

# Recapitulation of the forward nuclear auxin response pathway in yeast

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**Auxin influences nearly every aspect of plant biology through a simple signaling pathway; however, it remains unclear how much of the diversity in auxin effects is explained by variation in the core signaling components and which properties of these components may contribute to diversification in response dynamics. Here, we recapitulated the entire *Arabidopsis thaliana* forward nuclear auxin signal transduction pathway in *Saccharomyces cerevisiae* to test whether signaling module composition enables tuning of the dynamic response. Sensitivity analysis guided by a small mathematical model revealed the centrality of auxin/indole-3-acetic acid (Aux/IAA) transcriptional corepressors in controlling response dynamics and highlighted the strong influence of natural variation in Aux/IAA degradation rates on circuit performance. When the basic auxin response circuit was expanded to include multiple Aux/IAs, we found that dominance relationships between coexpressed Aux/IAs were sufficient to generate distinct response modules similar to those seen during plant development. Our work provides a new method for dissecting auxin signaling and demonstrates the key role of Aux/IAs in tuning auxin response dynamics.**

synthetic biology | signaling dynamics

**E**volution depends on the plasticity of existing signaling pathways. The small molecule auxin is linked to signaling modules that allowed plants to move to land, develop new organs, and respond to the environment (1, 2). Despite the wide range of auxin responses, the core auxin signal transduction pathway is quite simple, involving a small number of components from perception through transcription (Fig. 1A). Auxin triggers the rapid turnover of Aux/IAA (IAA) proteins, which repress the activity of auxin response factor (ARF) transcription factors through recruitment of TOPLESS (TPL) corepressors (3, 4). Auxin receptors, auxin signaling F-box (AFB) proteins, act as part of an E3 ubiquitin ligase to catalyze the ubiquitination and subsequent degradation of IAs when auxin is present (5, 6). ARFs bound to auxin-responsive *cis*-regulatory elements (AuxREs) are then free to regulate the expression of auxin target genes, which include the IAs themselves (7–9).

How such a simple pathway can orchestrate the large number of context-specific responses regulated by auxin is a long-standing question. Notably, each component in the auxin signaling pathway belongs to a large gene family (2, 10, 11). In *Arabidopsis*, there are 6 AFBs, 23 ARFs, and 29 IAs. Functional divergence between component family members could provide variation in response to a generic auxin signal (12). Members of the ARF family can be classified as transcriptional activators or repressors (7, 8), and the role of repressor ARFs in regulating auxin signaling is still a matter of debate (13). In addition, expression studies of auxin-signaling gene families support the idea of an “auxin pre-pattern” where certain combinations of coexpressed components generate auxin circuits with distinct auxin response capabilities (14, 15). However, owing to the high level of redundancy and coexpression of auxin signaling component family members, feedback, and interference from other signaling pathways (16), the contributions of individual components to auxin circuit behavior have remained elusive.

Here, we ported several variations of the auxin transcriptional response pathway from *Arabidopsis thaliana* to *Saccharomyces cerevisiae*. In so doing, we defined a minimal auxin response circuit (ARC) sufficient to recapitulate auxin-induced transcription in a heterologous context. Because no feedback was engineered into these ARCs, they captured the dynamic potential of the simplest forward response architecture. The ARC consists of five plant components: an AFB, an IAA, TPL, an activating ARF transcription factor, and an auxin-responsive plant promoter (Fig. 1). The implementation of ARCs in yeast (ARC<sup>Sc</sup>) relied on successful interfacing of the ARC with other essential elements of the yeast degradation and transcriptional machinery (Fig. 1B). This modular, plug-in nature of the ARC, similar to what was observed with auxin-induced degradation (17), highlights the likelihood that the ARC could be implemented in a variety of other eukaryotic contexts. The suite of ARC variants represented in the ARC<sup>Sc</sup> collection presented an opportunity to quantitatively investigate the dynamic capabilities of the auxin response in isolation.

Auxin response pivots on relief of transcriptional repression, yet neither expression of an IAA alone nor coexpression with TPL repressed induction of the IAA19 promoter by activator ARFs (Fig. 1C and *SI Appendix*, Fig. S1). A fusion of full-length TPL with the ARF-interaction domain of IAA12 is able to fully recapitulate the stabilized IAA12 mutant phenotype (3). Following this logic, we identified two TPL truncations (TPL-N100 and TPL-N300) that conferred repression in yeast when fused to a full-length IAA. However, only repression by TPL-N100 (hereafter referred to as TPL) was relieved by auxin treatment (Fig. 1C). In a yeast strain where we could simultaneously monitor degradation of the TPL:IAA fusion and reporter activation, we observed reporter activation dynamics similar to those observed in

## Significance

**Auxin is a simple small molecule, yet it elicits an enormous variety of responses. How a single molecule can encode such diverse and context-specific information is a long-standing question. One hypothesis is that variation in auxin response components generates diversity in local auxin responses. To test this, we transplanted the forward nuclear auxin signal transduction pathway from plants into yeast. We found that auxin circuit composition can tune the dynamic response, and we identified which properties of auxin response components had the largest impact on circuit performance. Reconstitution of eukaryotic pathways in novel contexts can inform our understanding of the pathways themselves and can generate tools for engineering novel cellular behaviors.**

