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Cell identity specification in plants: lessons from flower development

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Abstract

Multicellular organisms display a fascinating complexity of cellular identities and patterns of diversification. The concept of ‘cell type’ aims to describe and categorize this complexity. In this review, we discuss the traditional concept of cell types and highlight the impact of single-cell technologies and spatial omics on the understanding of cellular differentiation in plants. We summarize and compare position-based and lineage-based mechanisms of cell identity specification using flower development as a model system. More than understanding ontogenetic origins of differentiated cells, an important question in plant science is to understand their position- and developmental stage-specific heterogeneity. Combinatorial action and crosstalk of external and internal signals is the key to cellular heterogeneity, often converging on transcription factors that orchestrate gene expression programs.

Keywords: Cell lineage, cellular differentiation, flower development, phytohormone, plant cell type, positional regulation, transcription factor.

Introduction

How to define a cell type in plants

The cell type concept

A cell type is classically defined by its phenotype and function. More recently, molecular signatures such as gene expression profiles and epigenetic patterns have been introduced to assist in defining and distinguishing cell types (Brady et al., 2007; Yadav et al., 2014; X. Zhang et al., 2018). Less than 20 major cell types are classically assigned in vascular plants (Steeves and Sussex, 1989). Thus classical cell type concepts aim to generalize cellular identity and may not entirely cover the complexity and cellular diversity of cells within the major tissue and organ types in plants.

In 1665, Robert Hooke first discovered the cells in a piece of cork tissue, calling them ‘pores’ and later naming those structures ‘cells’, implying both the form and function of the cells (Mazzarello, 1999). The botanist Matthias Jakob Schleiden (1804–1881) later suggested that every structural element of plants is composed of cells or their products (Schleiden, 1838). With the improvement of microscopy and anatomy, cell types have historically been defined by morphology, localization, ontogeny, and function (e.g. Carter et al., 1986). Concepts of plant development have traditionally been strongly entangled with the question of ontogenetic (and later evolutionary) origins of
morphological structures. Classical concepts aim to trace the origins of ‘novel’ organs or structures from pre-existing structures, for example by exploring and conceptualizing ‘hidden’ evolutionary relationships and commonalities of shoots, and vegetative and floral organs (see, for example, Arber, 1950). With the origins of molecular genetics, evolutionary developmental biology has revolutionized our understanding of the molecular mechanisms underlying the evolutionary origins and diversification of plant organs and cell types. Gene duplications followed by sub- and neofunctionalization, co-option of genes, and gene regulatory modules were found to contribute to the striking morphological complexity in higher plant species. However, in very practical terms, this creates challenges, for instance since many developmental control genes act in more than one developmental process linked to evolutionary history. For example, many genes controlling vegetative leaf development are also expressed in floral organs, and their activities are modulated to give rise to specific structures of floral organs. This often makes it hard to identify marker genes that are very specifically expressed in only one cell type or differentiation stage. One might compare the problem of defining a ‘cell type’ in plants with that of the ‘species concepts’ in plants (Christenhusz, 2020), in that it can be difficult to distinguish cell ‘types’ from ‘states’, and cells with the same ‘identity’ may appear very different from each other in terms of gene expression profiles. The challenge is thus to deduce cell ‘history’, similar to deducing the natural history of species.

The idea that a generalized cell type concept does not fully reflect the cellular diversity in form and function in plants can be illustrated by the cell types that together constitute the epidermis tissue (Fig. 1A). The plant epidermis forms the outer cell layer of the plant like the skin of the human body, and its primary functions are to protect inner tissues and to act as the communication and exchange surface with the environment (Glover, 2000). Epidermal pavement cells of Arabidopsis leaves are usually shaped like the interlocking pieces of a jigsaw puzzle. Pavement cells in sepals instead are boxy and differentiated into giant and small cells, while the adaxial epidermal cells in petals are conical and uniform (Glover, 2000; Roeder et al., 2010; Huang and Irish, 2015). Meristemoids within the abaxial epidermis of developing leaves and in floral organs give rise to stomata. Trichomes develop typically at regular distances in the abaxial side of leaves and sepals; however, trichome anatomy is highly variable among organs of a plant. Cellular characteristics reflect functional differences of epidermis cells among organs. For example, epidermal cells in petals enhance attractiveness to pollinators. Specialized epidermal cells in carpels, the stigma, receive and induce the germination of pollen grains (Glover and Martin, 1998; Balanzà et al., 2014). Together, the example of the epidermis shows that the cell type composition of tissues can be variable across organs in a plant, and that individual cell types can have different phenotypes and function. Even more, cellular morphology and cell type frequencies within tissues can be plastic and affected by environmental factors (see, for example, Wardlaw, 1952; Casson and Gray, 2008).

