower than the 100 mentioned above. As a consequence, TDIF-PXY complex 1141 levels are considerably lower than the 100 mentioned above. As a consequence, TDIF-PXY complex 1142 numbers and receptor occupancy levels will be significantly smaller. Thus, intermediate values of PXY-1143 TDIF complex should be capable of inducing the high levels of ANT and PLT5 expression required for 1144 cambial identity, while low PXY-TDIF levels should not lead to their expression. Combined this sets a 1145 range of  $K_m$  values for PXY-TDIF induced expression for ANT between 15 and 35, and for PLT5 between 1146 20 and 40. We assigned slightly lower values for  $K_{CA}$  than for  $K_{CP}$  for two reasons. First, PLT5 represses 20 and 40. We assigned slightly lower values for  $K_{C,A}$  than for  $K_{CP}$  for two reasons. First, PLT5 represses 1147 PXY, so to somewhat protect PXY from this, PLT5 induction should require significant levels of PXY-1148 TDIF. Secondly, HD-ZIP III represses ANT, so to compensate for this, a slightly lower PXY-TDIF should 1149 already sufficiently induce ANT.

1150

 The repression of PXY by PLT5 results in a negative feedback loop that puts a cap on the PXY-TDIF induced *PLT5* and *ANT* expression. To still enable the high *PLT5* and *ANT* expression observed experimentally, we assume that PLT5 can maximally reduce PXY levels by 30%, and that for this maximum 1154 repressive activity high PLT5 levels  $(K_m>30)$  are needed. We speculate that such a parametrization may serve *in planta* as a sort of homeostatic mechanism, enabling cells to generate high expression of *PLT5* and *ANT* with moderate PXY-TDIF levels, while preventing even higher PLT5 and ANT levels when PXY- TDIF levels further increase. Since the maximum and *Km* of PLT5 mediated PXY repression have similar effects, we keep this maximum repression constant while varying the *Km* to vary overall repression in the 1159 performed parameter sweep.

1160

#### 1161 Cell fate threshold values

1162 Finally, we need to set the values for the threshold parameters determining how we translate gene 1163 expression patterns into vascular cell fate. Based on our experimental data it is the activity of PLT5 and 1164 ANT that induce cambium stem cell identity. Given that PLT5 and ANT individually have a maximum ANT that induce cambium stem cell identity. Given that PLT5 and ANT individually have a maximum 1165 protein level of 100, a threshold level above 100 implies that cambium stem cell identity requires both 1166 factors being present in significant amounts. Since knockout studies suggest this not to be the case (*20*), we 1167 chose a value of ANT+PLT5>75 for the cambium stem cell identity threshold.

1168

1169 Similarly, experimental data indicates that HD-ZIP III expression induces xylem cell fate. We set the 1170 threshold value for HD-ZIP III above which xylem fate is induced to 30. Note that the precise level of this threshold value for HD-ZIP III above which xylem fate is induced to 30. Note that the precise level of this 1171 threshold mainly determines the auxin level required to shift from phloem to xylem fate. Phloem fate occurs 1172 if neither the demands for xylem nor cambium stem cell fate have been met by the cell's expression state,<br>1173 thus PLT5+ANT <75 and HD-ZIP III<30.

- thus PLT5+ANT <75 and HD-ZIP III<30.
- 1174

## 1175 Parameter sweep

1176 To determine the robustness with which the above-described network reproduces the biological observation 1177 of high auxin levels resulting in xylem differentiation, high TDIF and low auxin levels resulting in phloem 1178 fate, and intermediate values inducing cambial identity, we applied a parameter sweep across the previously

1179 identified plausible ranges of parameter values. We varied each of the parameters from **tables S2 and table**  1180 **S3** in set increments between these extreme values. Given that  $m_{C,A} = 1 - m_{a,A}$ , and  $r_{H,a} = r_{H,C}$  this results in a total of 10 free parameters, with for 7 parameters 5 distinct values, and for 3 parameters 7 results in a total of 10 free parameters, with for 7 parameters 5 distinct values, and for 3 parameters 7 distinct values. Overall, this results in a 10-dimensional parameter search space subsampled in  $5^7$  times 1183  $7^3$ =26,796,875 different points. To allow interested readers to perform their own parameter sweep without 1183  $7<sup>3</sup>=26,796,875$  different points. To allow interested readers to perform their own parameter sweep without 1184 the use of extensive computational resources, in the provided code we increased the step size with the use of extensive computational resources, in the provided code we increased the step size with which 1185 parameter values are varied 2-fold reducing the number of sampled parameter combinations to  $3<sup>7</sup>$  times 1186  $4^{3}$  =139,968.

1187 For each selected combination of parameter values the single cell model is subjected to a matrix of auxin 1188 and (total, free+bound) TDIF levels. Auxin levels are simply superimposed and varied between 0 and 100 1189 in increments of 10 resulting in a total of 11 different auxin levels. TDIF levels are the result of TDIF<br>1190 production, sequestration by PXY, and degradation of free and bound TDIF. Due to the higher turnover of production, sequestration by PXY, and degradation of free and bound TDIF. Due to the higher turnover of 1191 bound relative to free TDIF, and the fact that bound TFY not only depends on total TDIF but also on PXY 1192 and hence auxin levels, total TDIF is hard to precisely control through simply varying its production rate<br>1193 in the full model. Therefore, to obtain the precise control of total TDIF levels independent of auxin lev in the full model. Therefore, to obtain the precise control of total TDIF levels independent of auxin level 1194 that is needed for our parameter sweep, we used TDIF and PXY levels to compute TDIF-PXY complex 1195 levels and the PLT/ANT induction this results in yet ignored PXY mediated TDIF sequestration and the 1196 differential degradation this results in (thus applying equations  $1^*$ ,  $2^*$ ,  $3^*$  instead of 1, 2, 3). Note that -1197 given that this is a single cell model in which TDIF cell-to-cell movement is irrelevant- identical simulation 1198 outcomes would occur if instead of ignoring sequestration the bound and unbound TDIF would have<br>1199 identical degradation rates, as well as that highly similar simulation outcomes would occur if the higher identical degradation rates, as well as that highly similar simulation outcomes would occur if the higher 1200 degradation rate of TDIF is compensated for by a higher production rate. Through varying  $p_T$  between 0<br>1201 and 0.02 with increments of 0.004, combined with  $d_T = 0.0002$  s<sup>-1</sup>, we vary total TDIF between 0 and 100 1201 and 0.02 with increments of 0.004, combined with  $d_T = 0.0002 \text{ s}^{-1}$ , we vary total TDIF between 0 and 100<br>1202 in steps of 20, resulting in a total of 6 different TDIF levels. Overall, each sampled parameter combin in steps of 20, resulting in a total of 6 different TDIF levels. Overall, each sampled parameter combination 1203 is thus subjected to a matrix of 66 auxin-TDIF combinations. For each pair of auxin and TDIF levels 1204 investigated, we score across the entire range of the parameter sweep the frequency with which the cell 1205 converges to the different possible cell types, enabling us to draw 2D auxin-TDIF cell fate maps. As long 1206 as the combined threshold for PLT5+ANT for cambium stem cell formation remains below 100, there is 1207 only a quantitative shift in model outcome (**fig. S8, B** and **C**), while the general behavior remains robust.

- 1208
- 1209 Role of HD-ZIP III in specifying xylem

1210 By taking specific subsets of the parameter sweep, where we keep one parameter constant, we can assess 1211 the effect of that parameter. While for most parameters we set relatively narrow ranges, the HD-ZIP III the effect of that parameter. While for most parameters we set relatively narrow ranges, the HD-ZIP III 1212 repression of ANT via  $r_{H,a}$  and  $r_{H,X}$  were left free to vary between barely repressing ANT at 0.4, to 1213 completely repressing ANT at 1.0. By comparing the results of the entire parameter sweep with the results 1214 of subset of simulations obtained for  $r_{H,a}$  =1 and  $r_{H,X}$  =1, we show how a maximum repression of ANT by 1215 HD-ZIP III is able to shift cells with high auxin and intermediate TDIF from cambial identity to xylem 1216 identity (compare **fig. S8, B** and **D**). Thus, this maximum HD-ZIP III activity can safeguard the high auxin 1217 xylem from intermediate TDIF mediated conversion to cambium stem cell fate.

#### 1219 Model analysis

1220 After having established (a range of) plausible parameter values for our model, we analyzed whether the 1221 defined model architecture results in a single equilibrium the location of which (variable values) depends<br>1222 on auxin and TDIF inputs, or rather that the model structure results in multi-stability with parameters 1222 on auxin and TDIF inputs, or rather that the model structure results in multi-stability with parameters 1223 impacting the presence and basin of attraction of the alternative equilibria. For this we analytically derived 1224 the steady states of the model (Appendix 1), demonstrating the model allows for only a single positive, real<br>1225 valued equilibrium. valued equilibrium.

1226

#### 1227 Extension to multicellular model

 In the multicellular model we created a 1D tissue strand with a xylem organizer cell on the left, a mature phloem cell on the right, and 1-3 cambial cells in between. On this 1D cell file we superimposed an auxin gradient which has its maximum at the xylem organizer cell and incorporated production of TDIF occurring in the mature phloem cells (see Eq. 2). The model was run till it reached steady state before analyzing outcomes.