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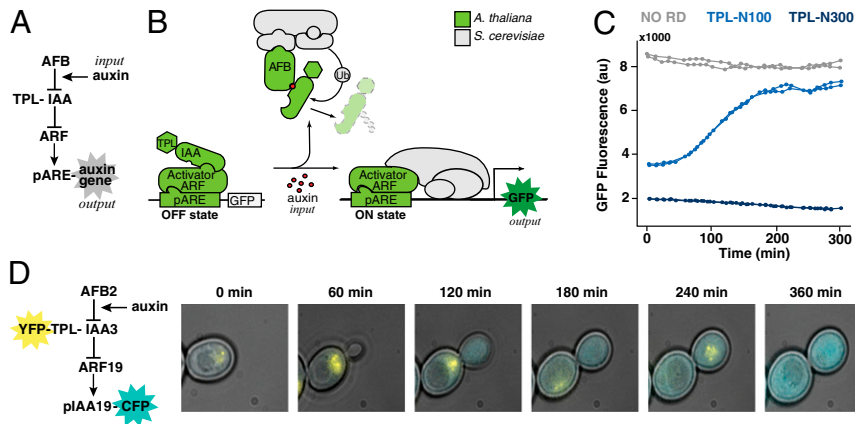
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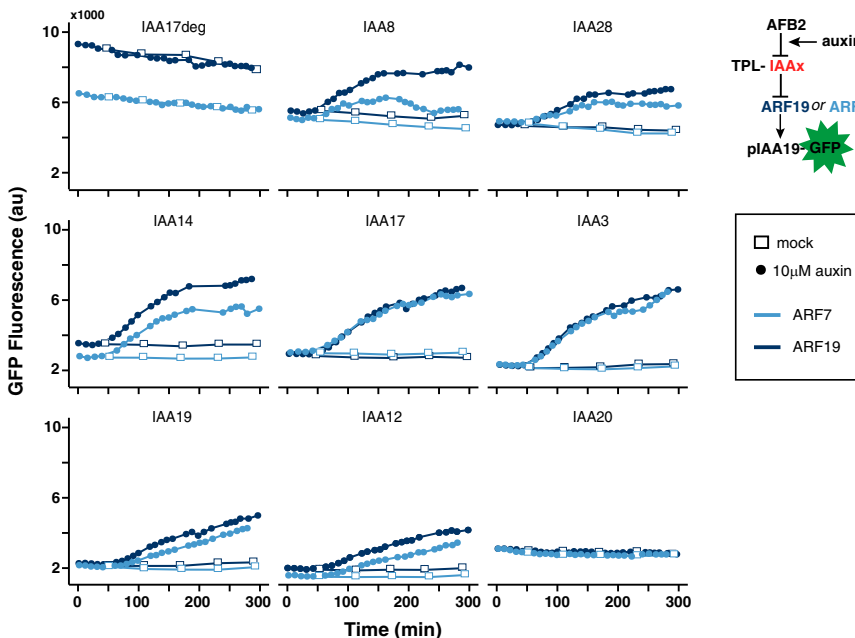


**Fig. 1.** Auxin-induced transcription in yeast. (A) Network diagram of the forward auxin response pathway in yeast. An auxin input increases association of a member of the TIR1/AFB family of F-box proteins (AFB) and an IAA protein fused to the first 100 aa of the TPL corepressor (TPL-IAA). Auxin-induced association of an AFB and a TPL-IAA leads to degradation of the TPL-IAA, thereby freeing a transcriptional activator in the ARF family to induce expression of an output gene driven by a promoter containing an auxin response element (pARE). (B) The five *A. thaliana* components needed to recapitulate auxin response in *S. cerevisiae* are shown in light green. They were an AFB F-box receptor, an IAA, a TPL corepressor, an ARF transcription factor, and an auxin-responsive promoter. The remaining cellular machinery (gray) was supplied by yeast. Fluorescence from a GFP reporter was used as a quantitative output. (C) Synthetic auxin-reversible repression required fusion of a specific TPL truncation to the IAA protein. Flow cytometry was used to monitor the induction of a GFP reporter following auxin treatment in circuits containing either IAA14 with no repression domain (NO RD), shown in gray, or two different C-terminal TPL truncations fused to IAA14. Auxin was added at time 0. TPL-N300 (dark blue) includes the first 300 aa of TPL and excludes the multiple C-terminal WD repeats. Fusion of TPL-N300 to IAA proteins results in reporter repression that is largely auxin-insensitive. TPL-N100 (light blue) includes only the first 100 aa of TPL. When fused to an IAA, TPL-N100 provides auxin-reversible repression. Two replicate induction curves are shown for each circuit. (D) Auxin-induced IAA degradation and subsequent transcriptional activation could be simultaneously monitored in dual-labeled yeast strains. YFP-TPL-IAA3 and a Cerulean reporter driven by the auxin-responsive IAA19 promoter (pIAA19-CFP) were monitored following auxin treatment using time-lapse microscopy and a microfluidic chamber.

cells where the TPL:IAA fusion was not labeled (Fig. 1D). Thus, performance of ARC<sup>Sc</sup> is consistent with the established model of plant auxin signaling, with the exception that the TPL and IAA components are acting as a single protein.