Cellular differentiation in plants can be reversible under specific conditions. Plants are sessile organisms and have to cope with injuries caused by environmental stimuli or biotic attacks.

**Fig. 1.** Diversity of epidermal cell types and control of trichome development. (A) Epidermal cell types, including stomatal guard cells and pavement cells on a leaf, trichome cells on a leaf, papillate cells on a carpel (scale bar, 500 μm), giant cells and small cells on a sepal (scale bar, 100 μm), and conical cells on a petal (scale bar, 15 μm). (B) Movement of R3 MYB functions as a positional signal for trichome patterning (after Grebe, 2012). AG serves as an upstream lineage signal in flowers to suppress trichome formation. Solid lines represent direct regulation of target genes, and dashed lines represent movement. SEM images in (A) are reprinted with permission from Riglet et al. (2020) (‘papillate cells’), Meyer et al. (2017) (‘giant cells and small cells’), Yang et al. (2019) (‘conical cells’), Emmanuel Boutet (Plant biocurator for UniProtKB/Swiss-Prot) (‘stomata and pavement cells’), and Stefan Eberhard (University of Georgia, USA; provided by Wellcome collection under a CC BY-NC 4.0 licence) (‘trichomes’).
Somatic cells can be reprogrammed to regenerate new organs or repair damaged tissues (Ikeuchi et al., 2019). Laser-assisted elimination of cells in Arabidopsis root triggers the cells adjacent to the injury to re-activate stem cell pathways, change the cell division orientation accordingly, and acquire the cell fates of the missing cells to replace them (Marhava et al., 2019). Other kinds of stresses, such as osmotic, heavy metal ion, and dehydration stress, can also induce plant cells to regenerate (Ikeda-Iwai et al., 2003). The questions remain of which factors and mechanisms determine stem cell identity in plants, how different stem cell niches can be distinguished from one another (beyond the activity of some master regulators), and what are the exact mechanisms underlying dedifferentiation versus terminal differentiation of specific cell types.

Altogether, the concept of ‘cell type’ aims to simplify the complex nature of cellular diversity in multicellular plants (and animals). Therefore, it is important to understand the origins and consequences of heterogeneity within and among cell types, and the mechanisms underlying their organ- and environment-specific modulation.

Cell lineage versus positional signals

Understanding the principles that govern cell type specification in multicellular organisms is one of the major challenges in developmental biology. The fundamental concept of ‘epigenetic landscape’ introduced by Waddington in 1957 visualizes cell differentiation as a ball rolling down a valley in a landscape that is sculpted by regulatory genes and their combinatorial activities (Waddington, 1957). This path forms the developmental trajectory or ‘lineage’ of the cell defined by its start position and by dynamic but predictable changes in gene activities determined by the regulators. In the animal field, reconstruction of cell lineage history is often used to understand cellular differentiation programs (Kretzschmar and Watt, 2012; Morris, 2019). A classical example for utilizing the lineage concept in plants is stomata development, which is initiated from a lineage-specific stem cell via a series of defined asymmetric cell divisions controlled by consecutively acting regulatory factors and enforced by cell–cell signaling (Han and Torii, 2016).

However, many experiments have shown that the developmental fate of a plant cell does not depend strictly on its lineage, but on its exact position within the growing plant body (Stewart and Dermer, 1975; van den Berg et al., 1995; Szymkowiak and Sussex, 1996; Berger et al., 1998; Kidner et al., 2000; Costa and Shaw, 2006; Costa, 2016). In early studies, plant scientists utilized mosaic or chimera experiments to trace the lineage of a cell type by labeling cell clones with genetically phenotypic traits, such as ploidy level or albinism, and tracking the mitotic descendants of marked cells (Poethig, 1987, 1989; Dawe and Freeing, 1991; Irish, 1991; Scheres, 2001). Chimera studies have shown that plant cells do not follow strict lineages, and their fates are not pre-determined but rely on positional information or cell–cell interactions (Szymbkowiak and Sussex, 1996). For instance, during leaf development, epidermal cells regularly undergo anticlinal divisions to form the outermost layer. However, in chimeras, occasionally small sectors derived from periclinal divisions of epidermal cells have been observed. These cells, although of epidermis lineage, adapt to their new position and differentiate as internal mesophyll cells (Stewart and Burk, 1970; Stewart and Dermer, 1975). This phenomenon has also been observed in the root meristem. If a cortical initial cell is laser ablated, the adjacent pericycle cell switches its fate and takes the position of the cortical cell to continue forming corresponding cell files (van den Berg et al., 1995). Plant cells may alter their identity when positional signals are changed. This is exemplified by observations on wound healing and de-/re-differentiation. In the past ~30 years, the molecular nature of many signaling mechanisms controlling cellular differentiation and cell type specification in plants has been elucidated. Combinations of genetic analyses and computational modeling have allowed us to gain insights into the regular nature of the positioning mechanisms.