1233 The model is grid based, with a space step of 0.5 microm with individual cells having an explicit width and 1234 height, and cell walls in between cells having the width of a single space step, i.e. 0.5 microm. To tak 1234 height, and cell walls in between cells having the width of a single space step, i.e. 0.5 microm. To take into<br>1235 account experimentally observed cell size differences, we applied the following cell widths: xylem ce 1235 account experimentally observed cell size differences, we applied the following cell widths: xylem cell 12<br>1236 um, cambium stem cell 8 um and phloem 16 um. The height of cells was set at 25 um independently of cell  $\mu$ m, cambium stem cell 8  $\mu$ m and phloem 16  $\mu$ m. The height of cells was set at 25  $\mu$ m independently of cell 1237 type.

1238

#### 1239 Superimposed auxin gradient

1240 Auxin is a key player in cambium development, being a major regulator for HD-ZIP III, PXY and ANT 1241 expression. In the cambium, experimental data show a characteristic auxin gradient with its maximum at 1242 the most cambium-ward adult xvlem cell that gradually decreases towards the phloem. In absence of 1242 the most cambium-ward adult xylem cell that gradually decreases towards the phloem. In absence of 1243 sufficient data on the relative importance of longitudinal and transversal auxin transport and local auxin sufficient data on the relative importance of longitudinal and transversal auxin transport and local auxin 1244 production in shaping this gradient, instead of explicitly modeling auxin dynamics, we superimposed an 1245 auxin gradient according to the following equation:

1246

$$
a_{xylem} = max_a
$$
  
\n
$$
a_1 = (max_a - drop_{xylem})
$$
  
\n
$$
a_{i>1} = \frac{(max_a - drop_{xylem})}{mod_{cambium}i^2}
$$
\n(8)

1247 ,where i is the cell number starting at 1 in the most xylem-ward cambial cell,  $max_a$  is the level of auxin at 1248 the auxin maximum in the xylem (default level 100),  $drop_{wlem}$  is the initial drop relative to the maximu the auxin maximum in the xylem (default level 100),  $drop_{xvlem}$  is the initial drop relative to the maximum 1249 for the first cambial cell (default level 60), and *mod<sub>cambium</sub>* cambial modulates the further reduction of auxin 1250 as distance (measured in number of cells) from the xylem increases (default level 1.25). Overall this results 1251 in a sharp drop of auxin levels from the xylem to the first neighboring cambium cells, followed with a in a sharp drop of auxin levels from the xylem to the first neighboring cambium cells, followed with a more 1252 gradual decline of auxin levels to subsequent cambium cells.

1253

#### 1254 TDIF and PLT/ANT transport

1255 Gene expression and thus protein level dynamics were modeled on a cellular level, using a single ODE per 1256 species and cell (see equations below). In contrast, transport of TDIF, ANT and PLT was modeled on the species and cell (see equations below). In contrast, transport of TDIF, ANT and PLT was modeled on the 1257 grid level. For this, first, cell level protein levels were assigned to the individual grid points within a cell.<br>1258 Next, we computed a concentration gradient based flux between the rightmost grid points of the lef Next, we computed a concentration gradient based flux between the rightmost grid points of the left 1259 neighbouring cell and the leftmost grid points of the right cell. This can be taken as a diffusive flux through 1260 the plasmodesmata from one cell's cytoplasm to the other in the case of PLT/ANT, or as the diffusive flux<br>1261 cross the cell wall separating two cells in the case of TDIF. To enable the use of a single diffusion cons across the cell wall separating two cells in the case of TDIF. To enable the use of a single diffusion constant 1262 we assumed plasmodesmatal density and aperture to be equal between the different cell types and incorporated the effects of plasmodesmatal mediated diffusional transport into lowered effective diffusion incorporated the effects of plasmodesmatal mediated diffusional transport into lowered effective diffusion 1264 rates. Similarly, we assumed homogeneous cell wall properties and hence a single diffusion constant for 1265 TDIF transport. No-flux boundary conditions were applied to the leftmost gridpoints of the leftmost cell 1265 TDIF transport. No-flux boundary conditions were applied to the leftmost gridpoints of the leftmost cell 1266 and the rightmost gridpoints of the rightmost cell. After computing the transport, grid-based levels within 1267 each cell were averaged to obtain cell level values. Note that by computing transport only between boundary<br>1268 grid points, we automatically scaled for cell volume and for length of cell-cell interface (see equation grid points, we automatically scaled for cell volume and for length of cell-cell interface (see equations 1269 below). Transport computations were performed at each timestep after updating the ordinary differential 1270 equations. Model simulations use the rate parameters of **table S1**, and the default auxin and PXY-TDIF parameter values of **table S2 and table S3** unless explicitly stated differently, as described in **table S5 and** 1271 parameter values of **table S2 and table S3** unless explicitly stated differently, as described in **table S5 and**  1272 **table S6.**

1273

1274 Apoplastic diffusion of the phloem secreted TDIF peptide into the cambium stem cells and towards the 1275 xylem is critical for achieving TDIF signaling. The TDIF peptide consists of 12 amino acids (aac) (*68*). 1276 While no experimental data for TDIF diffusion rates could be found in the literature, we found rates of 4.3<br>1277 \*  $10^{-10} \text{m}^2/\text{s}$  for oxytocin (9 aac) and 3.52 \*  $10^{-10} \text{m}^2/\text{s}$  for somatostatin (14 aac) (69 1277  $* 10^{-10}$ m<sup>2</sup>/s for oxytocin (9 aac) and 3.52  $* 10^{-10}$ m<sup>2</sup>/s for somatostatin (14 aac) (69), and 2.54  $* 10^{-10}$ m<sup>2</sup>/s 1278 for aprotinin (58 aac) (70) and  $1.26 * 10^{-10}$  m<sup>2</sup>/s for the approximately 50 amino acid long drkN SH3 domain 1279 (*71*). Given the relation between number of aac, molecular weight and protein size TDIF diffusion rates are 1280 expected to be closest to the higher values observed for the similar sized oxytocin and somatostatin peptides,<br>1281 i.e. approximately  $4 * 10^{-10} \text{m}^2/\text{s}$ . However, we found that in order to prevent TDIF from spre i.e. approximately 4  $*$  10<sup>-10</sup>m<sup>2</sup>/s. However, we found that in order to prevent TDIF from spreading 1282 homogeneously over the small number of cells in and around the cambium, diffusion coefficients that are 1283 5 orders of magnitude smaller were required in our model  $(D_T=0.00094 \text{ um}^2/\text{s}=0.94 * 10^{-15} \text{m}^2/\text{s})$ . 1284 Experimentally measured diffusion constants obtained from the literature are typically obtained in a watery 1285 solution. In contrast, in the cell wall viscosity is much higher than that of water and it is well established 1286 that viscosity is inversely related to the diffusion coefficient (Stokes-Einstein relation), additionally the cell wall is charged leading to interactions further slowing diffusion (72), and finally the cell wall is a 1287 wall is charged leading to interactions further slowing diffusion (72), and finally the cell wall is a porous<br>1288 medium with the cell wall fibrils forming a complex network of obstacles for diffusion (effective diff medium with the cell wall fibrils forming a complex network of obstacles for diffusion (effective diffusion 1289 inversely related to tortuosity). Indeed, a classical study by Kramer and co-workers demonstrated that the 1290 environment of the cell wall causes a one to two order of magnitude decrease in diffusion rates relative to 1291 water, with the two order of magnitude decreases occurring higher up in the root in the area where cambium **formation occurs (73).** Correcting the experimental values for this (i.e.  $4 * 10^{-12} \text{m}^2/\text{s}$ ), thus reduces the 1293 difference between experimental and model diffusion constants to 3 orders of magnitude.

1294

1295 Notably, our specific implementation of transport effectively assumes infinitely fast diffusion of TDIF<br>1296 across the cell wall overlaying the cells and limited TDIF diffusion across the cell walls in between cells 1296 across the cell wall overlaying the cells and limited TDIF diffusion across the cell walls in between cells in<br>1297 2D (or for PLT/ANT infinitely fast diffusion inside cells and limited diffusion through plasmodesmata 1297 2D (or for PLT/ANT infinitely fast diffusion inside cells and limited diffusion through plasmodesmata), 1298 while in reality, identical diffusion can be expected for the cell walls overlaying the cells as for the cell 1299 walls separating the cells in our 2D model. Consequently, in addition to diffusion being inhomogeneous, 1300 *effective* diffusion across the entire 1D tissue is larger than suggested by the value of the diffusion *effective* diffusion across the entire 1D tissue is larger than suggested by the value of the diffusion 1301 coefficients used to implement the in between cell transport. If we assume a periodic tissue, with an average 1302 cell width of 11 microm and a cell wall width of 0.5 microm, one can approximate this effective diffusion<br>1303 coefficient as the harmonic mean of the intracellular and intercellular diffusion coefficients: 1303 coefficient as the harmonic mean of the intracellular and intercellular diffusion coefficients:

1304 
$$
D_{eff} = \frac{11.5}{\frac{11}{D_{intra}} + \frac{0.5}{D_{inter}}}
$$

1305 which for an infinite intracellular diffusion coefficient translates to  $D_{eff} = 23 D_{inter} = 2.2 10^{-14}$ , that 1306 would further reduce the difference between (adjusted) experimental and (effective) model diffusion 1306 would further reduce the difference between (adjusted) experimental and (effective) model diffusion constants to 2 orders of magnitude. To validate this inference, we created an alternative model in which constants to 2 orders of magnitude. To validate this inference, we created an alternative model in which 1308 full, homogeneous 2D diffusion of TDIF across the tissue strand was simulated. In addition to generating 1309 intracellular gradients of TDIF and TDIF-PXY complex we found that while for the normal diffusion coefficient value a very limited TDIF diffusion leads to failure of PLT and ANT activation indeed a 23coefficient value a very limited TDIF diffusion leads to failure of PLT and ANT activation indeed a 23-1311 fold increased diffusion coefficient compared to the default model generates a similar TDIF gradient, PXY 1312 binding and PLT and ANT activation (Fig S10A).