The expansion of the IAA gene family has been linked to the increase in auxin signal complexity of land plants (2), suggesting

IAAs play an essential role in tuning the auxin response. To explore the range of behaviors possible in a basic ARC configuration, we therefore took advantage of the large number of naturally evolved variants in the IAA gene family. We tested the impact of different IAAs on response dynamics of ARC<sup>Sc</sup> variants containing either ARF19 or ARF7 using time-lapse flow cytometry (Fig. 2 and



**Fig. 2.** IAAs drive auxin response dynamics. Representative auxin-induced reporter fluorescence curves are shown for ARCs with ARF7 (light blue) or ARF19 (dark blue) in the context of different IAAs. Auxin was added at time 0 in all graphs. ARF7 and ARF19 showed qualitatively similar patterns of auxin response, whereas the identity of the IAA had a dramatic effect on ARC dynamics. ARC<sup>Sc</sup> recapitulated regulatory features of plant ARC function. Transcriptional repression required the known ARF-IAA interaction domain, because an IAA lacking this domain (IAA17deg) had no effect on ARF activity. In addition, auxin response was mediated by IAA degradation, because a naturally occurring IAA lacking a degron (IAA20) rendered the circuit insensitive to auxin treatment.

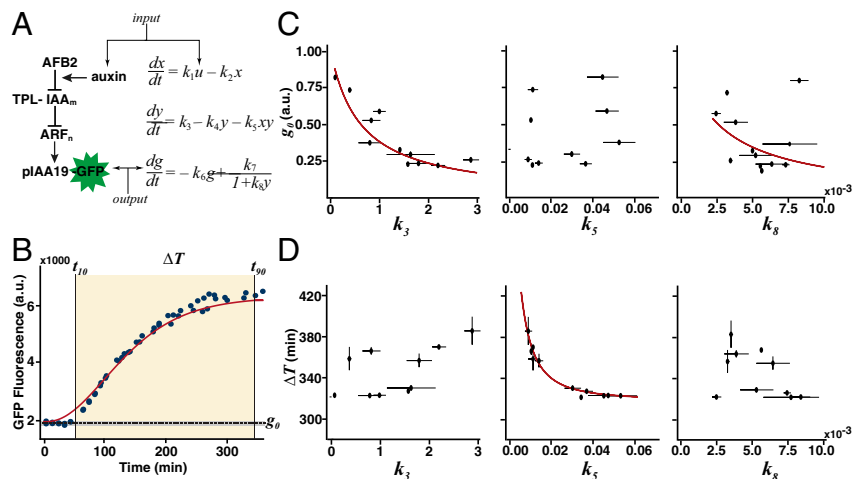
*SI Appendix, Fig. S2*). Our analysis focused primarily on IAAs that display strong mutant phenotypes when stabilized by point mutations that disrupt formation of the AFB–auxin–IAA complex (18–23). Additional IAAs found in distinct subgroups in the IAA family tree were also included (10). Circuits containing an IAA missing the ARF interaction domain (IAA17deg) failed to repress ARF activity, and thus indicated the maximum potential for reporter activation (Fig. 2). Yeast expressing IAA20, an IAA with a substantially diverged degron sequence rendering it insensitive to auxin, showed no reporter activation. Circuits with other IAAs exhibited substantial differences in the initial repressed state, as well as in the magnitude and rate of GFP reporter induction. The induction curves for a given IAA with either ARF19 or ARF7 were qualitatively quite similar, suggesting that ARF7 and ARF19 have similar or overlapping functions. This interpretation is supported by genetic and phylogenetic analysis in plants (24–26). This does not exclude the possibility that other ARF–IAA combinations, or higher-order interactions between multiple components from the ARF and IAA families, could generate additional diversity.

To quantify the differences in the behavior of ARC<sup>Sc</sup> IAA variants, we developed a small model that expands upon the model used to quantify our degradation system (27). The present model captures the overall flow of information while avoiding reference to specific molecules and kinetic parameters that cannot be measured and limiting the potential for overfitting (28, 29). In our model the variable  $u$  represents the concentration of the applied auxin input and the variable  $g$  represents the GFP output of the reporter (Fig. 3A). A lumped internal state, which combines multiple reactions including the binding of auxin to the AFB receptor and related molecular machinery, is represented by  $x$ . IAA protein levels are represented by  $y$ . The model has eight parameters ( $k_{1-8}$ ) that intuitively correspond to the biological processes listed in Table 1. We tested the validity of this interpretation numerically by evaluating fits of the data using different assumptions about which parameters change when different ARFs or IAAs are used in the circuit (*SI Appendix*). Our analysis, and the

interpretation in Table 1, suggests varying the IAA corresponds to tuning three parameters: expression level ( $k_3$ ), auxin-induced degradation ( $k_5$ ), and ARF affinity ( $k_8$ ) of each IAA. The rest of the parameters are independent of the changing components in the circuit except for  $k_7$ , which depends on which ARF is present.