Signals controlling cell type specification in plants

Different mechanisms explaining the relationships between relative position, spatial patterns, and cell fate in developing organisms have been proposed. A major and often considered mechanism is the formation of gradients of signaling molecules, such as morphogens, that result in the specification of distinct cell fates in a concentration-dependent manner (Wolpert, 1996). Mobile signaling molecules in plants reported to control patterning include phytohormones, mobile transcription factors (TFs), non-coding RNAs, and small signaling peptides. A second general model for explaining position-dependent cell specification does not rely on a gradient, but on biochemical signaling between neighboring cells in ‘boundary’ regions, essentially resulting in self-organization of the system (Sharpe, 2019). The idea that mechanical signals play a role in cellular differentiation and patterning in plants has been recognized for a long time (see, for example, Arber, 1950), but is also an exciting focus of ongoing research.

The phytohormone auxin in flower development

Graded auxin accumulation has been shown to play important roles in developmental patterning, while a primary effect is on cell expansion (Leyser, 2018), thereby controlling processes such as vascular development and specification of floral meristem founder cells. Auxin is actively transported in a polar manner between cells via transport proteins, such as PIN-FORMED (PIN) efflux carriers. The pin1 mutant fails to produce flowers and presents a pin-shaped inflorescence (Okada et al., 1991). Flower initiation can be rescued when indole acetic acid (IAA)
is applied exogenously to the pin1 mutant (Reinhardt et al., 2000). Cellular specificity of auxin responses is linked to the complexity of auxin sensing and response pathways in the cell (Leyser, 2018). After binding to receptors of the TRANSPORT INHIBITOR RESPONSE1/AUXIN SIGNALING FBOX (TIR1/AFB) family, auxin controls gene expression in the nucleus via activation of AUXIN RESPONSE FACTOR (ARF) TFs by targeted degradation of AUXIN/INDOLE-3ACETIC ACID (AUX/IAA) repressor proteins (Weijers and Wagner, 2016). The initiation of flower meristems is marked by the formation of local maxima of auxin (Benková et al., 2003; Heisler et al., 2005). A series of auxin responses is started by the activation of ARF5 (also known as MONOPTEROS), which promotes the expression of TFs mediating floral meristem identity, including LEAFY (LFY) (Yamaguchi et al., 2013, 2016). LFY further promotes the expression of APETALA1 (AP1) (Parcy et al., 1998; Wagner et al., 1999). LFY and AP1 play a central role in flower meristem specification and they regulate a large number of downstream genes for flower formation (Parcy et al., 1998; Kaufmann et al., 2010b; Winter et al., 2011). Besides a role in early activation of flower development, auxin has an instructive role in patterning of organs within the flower, as revealed by higher order yuca mutants that are defective in auxin biosynthesis (Zhao et al., 2001; Tobera-Santamaria et al., 2002; Cheng et al., 2006). Besides this, other ARF TFs were found to direct organ polarity as well as stamen differentiation (Przemeck et al., 1996; Sessions et al., 1997; Nagpal et al., 2005; Simonini et al., 2016; K. Zhang et al., 2018). To explain the multiple functions of auxin, we need a better understanding of the cell type-specific composition of the auxin response machineries. This requires knowledge of quantitative abundance of specific auxin signaling factors across cellular differentiation in the flower, direct ARF targets, along with their affinity and specificity for protein–protein and protein–DNA interactions. At the same time, the chromatin landscape of a cell type may affect auxin response. Addressing these questions will help to understand how transient peaks in auxin concentration can trigger cellular patterning responses, and to distinguish direct effects from indirect downstream developmental decisions.