1313

 Finally, as explained earlier we take a relativistic modeling approach and set maximum levels of all players 1315 to 100 arbitrary units by fixing the ratio of production over degradation rates to 100. While we applied some<br>1316 reasoning for the timescales of production and degradation rates, with e.g. ANT and PLT having slower 1316 reasoning for the timescales of production and degradation rates, with e.g. ANT and PLT having slower<br>1317 turnover based on experimental data than other proteins, applied timescales are still relatively arbitrary. turnover based on experimental data than other proteins, applied timescales are still relatively arbitrary. With regards to TDIF these timescales are relevant for the diffusion coefficients needed to achieve a particular gradient length. Assuming homogeneous TDIF diffusion and ignoring PXY mediated TDIF sequestration, the analytical solution for a production-degradation-diffusion system is given by:

1321 
$$
T(x) = \frac{p_T \cosh(\lambda x)}{d_T \cosh(\lambda L)}
$$

1322 with  $\lambda = \sqrt{d_T/D}$  and *L* the length of the tissue. From this it follows that identical steady state gradients can<br>1323 be generated if the TDIF diffusion coefficient and degradation constant are both increased or de be generated if the TDIF diffusion coefficient and degradation constant are both increased or decreased by 1324 the same factor. This analysis thus suggests that unless mechanisms are in place that cause TDIF to diffuse<br>1325 substantially slower than can be expected for a peptide of that size in planta TDIF turnover is approxim substantially slower than can be expected for a peptide of that size in planta TDIF turnover is approximately 1326 100-fold higher than assumed in our default model parametrization.

1327 We added a scale factor in our model code, allowing interested users to test that simultaneously increasing<br>1328 TDIF diffusion, TDIF and PXY production and degradation dynamics with the same factor although 1328 TDIF diffusion, TDIF and PXY production and degradation dynamics with the same factor although<br>1329 generating somewhat different temporal dynamics results in identical steady state model behavior. 1329 generating somewhat different temporal dynamics results in identical steady state model behavior.

1330

1331 The transcription factors ANT and PLT5 are capable of moving between cells through plasmodesmata (18).<br>1332 For their transport, our model uses as a default a diffusion coefficient that is 3 orders of magnitude smalle 1332 For their transport, our model uses as a default a diffusion coefficient that is 3 orders of magnitude smaller 1333 than that used for TDIF. As for TDIF, our specific implementation of ANT and PLT5 transport effectively 1334 assumes infinite fast diffusion inside cells and limited diffusion between cells, and following the harmonic 1335 mean approach we can again estimate that effective diffusion constant is 23-fold higher than the model 1336 value. However, since this applies for both TDIF and PLT/ANT this does not affect the size difference 1337 between the TDIF and PLT/ANT diffusion constants. Therefore, we performed additional simulations 1338 varying either TDIF or PLT/ANT diffusion rates, using strong sequestration settings (see below) (Fig. S10 varying either TDIF or PLT/ANT diffusion rates, using strong sequestration settings (see below) (Fig. S10 1339 B). We find that our results are reasonably robust to 2.5-fold increases in TDIF diffusion while showing 1340 significant deviations for 5-fold increases in TDIF diffusion results. In contrast, model outcomes are robust<br>1341 against a 10-fold increase in PLT/ANT diffusion rate which still results in spatially constrained ANT/ against a 10-fold increase in PLT/ANT diffusion rate which still results in spatially constrained ANT/PLT 1342 patterns with a clearly localized maximum, while for a 100-fold increase ANT/PLT patterns instead become 1343 smeared out, preventing localized cambium stem cell patterning. These results indicate that PLT/ANT<br>1344 diffusion rate could be increased 10-fold relative to default model parametrization, and that hence only a 2 1344 diffusion rate could be increased 10-fold relative to default model parametrization, and that hence only a 2<br>1345 order of magnitude smaller diffusion for PLT/ANT than for TDIF is required. However, if TDIF diffusion order of magnitude smaller diffusion for PLT/ANT than for TDIF is required. However, if TDIF diffusion 1346 rate is further increased to better agree with experimental data concomitant with increases in TDIF turnover 1347 (see above), this difference would further increase.

 To justify differences in TDIF and PLT/ANT diffusivity let us consider both differences in TDIF peptide and PLT/ANT protein properties and in where or how their transport is taking place. ANT and PLT5 proteins (555-558 amino acids) are far larger than the small TDIF peptides (12 aac). Assuming 110 Dalton per amino acid this results in a weight of approximately 61kDa for ANT/PLT5 and of 1.4kDa for TDIF. Next, assuming globular protein shape and assuming that 10kDa corresponds to a globular radius of 1.59nm this implies a radius of 2.9nm for ANT/PLT5 and 0.82nm for TDIF (for details see Appendix 2). Following the Stokes-Einstein relation the 3.5 larger radius of ANT/PLT5 compared to TDIF should already result in a 3.5 slower diffusion. Combined with the likely non-perfectly globular shape of particularly larger proteins like ANT/PLT5 this results in an order of magnitude difference in diffusion rates. On top of this TDIF is a signaling molecule that is excreted and diffuses in the apoplast, whereas ANT/PLT5 are transcription factors that need to leave the nucleus, diffuse in the cytoplasm and then enter the plasmodesmatal neck region to diffuse from cell to cell through the narrow plasmodesmatal sleeve. Currently we do not have data to determine whether ANT/PLT5 sizes exceed the size exclusion limit of the plasmodesmata in the cambium, nor whether ANT/PLT5 plasmodesmata transport involves protein unfolding or plasmodesmatal 1363 regulation. However, given the approximately 2.9nm molecular radius diffusional hindrance is expected to<br>1364 occur and can easily result in a further 10-100 fold reduction in diffusion rates. Combined, the size occur and can easily result in a further 10-100 fold reduction in diffusion rates. Combined, the size difference and the difference in where/how transport is taking place, this is expected to result in substantial 1366 differences in diffusion rates justifying the modeled differences between TDIF and ANT/PLT diffusion 1367 rates (see table S4). rates (see **table S4**).

- 
- Maximum HD-ZIP III repression versus strong sequestration settings

 In **(fig. S9; and Fig. 4)** of the main manuscript text we compare two alternative model settings. Specifically, we used either the maximum HD-ZIP III' repression of ANT **(Fig. 4, A and E; fig. S9C)** or we used the "strong sequestering" parameter values (**Fig. 4, B and F; fig. S9D**) to investigate the relevance of regulatory interactions versus immobilization of TDIF through PXY mediated sequestration for robust cambium patterning.

 For the maximum HD-ZIP III simulations, to enable us to focus on the importance of regulatory interactions we omitted effective TDIF sequestration, using multicellular versions of equations 1\*, 2\* and 3\* rather than 1, 2 and 3 (for explanation see above). Note that in theory a similar effect of strongly reduced TDIF 1378 sequestration while maintaining effective ANT/PLT induction can be achieved through reducing the 1379 binding of TDIF to PXY (increased  $K_d$ ), while simultaneously introducing a compensatory increase in the binding of TDIF to PXY (increased  $K_d$ ), while simultaneously introducing a compensatory increase in the 1380 effectiveness of PXY-TDIF mediated ANT and PLT5 expression (lowering  $K_{C,A}$  and  $K_{C,P}$ ).

 In contrast, under the strong sequestration settings, to focus on the importance of TDIF immobilization through sequestration we reduced the relevance of regulatory interactions for patterning. To achieve this, under these settings HD-ZIP III does not repress the PXY-TDIF based expression of ANT, and exclusively 1385 represses the auxin fraction, effectively halving the maximum achievable repression. To further enhance 1386 sequestration, we lowered the  $K_{off}$  from 0.1 to 0.02. sequestration, we lowered the  $K_{\text{off}}$  from 0.1 to 0.02.

- 
- 
- Variable gradients

1390 To investigate the capacity of the network to robustly integrate and respond to various auxin and TDIF 1391 signaling inputs, we varied the auxin and TDIF production levels in a 3 cells wide cambium (5 cells in total) signaling inputs, we varied the auxin and TDIF production levels in a 3 cells wide cambium (5 cells in total) **(fig. S12)**. In this larger cambium there is more room for differences in spatial overlap of gradients for different auxin gradient and TDIF production and diffusion parameter settings (**table S6**). These

1394 simulations used the strong sequestration parameters from **table S5**. We ascribe cell fates to the cells based on the cell fate threshold values section as described above.