To identify how IAA features influence the observed differences in response behavior, we defined two performance metrics: the preauxin steady state  $g_0$  and the activation time  $\Delta T$  (Fig. 3B). The metrics were plotted against the estimated IAA dependent parameter values for each variant to quantify the sensitivity of the pathway to these parameters. The preauxin steady state  $g_0$  was accurately predicted by  $k_3$  (expression level) and to a lesser extent by  $k_8$  (ARF affinity), whereas, as the model suggests,  $k_5$  did not reliably affect  $g_0$  (Fig. 3C and *SI Appendix*). In other words, the “OFF” state of the circuit before auxin treatment is largely determined by the amount of IAA present and only modestly affected by the affinity of the IAA for the ARF. Experimentally, we also confirmed that ARCs containing the high-expressing IAA1.1, encoded by a yeast codon-optimized *IAA1* (27), had a much lower preauxin steady state than those with IAA1 (*SI Appendix, Fig. S3*). In contrast, activation time  $\Delta T$  was predicted with high accuracy by  $k_5$  alone (Fig. 3D). This result indicates that natural variation in IAA degradation rates—and not similar levels of variation in the other parameters tested here—is sufficient to dramatically alter the dynamics of auxin response.

We next engineered ARC<sup>Sc</sup> variants with pairs of IAAs to test the impact of competition between IAAs on auxin response (Fig. 4A). This architecture was inspired by work suggesting that sequential modules of ARF–IAA partners drive discrete stages in the development of lateral roots (30–32). IAA28, IAA14, IAA12, and IAA3 have all been implicated in lateral root development (Fig. 4B). IAA28 regulates founder-cell specification (33) and IAA14 subsequently drives the initial cell divisions (22). IAA12 is involved in later divisions and outgrowth (30), and IAA3 has been shown to be involved in emergence as well as negative regulation of the IAA14 module (34). Limited expression studies available



**Fig. 3.** Model selection and sensitivity analysis of the auxin response pathway. (A) Gray-box model of the auxin response pathway. The auxin input is represented by the variable  $u$  and the GFP output of the reporter is represented by the variable  $g$ . The variable  $x$  represents a lumped internal state, which combines multiple reactions including the binding of auxin to the AFB receptor, and the variable  $y$  represents IAA protein levels. The model has eight parameters ( $k_{1-8}$ ) that intuitively correspond to the biological processes listed in Table 1. We focused on three parameters with strong effect on ARC dynamics: IAA expression level ( $k_3$ ), rate of auxin-induced IAA degradation ( $k_5$ ), and ARF–IAA affinity ( $k_8$ ). (B) A graphical representation of the two performance metrics used for sensitivity analysis: the preauxin steady state  $g_0$  and the activation time  $\Delta T$  ( $t_{90} - t_{10}$ ). Sample data from one ARC is shown as blue dots with a sample model fit in red. (C and D) Sensitivity analysis of preauxin steady state  $g_0$  (C) and activation time  $\Delta T$  (D) to model parameter values of  $k_3$ ,  $k_5$ , and  $k_8$  for each IAA. Each parameter was varied across its entire range of estimated values derived from our experimental dataset, and all other parameters were held constant. Each IAA is plotted as a single point, and the red line indicates the sensitivity curve computed from the model. The preauxin steady state  $g_0$  was accurately predicted by  $k_3$  (expression level) and to a lesser extent by  $k_8$  (ARF affinity), with  $k_5$  having little effect. In contrast, activation time  $\Delta T$  was predicted with high accuracy by  $k_5$  (auxin-induced degradation) alone. Error bars represent SD ( $n = 2$ ). More details can be found in *SI Appendix*.

**Table 1. Parameter interpretations**

Parameter	Biological interpretation
$k_1$	Assembly of auxin-primed AFB receptor
$k_2$	Basal degradation/dissociation of auxin-primed AFB receptor
$k_3$	IAA expression
$k_4$	Basal degradation/dilution of IAA
$k_5$	Auxin-induced degradation of IAA
$k_6$	Basal degradation/dilution of GFP
$k_7$	GFP expression
$k_8$	ARF affinity of IAA