**Mobile regulatory molecules in flower development**

Small peptides are usually <20 amino acids long in the mature form and rarely longer than 120 amino acids as a precursor, and they are often present in very low physiological concentrations in the nanomolar range (Murphy et al., 2012). Intercellular communications via mobile peptides and receptor signaling cascades have been identified as important regulators of cell identity in meristems and during organ tissue differentiation (see, for example, Lee et al., 2015; Lin et al., 2017). A classical example for this is the CLE (CLV3/ESR3, EMBRYONIC SURROUNDING REGION) family which controls stem cell identity in the meristems of the plant. By this, the CLE family impacts flower development, since the size of the stem cell niche in the floral meristem has a direct effect on the number of floral organs that are produced by it. The clv3 mutant leads to overproliferation of the shoot apical meristem (SAM) cell population, while overexpression of CLV3 eliminates the stem cell niche and results in termination of shoot development (Fletcher et al., 1999; Kondo et al., 2006). CLV3 is shown to be expressed in stem cells at the meristem apex. It diffuses toward inner layers of the meristem and organizing center, where it interacts with the leucine-rich receptor-like kinase (LRK-RLK) CLV1 and related proteins to restrict the expression of a homeodomain TF protein WUSCHEL (WUS), which is a positive regulator of the stem cell population (Mayer et al., 1998; Fletcher et al., 1999; Schoof et al., 2000; Kinoshita et al., 2010). Furthermore, WUS can also migrate toward the CLV3-expressing cell layers, where it activates CLV3 expression by directly binding to its genomic region (Yadav et al., 2011). This negative feedback loop between WUS and CLV3 is well established to maintain a proper cell population in the SAM. Consequently, this pathway regulates meristem size and by this controls the number of organs produced in a flower: clv3 mutants typically produce more floral organs, while overexpression results in flowers without inner whorls of stamens or carpels (Fletcher et al., 1999).

miRNAs function by binding to complementary sites in mRNA molecules to trigger the degradation or translational inhibition of target genes. Small RNAs can diffuse across tissues, resulting in concentration gradients, thereby potentially mediating tissue patterning through dose-dependent activity (D’Ario et al., 2017). miRNAs have been found to regulate the activity of TFs and other regulatory proteins in flower development. A classical example is miR172, which accumulates in the SAM at the onset of flowering, preventing the transition of the center of the SAM into a flower meristem by inhibiting AP2 expression (Aukerman and Sakai, 2003; Chen, 2004). miR172 also accumulates in the center of the flower primordia to suppress AP2, avoiding AP2 and AG co-expression and thereby setting the boundary between petal and stamen whorls. miR172 is itself down-regulated by AP2, establishing a negative feedback loop that is essential for the correct specification of organ identity (Zhao et al., 2007; Wollmann et al., 2010). Several miRNAs are involved in the process of floral organ development partially by crosstalking with plant hormones. For example, miR167 directly targets the transcripts of AUXIN RESPONSE FACTOR 6 (ARF6) and AUXIN RESPONSE FACTOR 8 (ARF8) to regulate stamen filament and pollen development (Wu et al., 2006). Furthermore, miR393 has been shown to target TIR1/AFB proteins, which are critical components of auxin signaling transduction (Parry et al., 2009).

Concentration gradients formed by mobile TFs also play roles in developmental patterning (Vadde et al., 2020). For instance, trichomes are equally distributed on leaves, because they can inhibit neighboring cells from acquiring trichome identity, and this program is modified in floral organs. In trichome-forming cells, the TF GLABRA 3 (GL3) forms a complex with
GL1, activating not only the expression of positive regulators for trichome cell fate determination but also the expression of R3–MYB TFs. R3–MYB proteins form a complex with GL3 in the neighboring cells, thus preventing the formation of the GL3–GL1 complex (Pattanaik et al., 2014). Cell fate decision during root hair development shares a similar mechanism, since movement of similar regulatory proteins between root hair and non-root hair cells reinforces their identity (Salazar-Henao et al., 2016) (Fig. 1B). During flower development, trichome formation in carpels is suppressed by the floral homeotic AGAMOUS (AG) TF via several target pathways, including direct repression of GL1 (Ó’Maoléidigh et al., 2018).

Together, these examples show that cell to cell signaling impacts developmental patterning and cell identity specification. Developmental programs thus represent the sum of ‘endogenous’ cellular status and different signaling cascades that emerge from cell to cell signaling.

Mechanical signals in floral organ differentiation

The role of mechanical forces in developmental patterning has long been acknowledged. For example, in The natural philosophy of plant form (1950), Agnes Arber states ‘Judging merely from inspection, it looks as if limitation of the space into which [plants] can expand, and the actual pressure which the developing parts exert upon one another, must be the efficient cause [for different shapes of various members of the plant body]’.