1396

#### 1397 Numerical procedures

1398 Differential equations were solved using simple Euler forward integration with a time step of 0.25s, and 1399 for the spatial 1D model a space step of 0.5 microm. For the spatial model no-flux boundary conditions 1400 were used. As initial conditions all protein levels for ANT, PLT5, PXY, TDIF, PXY-TDIF complex and 1401 HD-ZIP III were set to zero. Simulations were run for 60h of simulated time. Results shown represent 1402 steady state dynamics.

steady state dynamics.

## <sup>1403</sup> Appendix 1

1404 Differential equations

1405

1406 Let us first recap the relevant differential equations:

1407 
$$
\frac{dX}{dt} = \frac{p_X \cdot a^2}{a^2 + K_{a,X}^2} \cdot \left( (1 - f_{P,X}) + \frac{f_{P,X}}{1 + P^2 / K_{P,X}^2} \right) - d_X \cdot X - K_{on} \cdot X \cdot T + K_{off} \cdot C
$$

1408 where  $p_X, d_X, K_{a,X}, K_{p,X}, K_{on}, K_{off} \in \mathbb{R}^+, f_{p,X} \in \mathbb{R}^+$  and  $a(t)$  is provided as an input.

$$
\frac{dT}{dt} = p_T - d_T \cdot T - K_{\text{on}} \cdot X \cdot T + K_{\text{off}} \cdot C
$$

$$
1410 \t\t\t\t\t p_T, d_T \in \mathbb{R}^+
$$

$$
\frac{dC}{dt} = K_{\text{on}} \cdot X \cdot T - K_{\text{off}} \cdot C - d_C \cdot C
$$

$$
d_C \in \mathbb{R}^+
$$

1413 
$$
\frac{dP}{dt} = \frac{p_P \cdot C^2}{C^2 + K_{CP}^2} - d_P \cdot P
$$

$$
1414 \t\t\t p_p, d_p, K_{CP} \in \mathbb{R}^+
$$

1415 We will use  $X_{\infty}$  to indicate the steady state value of *X*,  $T_{\infty}$  to indicate the steady state value of *T*, etc. We 1416 aim to solve for  $\{X_{\infty}, T_{\infty}, C_{\infty}, P_{\infty}\}$  excluding  $\{H_{\infty}, A_{\infty}\}$ ; the first s 1416 aim to solve for  $\{X_{\infty}, T_{\infty}, C_{\infty}, P_{\infty}\}$  excluding  $\{H_{\infty}, A_{\infty}\}$ ; the first set is co-dependent and needs to be solved simultaneously. We will show below that  $T_{\infty}, C_{\infty}$  and  $P_{\infty}$  can be written as a 1417 simultaneously. We will show below that  $T_{\infty}$ ,  $C_{\infty}$  and  $P_{\infty}$  can be written as a function of  $X_{\infty}$ . This implies 1418 that to determine the number of different equilibria we can focus on solving  $X_{\infty}$ 1418 that to determine the number of different equilibria we can focus on solving  $X_{\infty}$ . Note that  $\{H_{\infty}, A_{\infty}\}$  are <br>1419 uncoupled since  $H_{\infty} = g(a_{\infty})$  (it is determined solely by the input  $a(t)$ ) while  $A_{\infty}$ 1419 uncoupled since  $H_{\infty} = g(a_{\infty})$  (it is determined solely by the input  $a(t)$ ) while  $A_{\infty} = h(H_{\infty}, a_{\infty}, C_{\infty}) =$ <br>1420  $h(X_{\infty})$  (there is one value of A for each value of X) and is not impacting  $X_{\infty}$ . Thus, A do 1420  $h(X_{\infty})$  (there is one value of A for each value of X) and is not impacting  $X_{\infty}$ . Thus, A does not contribute 1421 to the number of fixed points for the system of ODEs. to the number of fixed points for the system of ODEs.

1422 Steady-states

1423 We can directly solve for  $C_{\infty}$ :

$$
C_{\infty} = \frac{K_{\text{on}} \cdot X_{\infty} \cdot T_{\infty}}{K_{\text{off}} + d_{C}}
$$

1425 To solve for  $T_{\infty}$ , we set  $\frac{dT}{dt} + \frac{dC}{dt} = 0$ 

$$
1426 \t\t\t 0 = p_T - d_T \cdot T_{\infty} - d_C \cdot C = p_T - d_T \cdot T_{\infty} - d_C \cdot \frac{K_{\text{on}} \cdot X_{\infty} \cdot T_{\infty}}{K_{\text{off}} + d_C}
$$

1427 and thus:

1428 
$$
T_{\infty} = \frac{p_T}{d_T + d_C \cdot \frac{K_{\text{on}} \cdot X_{\infty}}{K_{\text{off}} + d_C}} = \frac{p_T \cdot K_{\text{off}} + p_T \cdot d_C}{d_T \cdot K_{\text{off}} + d_T \cdot d_C + d_C \cdot K_{\text{on}} \cdot X_{\infty}} = \frac{\alpha}{X_{\infty} + \beta}
$$

where  $\alpha := \frac{p_T \cdot K_{\text{off}} + p_T \cdot a_C}{d_C \cdot K_{\text{off}}}$  $\frac{K_{\text{off}} + p_T \cdot d_C}{d_C \cdot K_{\text{on}}}$  and  $\beta := \frac{a_T \cdot K_{\text{off}} + a_T \cdot d_C}{d_C \cdot K_{\text{on}}}$ 1429 where  $\alpha := \frac{p_T \cdot \alpha_{\text{off}} + p_T \cdot \alpha_C}{d_C \cdot K_{\text{on}}}$  and  $\beta := \frac{\alpha_T \cdot \alpha_{\text{off}} + \alpha_T \cdot \alpha_C}{d_C \cdot K_{\text{on}}}$  demonstrating that  $T_{\infty} = k(X_{\infty})$ , substituting  $T_{\infty}$  in  $C_{\infty}$ 1430 then shows that  $C_{\infty} = l(X_{\infty})$ .

1431 *P*<sub>∞</sub> is:

1432 
$$
P_{\infty} = \frac{p_P \cdot C_{\infty}^2}{d_P \cdot (C_{\infty}^2 + K_{CP}^2)} = \frac{p_P}{d_P} \cdot \frac{K_{\text{on}}^2 \cdot X_{\infty}^2 \cdot T_{\infty}^2}{K_{\text{on}}^2 \cdot X_{\infty}^2 \cdot T_{\infty}^2 + (K_{\text{off}} + d_C)^2 \cdot K_{CP}^2} = \frac{\gamma \cdot X_{\infty}^2 \cdot T_{\infty}^2}{X_{\infty}^2 \cdot T_{\infty}^2 + \delta}
$$

1433 where 
$$
\gamma
$$
 : =  $\frac{p_P}{d_P}$  and  $\delta$  : =  $\frac{(K_{\text{off}} + d_C)^2 \cdot K_{\text{CP}}^2}{K_{\text{on}}^2}$ .

1434 Substituting  $T_{\infty}$ :

1435 
$$
P_{\infty} = \frac{\gamma \cdot X_{\infty}^{2} \cdot \left(\frac{\alpha}{X_{\infty} + \beta}\right)^{2}}{X_{\infty}^{2} \cdot \left(\frac{\alpha}{X_{\infty} + \beta}\right)^{2} + \delta} = \frac{\gamma \cdot \alpha^{2} \cdot X_{\infty}^{2}}{(\alpha^{2} + \delta) \cdot X_{\infty}^{2} + 2 \cdot \beta \cdot \delta \cdot X_{\infty} + \beta^{2} \cdot \delta} = \frac{X_{\infty}^{2}}{\epsilon \cdot X_{\infty}^{2} + \zeta \cdot X_{\infty} + \eta}
$$

1436 where  $\epsilon := \frac{\alpha^2 + \delta}{\gamma \cdot \alpha^2}$ ,  $\zeta := \frac{2 \cdot \beta \cdot \delta}{\gamma \cdot \alpha^2}$  and  $\eta := \frac{\beta^2 \cdot \delta}{\gamma \cdot \alpha^2}$  demonstrating that also  $P_{\infty} = m(X_{\infty})$ .