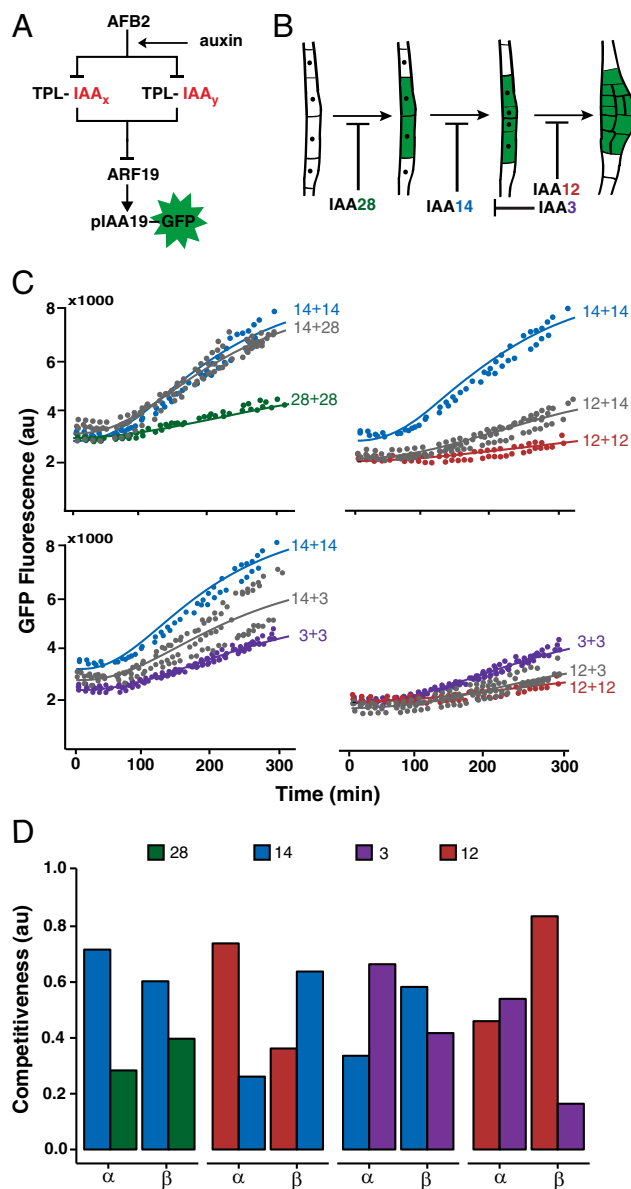
for IAA and ARF family members indicate that multiple family members are often coexpressed in plant cells, raising the question of how specificity or ordering is encoded (14, 15). Our data and those of others suggest that multiple IAAs can interact with and repress the same activator ARF (15). One attractive hypothesis is that a hierarchy of ARC function is encoded by a hierarchy of function within the IAAs themselves.

Although it was not possible to experimentally determine a mechanism for this potential for hierarchical action, we were able to test whether such a pattern can be simulated in a minimal synthetic context. Toward that end, we generated ARC<sup>Sc</sup> variants expressing pairwise combinations of IAA3, IAA12, IAA14, and IAA28. These competition strains made it possible to characterize the impact of an additional IAA on the dynamics of an ARC. Induction curves of ARC<sup>Sc</sup> competition strains indicated that transcriptional dynamics were largely biased toward the IAA that acts in later developmental modules (Fig. 4C). The 12+14 and 12+3 circuits exhibited response dynamics similar to that of a 12+12 circuit, whereas transcriptional dynamics of 14+28 circuits were nearly identical to a 14+14 circuit. The 14+3 transcriptional dynamics showed a high degree of variation from one experiment to another, following response dynamics of either 14+14 or 3+3 circuits. This leads to a dominance hierarchy of  $28 < 14, 3 < 12$ , which matches the proposed sequence of roles in lateral root development.

The preferential response to one of the two coexpressed IAAs suggests dominance relationships between IAAs could play a role in the generation of sequential auxin responses. Novel dynamics created by heterodimers may also be a factor in some cases, although the strong dominance of one IAA over another makes this less likely for most of the examples tested here. The dominance relationships uncovered here could explain how transcriptional dynamics could be switched from one regime to another over time. IAAs are among the earliest auxin response genes (9), suggesting that differences in synthesis and degradation rates among family members could lead to a dramatic change in the cellular complement of IAAs after exposure to auxin. In this way, IAAs may tune the intensity and duration of downstream transcriptional outputs similar to I $\kappa$ B repressors in the mammalian NF- $\kappa$ B pathway (35, 36).

We hypothesized that dominance between IAAs could reflect competition for two key interactions in the transmission of an auxin signal: binding to the AFB and binding to the ARF (Fig. 4A). These interactions are approximately captured by the parameters  $k_5$  (auxin-induced degradation rate) and  $k_8$  (ARF affinity). In plants, where IAAs are strongly and differentially auxin-induced, IAA expression level is also likely to play a critical role in determining which IAA drives transcriptional dynamics. However, because both IAAs in the yeast strains tested here are expressed from the same strong constitutive yeast promoters, we predicted that  $k_3$  (expression level) was unlikely to have a significant role in determining dominance in our assays. Indeed, we found only minor differences in expression levels between coexpressed IAAs, and these small differences had no

correlation to dominance (*SI Appendix*, Fig. S4). Because all of our rate estimates were obtained from experiments where only a single IAA is expressed, these estimates cannot be directly applied to predict competition outcomes in the more complex setting of multiple coexpressed IAAs. Instead, we modified our model to introduce two new parameters,  $\alpha$  and  $\beta$  (*SI Appendix*), which represent the relative contributions of the two competing IAAs to the parameters  $k_5$  and  $k_8$ , respectively. Thus,  $\alpha$  and  $\beta$