Plant cells sense internal and external forces that can be perceived as growth signals that have important effects on and affect the shape of cells and organs (Hervieux et al., 2017; Sapala et al., 2018). Mechanical signals defining epidermal cell morphology are at least in part perceived by stress-dependent, katanin-mediated alignment of microtubules in the cytoplasm that in turn guide cellulose-synthesizing complexes in the apoplast (Hamant et al., 2008; Jacques et al., 2013; Sampathkumar et al., 2014). Sepal development has been used as a model to study the contribution of mechanical signals to local growth because of its high variability in growth rate within tissues and its robustness in the final shape of the organ (Tauriello et al., 2015; Hervieux et al., 2017). Trichome precursor cells in sepals initially grow and expand much faster than surrounding epidermal cells, which potentially can distort the final shape of the sepal. However, neighboring cells of trichome precursors organize their microtubule arrays according to the mechanical changes caused by trichome precursor cells, and thus grow at a reduced rate to maintain sepal shape (Hervieux et al., 2017). Mechanical forces are involved in the control of sepal size. Tangential tension at the tip of the sepal causes the arrest of growth at the tip by reorientation of the microtubule array (Hervieux et al., 2016).

Another example of mechanical signaling comes from the SAM, where cells orient their cortical microtubules along the lines of mechanical stress generated during tissue formation, and this then affects the mechanical properties of the cell, thus establishing a feedback loop (Uyttewaal et al., 2012). Mechanical forces crosstalk with biochemical signals, for instance in the generation of phyllotactic patterns. Auxin minima reside in organ boundaries, and these regions are characterized by the expression of a specific group of TFs, such as CUP-SHAPED COTYLEDON 1 (CUC1), CUC2, and CUC3, which limit cell growth and thus create a creased shape in the boundaries (Heisler et al., 2005; Rast and Simon, 2008). Microtubule arrangement and polar auxin transport in the boundaries are regulated by mechanical forces (Heisler et al., 2010; Landrein et al., 2015). Mechanical signals interplay not only with auxin, but also with miRNA regulation. The expression of CUC genes in organ boundaries is regulated by both miRNA and mechanical forces (Fal et al., 2016). Interestingly, it was reported that the shape of nuclei correlates with cell shape and size in plants (Meier et al., 2016). Mechanotransduction can affect the shape of the nucleus via interaction with the cytoplasmic microtubule cytoskeleton, and ‘nuclear stiffness’ may affect transcriptional regulation and gene activity (Finan and Guilak, 2010; Goswami et al., 2020b; Irianto et al., 2013; Lovett et al., 2013).

In summary, mechanical forces provide positional signals by regulating organ growth rate, cell and organ shape, thereby contributing to cellular differentiation and cell type frequencies within organs.

Insights into lineage-based mechanisms

The cell lineage concept suggests that a cell’s fate is determined early by its progenitors. Cells pass on specific cell fate decisions to their progeny across cell division (Stent, 1985). Although positional mechanisms play an important role in cell fate specification in plants, lineage-based mechanisms cannot be neglected. If cell fate is specified and maintained solely by positional information, cells would have to re-establish their expression programs based on their new position at every division (Costa, 2016). However, data from roots suggest that cells that misexpress a cell identity marker gene frequently pass on the ‘wrong’ identity to their progeny (Costa, 2016). Once cell fate is specified by positional information, it is clonally maintained by lineage until they receive new positional input (Costa, 2016). The critical role of TFs and epigenetic regulators also indicates the existence of a lineage-based component in plant cell fate determination. Stomatal differentiation follows an evident cell lineage from meristemoid mother cells to mature guard cells (Yang and Sack, 1995; Larkin et al., 1997; Nadeau and Sack, 2002). Several basic helix–loop–helix (bHLH) TFs—SPEECHLESS (SPCH), MUTE, and FAMA—function at each stage to control the lineage. The overexpression of SPCH induces extra asymmetric divisions and the production of excess stomata (MacAlister et al., 2007). Ectopic MUTE expression in the petal that is normally devoid of stomata converted petal epidermal cells into stomata (Pillitteri et al., 2008).
Induced expression of FAMA transforms cotyledon epidermal cells into guard cells (Ohashi-Ito and Bergmann, 2006).

In more general terms, key TFs have been identified that trigger cell lineage differentiation, while the initial activation of these TFs may be dictated by positional signaling at early stages in plants and animals (Scott and Carroll, 1987; St Johnston and Nüsslein-Volhard, 1992; Scheres, 2001).