1437 To solve for 
$$
X_{\infty}
$$
, we set  $\frac{dX}{dt} + \frac{dC}{dt} = 0$ :

1438 
$$
0 = \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{aX}^2} \cdot \left( (1 - f_{P,X}) + \frac{f_{P,X}}{1 + P_{\infty}^2 / K_{PX}^2} \right) - d_X \cdot X_{\infty} - d_C \cdot C_{\infty}
$$

1439 Since we can rewrite  $C_{\infty}$  solely as a function of  $X_{\infty}$  as:

$$
C_{\infty} = \frac{K_{\text{on}} \cdot X_{\infty} \cdot T_{\infty}}{K_{\text{off}} + d_{C}} = \frac{\alpha \cdot K_{\text{on}}}{K_{\text{off}} + d_{C}} \cdot \frac{X_{\infty}}{X_{\infty} + \beta}
$$

1441 we can expand equation  $[eq10]$  to:

$$
1442 \qquad 0 = \frac{1}{d_X} \cdot \frac{p_X \cdot a_\infty^2}{a_\infty^2 + K_{aX}^2} \cdot \left(1 - f_{P,X}\right) + \frac{1}{d_X} \cdot \frac{p_X \cdot a_\infty^2}{a_\infty^2 + K_{aX}^2} \cdot \frac{f_{P,X}}{1 + P_\infty^2 / K_{PX}^2} - X_\infty - \frac{d_C}{d_X} \cdot \frac{\alpha \cdot K_{on}}{K_{off} + d_C} \cdot \frac{X_\infty}{X_\infty + \beta}
$$

We define the following constants:  $p_0 := \frac{1}{d_1}$  $\frac{1}{d_X} \cdot \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{a}^2}$  $\frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{aX}^2} \cdot (1 - f_{P,X}), p := \frac{f \cdot p_X \cdot a_{\infty}^2}{d_X \cdot (a_{\infty}^2 + K_{cX}^2)}$ 1443 We define the following constants:  $p_0 := \frac{1}{d_X} \cdot \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{aX}^2} \cdot (1 - f_{P,X})$ ,  $p := \frac{f \cdot p_X \cdot a_{\infty}^2}{d_X \cdot (a_{\infty}^2 + K_{aX}^2)}$  and  $d := \frac{d_C \cdot a \cdot K_{\text{on}}}{d_X \cdot (K_{\text{off}} + d_C)}$ 

and thus equation [eq11] can be condensed to:

1445 
$$
0 = p_0 + \frac{p}{1 + P_{\infty}^2 / K_{PX}^2} - X_{\infty} - d \cdot \frac{X_{\infty}}{X_{\infty} + \beta}
$$

1446 Focusing on the term  $\frac{p}{1+P_{\infty}^2/K_{PX}^2}$ , we can expand it as:

1447 
$$
\frac{p}{1 + P_{\infty}^2 / K_{PX}^2} = \frac{p}{1 + \frac{X_{\infty}^4 / K_{PX}^2}{(\epsilon \cdot X_{\infty}^2 + \zeta \cdot X_{\infty} + \eta)^2}} = \frac{p \cdot (\epsilon \cdot X_{\infty}^2 + \zeta \cdot X_{\infty} + \eta)^2}{(\epsilon \cdot X_{\infty}^2 + \zeta \cdot X_{\infty} + \eta)^2 + X_{\infty}^4 / K_{PX}^2}
$$

1448 that expands to:

1449 
$$
\frac{p}{1 + P_{\infty}^2 / K_{PX}^2} = \frac{p\epsilon^2 X_{\infty}^4 + 2p\epsilon \zeta X_{\infty}^3 + p(2\epsilon \eta + \zeta^2) X_{\infty}^2 + 2p\zeta \eta X_{\infty} + p\eta^2}{(\epsilon^2 + 1/K_{PX}^2)X_{\infty}^4 + 2\epsilon \zeta X_{\infty}^3 + (2\epsilon \eta + \zeta^2) X_{\infty}^2 + 2\zeta \eta X_{\infty} + \eta^2}
$$

This is:

1451 
$$
\frac{p}{1 + P_{\infty}^2 / K_{PX}^2} = \frac{a_4 X_{\infty}^4 + a_3 X_{\infty}^3 + a_2 X_{\infty}^2 + a_1 X_{\infty} + a_0}{b_4 X_{\infty}^4 + b_3 X_{\infty}^3 + b_2 X_{\infty}^2 + b_1 X_{\infty} + b_0}
$$

1452 where 
$$
a_0 := p\eta^2
$$
,  $a_1 := 2p\zeta\eta$ ,  $a_2 := p(2\epsilon\eta + \zeta^2)$ ,  $a_3 := 2p\epsilon\zeta$ ,  $a_4 := p\epsilon^2$ ,

1453 where 
$$
b_0 := \eta^2
$$
,  $b_1 := 2\zeta\eta$ ,  $b_2 := 2\epsilon\eta + \zeta^2$ ,  $b_3 := 2\epsilon\zeta$ ,  $b_4 := \epsilon^2 + 1/K_{PX}^2$ 

1454 
$$
0 = p_0 + \frac{a_4 X_{\infty}^4 + a_3 X_{\infty}^3 + a_2 X_{\infty}^2 + a_1 X_{\infty} + a_0}{b_4 X_{\infty}^4 + b_3 X_{\infty}^3 + b_2 X_{\infty}^2 + b_1 X_{\infty} + b_0} - X_{\infty} - \frac{d \cdot X_{\infty}}{X_{\infty} + \beta}
$$

1455 
$$
0 = \frac{c_6 X_{\infty}^6 + c_5 X_{\infty}^5 + c_4 X_{\infty}^4 + c_3 X_{\infty}^3 + c_2 X_{\infty}^2 + c_1 X_{\infty} + c_0}{(b_4 X_{\infty}^4 + b_3 X_{\infty}^3 + b_2 X_{\infty}^2 + b_1 X_{\infty} + b_0)(X_{\infty} + \beta)}
$$

where

$$
c_0 := p_0 \beta b_0 + a_0 \beta
$$
  
\n
$$
c_1 := p_0(b_0 + \beta b_1) - \beta b_0 - db_0 + a_0 + a_1 \beta
$$
  
\n
$$
c_2 := p_0(b_1 + \beta b_2) - b_0 - \beta b_1 - db_1 + a_1 + a_2 \beta
$$
  
\n
$$
c_3 := p_0(b_2 + \beta b_3) - b_1 - \beta b_2 - db_2 + a_2 + a_3 \beta
$$
  
\n
$$
c_4 := p_0(b_3 + \beta b_4) - b_2 - \beta b_3 - db_3 + a_3 + a_4 \beta
$$
  
\n
$$
c_5 := p_0 b_4 - b_3 - \beta b_4 - db_4 + a_4
$$
  
\n
$$
c_6 := -b_4
$$

 This expression is equal to zero only when the numerator is zero. Since the numerator is a hexic polynomial, 1459 there must be 6 fixed points in  $\mathbb C$  for  $X_{\infty}$ .

#### 1460 Interpretation

1461 Importantly, we are modeling a biological system. Thus, only positive, real valued equilibria are relevant. 1462 The question thus is how many of these equilibria may exist. For this we make use of the fact that the 1463 coefficients of the hexic polynomial are not unconstrained. Firstly, since:

1464 
$$
p_X, d_X, K_{a,X}, K_{p,X}, K_{\text{on}}, K_{\text{off}}, a, d_C, p_T, d_T, p_P, d_P, K_{C,P} \in \mathbb{R}^+
$$

$$
f_{P,X} \in [0,1]
$$

1465 From which follows that:

1466

$$
\alpha, \beta, \gamma, \delta, \epsilon, \zeta, \eta, p_0, p, d \in \mathbb{R}^+
$$

$$
a_0, a_1, a_2, a_3, a_4 \in \mathbb{R}^+
$$

$$
b_0, b_1, b_2, b_3, b_4 \in \mathbb{R}^+
$$

$$
c_0, c_1, c_2, c_3, c_4, c_5, c_6 \in \mathbb{R}
$$

1467 The complex conjugate root theorem thus applies: given a polynomial with real coefficients, if z is a root 1468 such that  $Im(z) \neq 0$ , then its complex conjugate  $z^*$  will also be a root of this polynomial (complex r 1468 such that Im( $z$ ) ≠ 0, then its complex conjugate  $z^*$  will also be a root of this polynomial (complex roots 1469 come in pairs). Thus, given that complex roots necessarily occur in pairs, without taking the constra come in pairs). Thus, given that complex roots necessarily occur in pairs, without taking the constraints of 1470 the problem into account, there could be zero, two, four or six real roots.

To investigate whether one or more equilibria exist in ℝ<sup>+</sup> we applied a heuristic approach. We sampled 1472 random parameter values (drawn uniformly from  $U(0,10 \times v_i)$ , where  $v_i$  is the parameter value used in the 1472 random parameter values (drawn uniformly from  $U(0,10 \times v_i)$ , where  $v_i$  is the parameter value used in the 1473 final model settings (tables S1-3). An exception is  $f_{PX}$  which was sampled from  $U(0,1)$ . The number 1473 final model settings (tables S1-3). An exception is  $f_{P,X}$  which was sampled from  $U(0,1)$ . The number of 1474 parameter value combinations sampled was  $n_{\text{compling}} = 10^6$ , for each of which we evaluated the number of 1474 parameter value combinations sampled was  $n_{\text{sampling}} = 10^6$ , for each of which we evaluated the number of 1475 real valued solutions r. real valued solutions r.