**Fig. 4.** IAA coexpression reveals dominance relationships between IAAs. (A) Modified ARF19 circuits were built that contained pairs of IAAs (IAA<sub>x</sub> and IAA<sub>y</sub>). (B) Schematic showing the simplified sequence of lateral root development: founder cell specification, first cell division and patterning/outgrowth. Auxin response modules are proposed to be sequentially triggered by degradation of the indicated IAAs. (C) One IAA could dictate the auxin response dynamics of yeast cells expressing multiple IAAs. Comparisons are shown between circuits containing two copies of the same IAA (colored points) or one copy of each IAA (gray points). Model fits are shown as solid lines. (D)  $\alpha$  (AFB binding competitiveness) or  $\beta$  (ARF binding competitiveness) can affect the contribution of an IAA to ARC dynamics. Estimated  $\alpha$  and  $\beta$  values from the modified model are shown for each of the IAAs tested here. More details can be found in *SI Appendix*.



describe which IAA is dominant in terms of its affinity to either the AFB ( $\alpha$ ) or the ARF ( $\beta$ ). IAAs that demonstrated clear dominance in the behavior of the induction curve in mixed IAA circuits had a higher  $\alpha$  and/or  $\beta$  than the other coexpressed IAA (Fig. 4D). It is worth noting that each of the original parameters, as well as the derived  $\alpha$  and  $\beta$ , encompass multiple rates controlled by distinct biochemical mechanisms. Given the current limits for experimental measurement of each of the constituent rates within these composite parameters, it is impossible to predict the dominance relationship of any IAA pair or to explain observed dominance by an explicit molecular mechanism. Instead, these competition experiments serve to underline the striking complexity possible with even relatively simple circuits and generate a new hypothesis for sequential auxin circuit function during development.

By transplanting auxin response into yeast and defining a small mathematical model able to explain ARC dynamics, we were able to rigorously test how auxin response is controlled and tuned using only the core ARC components. When combined with feedback and interactions with other pathways, such tuning knobs likely have even more dramatic effects. By finding the ARC tuning knobs, this approach generated hypotheses about the evolution of natural ARC variants in plants, as well as showing a path to engineer novel circuit behaviors in synthetic contexts. The inherent modularity of the ARC lends itself to adding additional layers of complexity and quantitative analysis of other challenging features of auxin response, such as feedback or higher-order complexes between ARFs and IAAs. Indeed, recent structural studies indicate the potential for higher-order ARF–IAA complexes than previously suspected (37–39). ARC<sup>Sc</sup> may be a useful tool for quantifying the impact of this structural complexity on response dynamics. In addition, incorporation of additional auxin-responsive promoters could expand the functionality of our system and allow incorporation of combinatorial control by multiple transcription factors. More generally, ARC<sup>Sc</sup> demonstrates how the reconstitution of eukaryotic pathways can serve as a powerful companion to systems biology-driven studies that generate predictions about functional signaling modules.

## Materials and Methods

**Strain Construction.** A library of IAAs was cloned downstream of a GPD promoter in *TRP1* integration vectors using a standard Gateway LR reaction (LRClonase II; Life Technologies) and transformed into W814-29B *MATa* strains (YKL381) containing pGP5G-AFB2 [AFB2 integration at *LEU2*, expression from the GPD promoter (27)]. ARFs were amplified from *Arabidopsis* cDNA and subcloned into the Gateway pDONR221 plasmid using a standard Gateway BP reaction (BP Clonase II; Life Technologies). Each ARF was then subsequently cloned into pGP8A-ccdB (integration at *HIS3*, expression from the *ADH1* promoter) and transformed into a W303-1A *MATa* strain containing the pIAA19-GFP reporter integrated at *URA3* (YKL12). Strains containing AFB2 and an IAA were then mated to strains containing the reporter and an ARF to generate the library of ARC<sup>Sc</sup> variants. See *SI Appendix*, Table S1 for strains used in this study.

**Plasmid Construction.** *Arabidopsis* genes were inserted into single integrating derivatives of pAG Gateway vectors (27, 40) (*SI Appendix*, Fig. S5 and Table S2). *TRP1* targeting single integrating Gateway vectors (27) were further modified to contain an N-terminal TPL repression domain (RD) for auxin response assays, or an N-terminal YFP-RD tag to monitor the degradation of RD–IAA fusions. Repression domains TPL-N100 and TPL-N300 were amplified from *A. thaliana* Columbia ecotype cDNA. For the GFP auxin induction reporter plasmid, 300 bp immediately upstream of the IAA19 start codon was amplified from

*Arabidopsis* gDNA and inserted upstream of GFP in the pAG306 (*URA3*, multiply inserting) targeting vector.

**Yeast Methods.** Standard yeast drop-out and yeast extract–peptone–dextrose plus adenine (YPAD) media were used, with care taken to use the same batch of synthetic complete (SC) media for related experiments. A standard lithium acetate protocol (41) was used for transformations of digested plasmids. All cultures were grown at 30 °C with shaking at 220 rpm.

**Flow Cytometry.** Fluorescence measurements were taken with a BD Accuri C6 flow cytometer and CSampler plate adapter using an excitation wavelength of 488 and an emission detection filter at 533 nm. A total of 10,000 events above a 400,000 FSC-H threshold (to exclude debris) were measured for each sample and data exported as FCS 3.0 files for processing using the flowCore R software package and custom R scripts (27).

**Induction Assays.** A freshly grown colony for each strain was inoculated in SC and the cell density (in events per microliter) was estimated using cytometry data gated for yeast by a custom R script (27). Each culture was then diluted to 0.25 events per microliter in 12 mL of SC. This dilution was split into duplicate 5-mL aliquots and incubated for 16 h. At 16 h, duplicate aliquots were combined, mixed, and resplit into two tubes and initial measurements taken in the cytometer (starting density was ~200 events per microliter). For each strain, one replicate was mock-treated [95% (vol/vol) ethanol] and one replicate was treated with 10  $\mu$ M indole-3-acetic acid. Immediately after auxin addition, fluorescence for the 0-min time point was recorded and subsequent measurements acquired at 10-min intervals for 5 h. Up to 12 strains were measured in parallel.

**Steady-State Expression Measurements.** A freshly grown colony was inoculated into 1 mL of SC in a 96-well plate and grown to stationary overnight. The next morning each culture was diluted into fresh SC and fluorescence measurements taken ~6 h later.

**Microscopy.** Yeast cells grown overnight in SC at 30 °C were diluted 1:100 in SC, incubated for 4–5 h, and then diluted 1:20 before loading onto a Y04D plate (CellASIC). Using the CellASIC-ONIX microfluidic system and associated software, cells were treated with 10  $\mu$ M auxin in SC medium. An inverted Nikon Eclipse Ti microscope with a 60 $\times$ , numerical aperture and 1.4 oil objective was used to image the yeast cells at 1-h intervals using a YFP-HYQ and CFP bandpass filter (Nikon) with a CoolSNAP HQ2 14-bit camera.

**Expression Analysis.** Yeast cells were grown in YPAD and collected at an OD of 0.9. RNA was extracted using the RiboPure RNA purification kit (Life Technologies). One microgram of RNA was used for cDNA synthesis using the iScript cDNA Synthesis kit (Bio-Rad). Samples were analyzed using the iQ SYBR Green Supermix (Biorad) reactions in a C100 Thermal Cycler fitted with a CFX96 Real-Time Detection System (Biorad). Relative expression levels were calculated using the formula  $(E_{\text{target}})^{-C_{\text{Ptarget}}}/(E_{\text{ref}})^{-C_{\text{Pref}}}$  (42) and normalized to the *ACT1* reference gene.

**Quantitative Analysis.** A gray-box modeling approach was used to identify a minimal mathematical model for the auxin response characterized by ARC<sup>Sc</sup> variants. A more detailed description of the modeling and quantitative analysis is found in *SI Appendix*. The annotated code and supporting data are available for download at <http://klavinslab.org/data.html>.

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1. Finet C, Berne-Dedieu A, Scutt CP, Marlétaz F (2013) Evolution of the ARF gene family in land plants: Old domains, new tricks. *Mol Biol Evol* 30(1):45–56.
2. Paponov IA, et al. (2009) The evolution of nuclear auxin signalling. *BMC Evol Biol* 9:126.
3. Szemenyei H, Hannon M, Long JA (2008) TOPELESS mediates auxin-dependent transcriptional repression during *Arabidopsis* embryogenesis. *Science* 319(5868):1384–1386.
4. Ulmasov T, Murfett J, Hagen G, Guilfoyle TJ (1997) Aux/IAA proteins repress expression of reporter genes containing natural and highly active synthetic auxin response elements. *Plant Cell* 9(11):1963–1971.

5. Dharmasiri N, Dharmasiri S, Estelle M (2005) The F-box protein TIR1 is an auxin receptor. *Nature* 435(7041):441–445.
6. Kepinski S, Leyser O (2005) The *Arabidopsis* F-box protein TIR1 is an auxin receptor. *Nature* 435(7041):446–451.
7. Ulmasov T, Hagen G, Guilfoyle TJ (1999) Activation and repression of transcription by auxin-response factors. *Proc Natl Acad Sci USA* 96(10):5844–5849.
8. Tiwari SB, Hagen G, Guilfoyle T (2003) The roles of auxin response factor domains in auxin-responsive transcription. *Plant Cell* 15(2):533–543.