1476 The number of real valued solutions for all the sampled parameter combinations are shown in the table 1477 below:

 $R^-: 1; R^+: 1 \quad R^-: 3, R^+: 1$ 999063 937

- 1478 These numbers support the high likelihood of a single equilibrium existing in  $R^+$ .
- 1479
- 1480 Phase plane analysis
- 1481

1482 As an alternative to solving for the equilibrium of our system and to further proof there is only a single<br>1483 equilibrium in  $R^+$ , we performed a graphical phase plane analysis to determine null clines and their p 1483 equilibrium in  $R^+$ , we performed a graphical phase plane analysis to determine null clines and their points 1484 of intersection and thereby the number of equilibria in  $R^+$ . For this we needed to reduce our model 1484 of intersection and thereby the number of equilibria in  $R^+$ . For this we needed to reduce our model to 2<br>1485 dimensions. As before we ignored H and A, focussing on X, T, C and P, which together constitute a 4 dimensions. As before we ignored *H* and *A*, focussing on *X*, *T*, *C* and *P*, which together constitute a 4 1486 dimensional system. We reduced this to a 2 dimensional system by taking a so-called quasi steady state 1487 assumptions for both the C variable and the T variable, eliminating these as variables and replacing them 1488 by algebraic expressions. This procedure makes this problem tractable without affecting the steady-state 1489 properties of the system.

1490

1491 We previously showed that:

1492 
$$
T_{\infty} = \frac{\alpha}{X_{\infty} + \beta}; \quad C_{\infty} = \frac{K_{\text{on}} \cdot X_{\infty} \cdot T_{\infty}}{K_{\text{off}} + d_{C}} = \frac{\alpha \cdot K_{\text{on}}}{K_{\text{off}} + d_{C}} \cdot \frac{X_{\infty}}{X_{\infty} + \beta}
$$

1493 Assuming quasi steady state, we can similarly replace T and C with the following algebraic expressions for 1494  $T(X)$  and  $C(X)$ :

1495 
$$
T(X) = \frac{\alpha}{X + \beta}; \quad C(X) = \frac{\alpha \cdot K_{\text{on}}}{K_{\text{off}} + d_C} \cdot \frac{X}{X + \beta}
$$

1496 Trivially, these expressions tend to  $T_{\infty}$  and  $C_{\infty}$  as  $t \to \infty$ .

1497 Next we need to find the null clines for X and P by solving for  $\ddot{x} = \frac{dx}{dt} = 0$  and  $\dot{P} = \frac{dP}{dt} = 0$  while making

1498 use of the algebraic expressions for C and T.

1499 For P this results in the following equation of the P null cline

$$
P_{\dot{P}=0}(X) = \frac{X^2}{\epsilon \cdot X^2 + \zeta \cdot X + \eta}; \quad X \in \mathbb{R}^+
$$

1501 The P null cline thus corresponds to a monotonically increasing non-linear saturating function in the 1502 positive quadrant since

1503 
$$
\frac{d}{dX}P_{p=0}(X) = \frac{\zeta \cdot X^2 + 2 \cdot \eta \cdot X}{(\epsilon \cdot X^2 + \zeta \cdot X + \eta)^2} > 0 \text{ if } X \in \mathbb{R}^+
$$

1504 Writing an analytical expression for the X null cline is more complex:  $\lambda$ 

1505 
$$
\frac{dX}{dt} = \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{ax}^2} \cdot \left( (1 - f_{P,X}) + \frac{f_{P,X}}{1 + \frac{P^2}{K_{PX}^2}} \right) - d_X \cdot X - d_C \cdot \frac{\alpha \cdot K_{\text{on}}}{K_{\text{off}} + d_C} \cdot \frac{X}{X + \beta} = 0
$$

1506 Which can be rewritten as

1507 
$$
0 = p_0 + \frac{pK_{PX}^2}{K_{PX}^2 + P^2} - X - d \cdot \frac{X}{X + \beta}
$$

where as before  $p_0 := \frac{1}{d_1}$  $\frac{1}{d_X} \cdot \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_a^2}$  $\frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{aX}^2} \cdot (1 - f_{P,X}), p := \frac{f \cdot p_X \cdot a_{\infty}^2}{d_X \cdot (a_{\infty}^2 + K_{cX}^2)}$ 1508 where as before  $p_0 := \frac{1}{d_X} \cdot \frac{p_X \cdot a_{\infty}^2}{a_{\infty}^2 + K_{aX}^2} \cdot (1 - f_{P,X}), p := \frac{f \cdot p_X \cdot a_{\infty}^2}{d_X(a_{\infty}^2 + K_{aX}^2)}$  and  $d := \frac{d_C \cdot a \cdot K_{on}}{d_X \cdot (K_{off} + d_C)}$ 

1509 this can be reordered as

1510 
$$
\frac{pK_{PX}^2}{K_{PX}^2 + P^2} = p_0 + X + d \cdot \frac{X}{X + \beta}
$$

1511 rewritten as

1512 
$$
\frac{pK_{PX}^2}{K_{PX}^2 + P^2} = \frac{p_0(X + \beta) + X(X + \beta) + dX}{X + \beta}
$$

1513 and next as

1514 
$$
\frac{K_{PX}^2 + P^2}{pK_{PX}^2} = \frac{X + \beta}{p_0(X + \beta) + X(X + \beta) + dX}
$$

1515 inverted as

1516 
$$
P^2 = \frac{pK_{PX}^2(X+\beta)}{p_0(X+\beta)+X(X+\beta)+dX}-K_{P}^2
$$

1517 And finally written as

1518 
$$
P_{\dot{X}=0}(X) = \pm K_{PX} \sqrt{\frac{p \cdot (X + \beta)}{p_0 \cdot (X + \beta) + X \cdot (X + \beta) + d \cdot X}} - 1
$$

1519 where only the positive expression is biologically relevant. This can be expanded as:

1520 
$$
P_{\dot{X}=0}(X) = K_{PX} \sqrt{\frac{-X^2 + (p - p_0 - \beta - d) \cdot X + p \cdot \beta - p_0 \cdot \beta}{X^2 + (p_0 + \beta + d) \cdot X + p_0 \cdot \beta}} = K_{PX} \sqrt{\frac{P(X)}{Q(X)}}
$$

defined in the domain of  $S := \{X \in \mathbb{R}^+ | P(X) > 0\}$  since  $Q(X) > 0$   $(p_0, \beta, d \in \mathbb{R}^+)$ . To assess it potential 1522 monotonicity, we differentiate with respect to X and study its sign: monotonicity, we differentiate with respect to  $X$  and study its sign:

1523 
$$
\frac{d}{dX} P_{\dot{X}=0}(X) = K_{PX} \frac{P'(X) \cdot Q(X) - Q'(X) \cdot P(X)}{2 \sqrt{\frac{P(X)}{Q(X)}} Q(X)^2}
$$

1524 Since for  $X \in S$ , both  $P(X)$ ,  $Q(X) > 0$ , the denominator is strictly positive in this domain. We thus focus 1525 on the sign of the numerator: on the sign of the numerator:

1526 
$$
P'(X) \cdot Q(X) - Q'(X) \cdot P(X) = -p \cdot (X^2 + 2\beta X + \beta^2 + \beta \cdot d) < 0
$$

1527

1528 Since the expression for the X null cline describes a monotonically decreasing function while the expression for the P null cline describes a monotonically increasing function only a single point of intersection in for the P null cline describes a monotonically increasing function only a single point of intersection in  $R^+$  1530 is possible, proving there is only a single biologically relevant equilibrium in our system. is possible, proving there is only a single biologically relevant equilibrium in our system.

- 1531
- 1532
- 1533
- 1534
- 1535
- 1536
- 1537
- 
- 1538
- 1539
- 1540

# Appendix 2

- The precise size of a specific protein will depend on the number and type of amino acids it contains and
- how these determine the shape the protein will fold into. To get an estimate of protein size (radius), typically the following approach is used:
- 1548 1. It is assumed that the protein has a globular shape, and hence a volume that can be described as 1549  $V = \frac{4}{3} \pi r^3$ , with *V* volume in nm<sup>3</sup> and *r* radius in nm.
- 1550 2. It is assumed that the protein has an average protein density  $\rho$  in Da/nm<sup>3</sup> enabling the conversion 1551 between molecular weight and volume as  $V = \frac{1}{\rho} MW$ , where  $MW$  is the molecular weight of the protein in Dalton.
- An explanation of this approach can also be found in the following documentation:
- https://biologicalproceduresonline.biomedcentral.com/counter/pdf/10.1007/s12575-009-9008-x.pdf
- Online different calculators for protein radius from molecular weight are available which use slightly
- 1557 different assumptions for the value of average protein density  $\rho$  in their calculations resulting in slightly
- different estimates for protein radius *r.*
- We made use of the following two online calculators:
- [https://nanocomposix.com/pages/molecular](https://nanocomposix.com/pages/molecular-weight-to-size-calculator)-weight-to-size-calculator
- https://www.fidabio.com/molecular[-weight-to-size-protein-radius-calculator](https://www.fidabio.com/molecular-weight-to-size-protein-radius-calculator)
- The first website assumes a molecular density that results in a predicted 1.42nm radius for a globular
- protein of 10kDa, whereas the second website assumes a somewhat different molecular density that
- results in a predicted 1.78nm radius for a 10kDa globular protein. We decided to use an average value of
- a 1.59nm radius for 10kDa protein to calculate the radii for the TDIF peptide and ANT/PLT proteins.