9. Abel S, Nguyen MD, Theologis A (1995) The PS-IAA4/5-like family of early auxin-inducible mRNAs in *Arabidopsis thaliana*. *J Mol Biol* 251(4):533–549.
10. Remington DL, Vision TJ, Guilfoyle TJ, Reed JW (2004) Contrasting modes of diversification in the Aux/IAA and ARF gene families. *Plant Physiol* 135(3):1738–1752.
11. Dharmasiri N, et al. (2005) Plant development is regulated by a family of auxin receptor F box proteins. *Dev Cell* 9(1):109–119.
12. Lokere AS, Weijers D (2009) Auxin enters the matrix—assembly of response machineries for specific outputs. *Curr Opin Plant Biol* 12(5):520–526.
13. Pierre-Jerome E, Moss BL, Nemhauser JL (2013) Tuning the auxin transcriptional response. *J Exp Bot* 64(9):2557–2563.
14. Rademacher EH, et al. (2012) Different auxin response machineries control distinct cell fates in the early plant embryo. *Dev Cell* 22(1):211–222.
15. Vernoux T, et al. (2011) The auxin signalling network translates dynamic input into robust patterning at the shoot apex. *Mol Syst Biol* 7:508.
16. Del Bianco M, Kepinski S (2011) Context, specificity, and self-organization in auxin response. *Cold Spring Harb Perspect Biol* 3(1):a001578.
17. Nishimura K, Fukagawa T, Takisawa H, Kakimoto T, Kanemaki M (2009) An auxin-based degron system for the rapid depletion of proteins in nonplant cells. *Nat Methods* 6(12):917–922.
18. Tian Q, Reed JW (1999) Control of auxin-regulated root development by the *Arabidopsis thaliana* SHY2/IAA3 gene. *Development* 126(4):711–721.
19. Nagpal P, et al. (2000) AXR2 encodes a member of the Aux/IAA protein family. *Plant Physiol* 123(2):563–574.
20. Ouellet F, Overvoorde PJ, Theologis A (2001) IAA17/AXR3: Biochemical insight into an auxin mutant phenotype. *Plant Cell* 13(4):829–841.
21. Rogg LE, Lasswell J, Bartel B (2001) A gain-of-function mutation in IAA28 suppresses lateral root development. *Plant Cell* 13(3):465–480.
22. Fukaki H, Tameda S, Masuda H, Tasaka M (2002) Lateral root formation is blocked by a gain-of-function mutation in the SOLITARY-ROOT/IAA14 gene of *Arabidopsis*. *Plant J* 29(2):153–168.
23. Hamann T, Benkova E, Bäurle I, Kientz M, Jürgens G (2002) The *Arabidopsis* BODENLOS gene encodes an auxin response protein inhibiting MONOPTEROS-mediated embryo patterning. *Genes Dev* 16(13):1610–1615.
24. Fukaki H, Taniguchi N, Tasaka M (2006) PICKLE is required for SOLITARY-ROOT/IAA14-mediated repression of ARF7 and ARF19 activity during *Arabidopsis* lateral root initiation. *Plant J* 48(3):380–389.
25. Okushima Y, Fukaki H, Onoda M, Theologis A, Tasaka M (2007) ARF7 and ARF19 regulate lateral root formation via direct activation of LBD/ASL genes in *Arabidopsis*. *Plant Cell* 19(1):118–130.
26. Wilmoth JC, et al. (2005) NPH4/ARF7 and ARF19 promote leaf expansion and auxin-induced lateral root formation. *Plant J* 43(1):118–130.
27. Havens KA, et al. (2012) A synthetic approach reveals extensive tunability of auxin signaling. *Plant Physiol* 160(1):135–142.
28. Ljung L (1999) *System Identification: Theory for the User* (Prentice Hall, Upper Saddle River, NJ), 2nd Ed.
29. Neuert G, et al. (2013) Systematic identification of signal-activated stochastic gene regulation. *Science* 339(6119):584–587.
30. De Smet I, et al. (2010) Bimodular auxin response controls organogenesis in *Arabidopsis*. *Proc Natl Acad Sci USA* 107(6):2705–2710.
31. De Smet I (2012) Lateral root initiation: One step at a time. *New Phytol* 193(4):867–873.
32. Lavenus J, et al. (2013) Lateral root development in *Arabidopsis*: Fifty shades of auxin. *Trends Plant Sci* 18(8):450–458.
33. De Rybel B, et al. (2010) A novel aux/IAA28 signaling cascade activates GATA23-dependent specification of lateral root founder cell identity. *Curr Biol* 20(19):1697–1706.
34. Goh T, Kasahara H, Mimura T, Kamiya Y, Fukaki H (2012) Multiple AUX/IAA-ARF modules regulate lateral root formation: The role of *Arabidopsis* SHY2/IAA3-mediated auxin signalling. *Philos Trans R Soc Lond B Biol Sci* 367(1595):1461–1468.
35. Kearns JD, Basak S, Werner SL, Huang CS, Hoffmann A (2006) I $\kappa$ B provides negative feedback to control NF- $\kappa$ B oscillations, signaling dynamics, and inflammatory gene expression. *J Cell Biol* 173(5):659–664.
36. Tian B, Nowak DE, Brasier AR (2005) A TNF-induced gene expression program under oscillatory NF- $\kappa$ B control. *BMC Genomics* 6:137.
37. Boer DR, et al. (2014) Structural basis for DNA binding specificity by the auxin-dependent ARF transcription factors. *Cell* 156(3):577–589.
38. Korasick DA, et al. (2014) Molecular basis for AUXIN RESPONSE FACTOR protein interaction and the control of auxin response repression. *Proc Natl Acad Sci USA* 111(14):5427–5432.
39. Nanao MH, et al. (2014) Structural basis for oligomerization of auxin transcriptional regulators. *Nat Commun* 5:3617.
40. Alberti S, Gitler AD, Lindquist S (2007) A suite of Gateway cloning vectors for high-throughput genetic analysis in *Saccharomyces cerevisiae*. *Yeast* 24(10):913–919.
41. Gietz RD, Woods RA (2002) Transformation of yeast by lithium acetate/single-stranded carrier DNA/polyethylene glycol method. *Methods Enzymol* 350:87–96.
42. Pfaffl MW (2001) A new mathematical model for relative quantification in real-time RT-PCR. *Nucleic Acids Res* 29(9):e45.