Upregulated genes in p35S:CLE41 root









#### **Fig. S1. Transcriptomic analysis of** *pxy* **and CLE41 overexpression**

 **(A)** Seedling material for the RNA-seq analysis. Left, an image of a seedling showing the position of cross section shown on the right. Cross-sections of 7-day old Wild type Col-0, *pxy, p35S:CLE41* of hypocotyls and roots. **(B)** Gene Ontology terms reduced in *pxy* and enriched in *p35S:CLE41* root respectively. Red arrows mark the primary xylem axis. (**C**) Venn diagram showing numbers of genes with differential expression in *pxy* and *p35S:CLE41* shoots or roots relative to Col-0 in 7-day-old seedlings. Genes upregulated in *p35S:CLE41* and down-regulated in *pxy*, were considered more likely to be PXY-signaling targets. Sections within the Venn diagram where *PLT3*, *PLT5*, *PLT7*, and *ANT* fall are marked. Scale bars 20 µm.

- 
- 



**Fig. S2. Four** *AIL***/***PLT***s are expressed in root cambium**

 **(A)** Confocal cross-sections of 14-day-old *pPLT3:erVenus, pPLT5:erRFP, gPLT7-YFP* and *pANT:erRFP*  roots. **(B)** Confocal cross-sections of 14-day-old roots and longitudinal view of root tips of *gPLT1-YFP, gPLT2-YFP* and *gPLT4-YFP* show expression in root tip as previously reported (*18*), however they show no fluorescence in root vascular cambium. White arrowheads mark recent cell division. Vessels (v), cambium (c), sieve element (se). Scale bars 10 µm.



**Fig. S3. Histological analysis of higher order** *ail/plt* **mutants generated by crossing.**

 **(A)** Cross section of 14-day-old Col-0 and *ant-GK;plt5* double mutant together with the quantification of the vascular diameter. **(B)** Gross morphology of 11-day-old Col-0, *plt3plt5plt7-cr* and *plt3plt5plt7-cr;ant- GK* seedlings. (**C**) Root cross-sections of 12-day-old Col-0, *plt3plt5plt7-cr*, and *plt3plt5plt7-cr;ant-GK.* 1593 Quantification of the vascular diameter presents on the right panel. Significance difference was tested by<br>1594 Student t-test in (A). Letters indicate significant differences using one-way ANOVA with a Tukey post hoc 1594 Student t-test in (A). Letters indicate significant differences using one-way ANOVA with a Tukey post hoc<br>1595 test in (C). Yellow arrows mark the primary xylem axis. Vessels (v), sieve elements (se), Scale bars 10 µm test in (C). Yellow arrows mark the primary xylem axis. Vessels (v), sieve elements (se), Scale bars 10  $\mu$ m (A and C) and 1 cm (B). 



**Fig. S4. Histological analysis of higher order** *ail/plt* **mutants generated by gene editing**

 (**A**) Gross morphology of 11-day-old plants of Col-0, *plt3plt5plt7-cr*, *plt3plt5plt7-cr;IGE-ant*, and *plt3plt5plt7-cr;ant-cr*. (**B**) Root cross-sections of 13-day-old Col-0, *plt3plt5plt7-cr*, *plt3plt5plt7-cr;IGE-ant*, and *plt3plt5plt7-cr;ant-cr* seedlings. The quantification of vascular tissue diameter is shown on the 1604 right. (C) Confocal cross-section of 13-day-old roots of Col-0, *plt3plt5plt7-tdna*, *plt3plt5plt7-tdn*  right. (**C**) Confocal cross-section of 13-day-old roots of Col-0, *plt3plt5plt7-tdna, plt3plt5plt7-tdna;IGE-ant*  and the quantification of the vascular tissue diameter and the number of sieve elements (right panels). 8/15 of *plt3plt5plt7-tdna;IGE-ant* roots showed sectors without vessel production. 6/8 of these sectors occurred in the position of phloem pole, and these 6 sectors were used in quantification of sieve elements (lower right panel). Inset images show reduced number of sieve elements (marked with green dot) and the sector marked with dotted line (magenta). Cell walls are stained with SR2200 (grey), lignified cell walls are stained with 0.1% basic fuchsin (red). Letters indicate significant differences using one-way ANOVA with a Tukey post hoc test in (B), or using Kruskal-Wallis with Dunn post hoc test in (C). Numbers in (C)

- 1612 represent the frequency of the observed phenotypes. Yellow arrows mark the primary xylem axis. Vessels 1613 (v), sieve elements (se), Scale bars 10  $\mu$ m (B), 50  $\mu$ m (C) and 1 cm (A).
- (v), sieve elements (se), Scale bars 10  $\mu$ m (B), 50  $\mu$ m (C) and 1 cm (A).



## 1615

## 1616 **Fig. S5. Expression patterns of key cambial regulators**

1617 **(A)** Confocal cross-sections of 12-day-old roots expressing *ANT* double markers. **(B)** Confocal cross-<br>1618 sections of 13-day-old roots expressing *pAtHB8:AtHB8-YFP* double markers. **(C)** Confocal cross-sections 1618 sections of 13-day-old roots expressing *pAtHB8:AtHB8-YFP* double markers. **(C)** Confocal cross-sections of 14-day-old *pPXY:erRFP;pCLE41:erVenus* root. **(D)** Confocal cross-sections of *pANT:erRFP* after 1-1619 of 14-day-old *pPXY:erRFP;pCLE41:erVenus* root. **(D)** Confocal cross-sections of *pANT:erRFP* after 1- 1620 day TDIF treatment in 14-day-old plants. Numbers in (D) represent the frequency of the observed 1621 phenotypes. White arrowheads mark recent cell division. Phloem (p). Vessels (v). Scale bars 10 µm (A to 1622 D).



#### 1625 **Fig. S6. Analysis of the consequences of inducible** *PLT5* **overexpression**

1626 **(A)** Volcano plot of all the transcript after 8 hours or 24 hours induction of *35S:XVE>>PLT5-TagRFP* 1627 RNA-seq. The dotted horizontal line corresponds to a Benjamini–Hochberg corrected significance of Padj<br>1628 Value < 0.05. The dotted vertical lines bound the minimal fold-change for the most-differentially-expressed 1628 Value <0.05. The dotted vertical lines bound the minimal fold-change for the most-differentially-expressed genes. (B), Bar plot shows the percentage of the genes that are upregulated, downregulated and 1629 genes. **(B),** Bar plot shows the percentage of the genes that are upregulated, downregulated and 1630 differentially expressed genes (DEGs) in categories of All genes, xylem, phloem and cell cycle genes for 8<br>1631 and 24 hours using PValue < 0.05. (C), Confocal cross-section of  $35S: XVE \gg PLT5-TagRFP$  after 2-day 1631 and 24 hours using PValue <0.05. **(C),** Confocal cross-section of *35S:XVE>>PLT5-TagRFP* after 2-day 1632 induction (Ind) (in 12-day-old plants) showing ectopic cell division in xylem parenchyma. These ectopic<br>1633 divisions are not present in the Mock. Most recent cell divisions in cambium are marked with dotted lines divisions are not present in the Mock. Most recent cell divisions in cambium are marked with dotted lines

1634 (white). **(D),** Cross-section of 35S: XVE>>PLT5-TagRFP;gVND6-GUS after 2-day induction (in 8-day-old plants). **(E),** Cross-section of 35S: XVE>>PLT5-TagRFP after 7-day induction (in 8-day-old plants). **(F),** 1635 plants). **(E),** Cross-section of *35S:XVE>>PLT5-TagRFP* after 7-day induction (in 8-day-old plants). **(F),** 1636 RT-qPCR showing the reduced *PXY* transcript levels in 35S:XVE>>PXY-RNAi;pANT:erRFP. Barplot shows average with  $\pm$  sd. Numbers in panels (C to E) represents the frequency of the observed phenotypes. 1637 shows average with  $\pm$  sd. Numbers in panels ( $\overline{C}$  to E) represents the frequency of the observed phenotypes.<br>1638 Yellow arrowheads mark recent, ectopic cell division, red arrows mark the primary xylem axis, an 1638 Yellow arrowheads mark recent, ectopic cell division, red arrows mark the primary xylem axis, and blue arrows marks the expanding xylem vessels. Vessels (v). Sieve elements (se). Estradiol induction (Ind). 1639 arrows marks the expanding xylem vessels. Vessels (v). Sieve elements (se). Estradiol induction (Ind). 1640 Scale bars 20  $\mu$ m (C to E). Scale bars 20  $\mu$ m (C to E). 1641







#### 1643 **Fig. S7. Signaling network in vascular cambium used in the modelling**

1644 Top panel indicates the same network of modelled regulatory interactions driving cambial cell fate decision<br>1645 making as shown in Fig. 3C. Bottom panels indicate the experimental support for the different incorporat 1645 making as shown in Fig. 3C. Bottom panels indicate the experimental support for the different incorporated regulatory interactions categorized according to being published elsewhere (yellow, PE) or observed in this 1646 regulatory interactions categorized according to being published elsewhere (yellow, PE) or observed in this current study (orange, TS). current study (orange, TS).





#### **Fig. S8. Fate map for large scale Parameter Sweeps**

 **(A)** Model simulation examples illustrating convergence to alternative cell fates depending on varying auxin and TDIF input levels. a=auxin, H=HD-ZIP III, X=PXY, T=TDIF, C=TDIF-PXY complex, P=PLT5, A=ANT. Parameter settings used are the final values shown in Tables S1-S5, combined with maximum 1654 HD-ZIP III mediated ANT repression, and normal TDIF-PXY complex formation dynamics  $(K_d=5)$ . Left figure: auxin level of 10 and (total) TDIF level of 80 leads to phloem fate (HD-ZIP III<30 and PLT5+ANT<75); middle figure: auxin level of 40 and (total) TDIF level of 60 leads to stem cell fate (PLT5+ANT>75); right figure: auxin level of 100 and (total) TDIF level of 0 leads to xylem fate (HD-ZIP III>30). Left figure: Since PXY (X) and HD-ZIPIII (H) are both auxin-dependent, X is plotted on top of H, and thus H is not visible. Similarly, in the right figure in absence of TDIF (T), both T, TDIF-PXY complex 1660 (C), total TDIF (T+C) and PLT (P) equal zero, P is plotted on top making T, C, and T+C not visible, while 1661 free PXY (X) and total PXY (X+C) are equal causing X to not be visible and ANT (A) and ANT plus PLT (P+A) being equal causing A to not be visible. (**B**) Fate map of overall parameter sweep including both 1663 strong and weak HD-ZIP III mediated ANT repression for a cambial fate threshold of 75. For the range and sampling interval of parameter values used for the parameter sweep see Supplementary Modeling Methods. sampling interval of parameter values used for the parameter sweep see Supplementary Modeling Methods. Each Auxin-TDIF combination is colored according to the fraction of simulations that acquire xylem, cambium, or phloem identity according to the color triangle. **(C)** Alternative fate map of an overall parameter sweep using a cambial fate threshold of 87.5. Raising this threshold to 87.5 reduces the cambial

- domain but retains the same qualitative behavior as shown in (B). **(D)** Fate map for a parameter sweep with
- default (same as in (B)) cell fate thresholds but now constrained to maximum HD-ZIP III mediated ANT
- repression (see Modelling Methods), showing an expansion of xylem over cambial fate for high auxin and
- 1671 high TDIF levels, when compared to (B).





**Fig. S9. Robustness to variable TDIF gradients and cambial size.**

1674 (A) Robust cell fate decision making in a 3-cell tissue for variable production and diffusion rates of TDIF<br>1675 for maximum HD-ZIP III mediated ANT repression and moderate TDIF sequestration. (**B-D**) Robustness 1675 for maximum HD-ZIP III mediated ANT repression and moderate TDIF sequestration. (**B-D**) Robustness 1676 of cell fate decision making to variation in cambial cell number for the same settings as in (A) except that 1677 TDIF production was fixed to a value of 0.04 (B), for maximum HD-ZIP III mediated ANT repression in 1677 TDIF production was fixed to a value of 0.04 (B), for maximum HD-ZIP III mediated ANT repression in 1678 absence of TDIF sequestration (C), and for low (halved) HD-ZIP III mediated ANT repression combined 1679 with enhanced TDIF sequestration (5-fold reduced TDIF-PXY dissociation) (D). C and D are both for 1679 with enhanced TDIF sequestration (5-fold reduced TDIF-PXY dissociation) (D). C and D are both for 1680 default TDIF production and diffusion rates. For A, B and D Equations 1, 2 and 3 were used, while for C default TDIF production and diffusion rates. For A, B and D Equations 1, 2 and 3 were used, while for C 1681 Equations 1<sup>\*</sup>, 2<sup>\*</sup>, 3<sup>\*</sup> were used (for details see Supplemental Methods) (**E-F**) Uncropped images presented in Fig. 4C and Fig. 4G. The yellow dashed square shows the region that is magnified in Fig. 4C and Fig. in Fig. 4C and Fig. 4G. The yellow dashed square shows the region that is magnified in Fig. 4C and Fig. 1683 4G. Vessels (v). Scale bars 10 µm. 1684





 **Fig. S10. Robustness of cambium patterning to variations in diffusion implementation and rates of TDIF and PLT/ANT.** (**A**) Cambium patterning for alternative TDIF diffusion implementation in which

1690 TDIF diffuses homogeneously across the tissue instead of only in between cells. Left: cambium patterning<br>1691 if the same diffusion coefficient is used as for the default TDIF diffusion implementation. Right: cambium if the same diffusion coefficient is used as for the default TDIF diffusion implementation. Right: cambium patterning for a 23 fold increased diffusion coefficient. (**B**) From left to right: cambium patterning for the default strong sequestration modeling setting and default diffusion implementation; cambium patterning for a 5 fold increase in TDIF diffusion rate; cambium patterning for a 2.5 fold increase in TDIF diffusion 1695 rate; cambium patterning for a 10 fold increase in PLT/ANT diffusion rate; and cambium patterning for a 1696 100 fold increase in PLT/ANT diffusion rate. 100 fold increase in PLT/ANT diffusion rate.





 

**Fig. S11. Complementation analysis of different versions of PXY.**

 **(A)** Complementation of the *pxy* phenotype with different versions of PXY-YFP. Confocal cross-sections of 14-day-old Col-0, *pxy*, *pPXY:gPXY-YFP;pxy, pPXY: gPXYK747E-YFP;pxy* and *pPXY:gPXY<sup>∆</sup>KD -YFP;pxy* 1703 roots While the transgenic line containing *pPXY* complements the *pxy* mutant, lines containing *PXY<sup>K747E</sup>* or *PXY<sup>* $\Delta K$ *D*</sup> failed to do so, and thus, show *pxy*-like stem cell differentiation phenotype: a vessel (v) adjacent to a sieve element (se). The numbers in the top right corner of subpanels represent the frequency of the observed phenotype. **(B)** Expression of PXY-YFP variants after 1d of induction. Confocal root cross sections of 6-day-old mock, *pxy*, and 5-day-old *pPEAR1:XVE>>gPXY-YFP*, *pPEAR1:XVE>>gPXYK747E - YFP* and *pPEAR1:XVE>>gPXY<sup>∆XD</sup>-YFP* seedlings induced for 1 day. Note, *pxy* phenotype is typically not yet visible in 7-day-old roots (such as *pxy* in panel B). Cell wall stained with SR2200 (grey), lignified cell walls are stained with 0.1% basic fuchsin (magenta)**.** White arrowheads mark recent cell divisions. Red arrows mark the primary xylem axis. Sieve element (se), xylem vessels (v). Scale bars 10 µm (A and B).





## 1714 **Fig. S12. Flexibility to respond to variation in auxin and TDIF gradients in variable size cambium**

1715 Overview of cell fate decision making in a 3 to 5-cell vasculature exposed to variable auxin gradient lengths 1716 (left to right) and variable TDIF production levels in the phloem (top to bottom) under strong sequest (left to right) and variable TDIF production levels in the phloem (top to bottom) under strong sequestration 1717 parameter settings. Gray bars show which cells will retain cambial identity (grey zone) as a result of 1718 expressing sufficient ANT+PLT5. Light blue cells express sufficient HD-ZIP III to differentiate to xylem,<br>1719 while light green cells lack both high levels of ANT+PLT5 and HD-ZIP III and will thus differentiate to while light green cells lack both high levels of ANT+PLT5 and HD-ZIP III and will thus differentiate to 1720 phloem. A strong auxin gradient (top right) pushes the cambial cell identity towards the phloem, allowing 1721 differentiation of new xylem cells. A strong TDIF signal pushes the cambial cell identity towards the xylem 1722 (bottom left) allowing phloem cells to differentiate. A combination of these two gradients allows for a larger<br>1723 total overlap (bottom right) that generates a larger cambium. When the overlap is too weak (top left) total overlap (bottom right) that generates a larger cambium. When the overlap is too weak (top left) a 1724 meristem cannot be sustained. The dashed line indicates the highest current ANT+PLT5 level, where a 1725 smaller cambium (with increased PXY-TDIF overlap) could be maintained.

Parameter name	Parameter value	Unit
$p_H$	0.02	$[]s^{-1}$
$d_H$	0.0002	$s-1$
$p_A$	0.002	$[$ $]$ s <sup>-1</sup>
$d_A$	0.00002	$s-1$
$p_P$	0.002	$[$ $]$ s <sup>-1</sup>
$d_{P}$	0.00002	$s-1$
$K_{\text{on}}$	0.02	$[1^{-1}S^{-1}]$
$K_{\text{off}}$	0.1	$s-1$
$p_X$	0.002	$\prod$ s <sup>-1</sup>
$d_X$	0.00012	$S^{-1}$
$d_C$	0.00012	$S-1$
$d_{\mathcal{T}}$	0.0002	$S^{-1}$

1727 **table S1. Constant valued model parameters, their values and units.**

#### 1730 **table S2. Parameters for auxin and HDZIPIII, the range of values investigated, sampling**  1731 **interval used and values used for final model settings and units.**



1732

#### 1734 **table S3. PXY-TDIF dependent parameters, the range of values investigated, sampling**  1735 **interval used and values used for final model settings and units.**



## 1738 **table S4. Diffusion parameters**



1739

## 1741 **table S5. Alternative parameter regimes**



## 1744 **table S6. Auxin and TDIF gradient values**



1745

1746

# 1747 **Data S1**

Source and RNA-seq data

# 1749

1750 **Data S2**

List of Primers, constructs, seeds



## **Citation on deposit:**

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