**Pioneer transcription factors, organ identity, and cell type specification**

Over the past decades, genetic analyses have identified floral regulators and discovered detailed insights into how they interact and cooperate to control flower development. Plant morphogenesis depends on the combinatorial interplay of TFs to mediate distinct and dynamic spatiotemporal gene expression, associated with feedback control (Zik and Irish, 2013). Floral organ specification essentially requires modification of leaf developmental pathways, including changes in growth, cell type frequencies, cellular morphologies (e.g. trichomes in sepals versus leaves; conical cells in the petal epidermis) (Fig. 1A), and the establishment of flower-specific cell and tissue types that are not found in leaves (e.g. in reproductive organs).

Cell type specification requires orchestrated changes in global gene expression programs. So-called ‘pioneer TFs’ control cell type programming and reprogramming by promoting chromatin opening to make it accessible for other TFs (Zaret and Carroll, 2011; Iwafuchi-Doi and Zaret, 2014, 2016; Zaret et al., 2019). Growing evidence suggests that pioneer TFs contribute to the regulation of developmental switches in plants (see, for example, Tao et al., 2017, 2019; Lai et al., 2018).

Several key TFs are required for the leaf-to-flower transition. LFY specifies flower meristem identity. Combinatorial expression of LFY and WUS induces the generation of floral organs on primary and lateral root tips (Gallois et al., 2004), and inducible expression of LFY in root explants is sufficient to trigger flower formation, bypassing elaboration of a shoot (Weigel et al., 1992; Wagner et al., 2004). Furthermore, expression of the LFY co-regulator UNUSUAL FLORAL ORGANS (UFO) fused to a VP16 activation domain resulted in ectopic formation of flowers and inflorescences in vegetative leaves in the presence of a functional LFY gene (Risseeuw et al., 2013). Protein oligomerization via a SAM domain enabled LFY to bind to closed chromatin regions (Sayou et al., 2016). The functions of the floral meristem factors LFY and AP1 are closely linked, and they regulate each other’s expression. For example, the expression level of LFY is reduced in ap1 cal double mutants, and the onset of expression of AP1 is delayed in the lfy mutant (Weigel and Nilsson, 1995). The ap1 mutant can suppress the terminal flower phenotype of the constitutive expression of LFY (Weigel and Nilsson, 1995). This cross-regulation is mediated by direct promoter interactions (Kaufmann et al., 2010b; Winter et al., 2011). Recent work also explains how LFY binds DNA in a nucleosomal context and enhances chromatin accessibility at its target loci such as AP1 (Jin et al., 2021; Lai et al., 2021). The activation of LFY and AP1 is furthermore more under positional control by ARF5 (Wagner et al., 1999; Yamaguchi et al., 2013, 2016).

Floral organ identity is specified by homeotic TFs of the MADS-box family that interact in a combinatorial manner to specify different types of floral organs (Causier et al., 2010). Combined loss of function of floral homeotic proteins results in the transformation of all floral organs into cauline leaf-like organs (Bowman et al., 1991). Furthermore, the redundantly acting SEPALLATA (SEP) TFs are essential for the specification of all floral organ types. Accordingly, loss of SEP function results in the conversion of the floral organs into leaf-like structures (Pelaz et al., 2000). SEP proteins act as mediators of higher order complex formation of other homeotic TF classes (Theißen and Sauder, 2001; Immink et al., 2009). According to the floral quartet model, petals are specified by a complex consisting of AP1, APETALA3 (AP3), PISTILLATA (PI), and SEP proteins. In contrast, stamens are specified by a complex of AP3, PI, SEP, and AG proteins. Sepals are specified by complexes formed by SEP and AP1 proteins, while carpels are specified by a SEP/AG tetramer (Honma and Goto, 2001; Theissen, 2001).

Cooperative and combinatorial interactions are important for floral homeotic TFs and may facilitate their action as pioneer factors (Kaufmann and Airoldi, 2018). Vegetative leaves of transgenic plants that constitutively express SEP3–AP3–PI or AP1–AP3–PI are converted into petals, showing that these TFs are not only required but also sufficient to specify floral organ identity (Honma and Goto, 2001). SEM revealed that cells on both the abaxial and adaxial surface of the converted rosette leaves closely resembled cells on the surface of the petals (Pelaz et al., 2001). Cauline leaves of AP3–PI–SEP3–AG ectopic expression lines were converted into staminoid organs, and all floral organs are transformed into stamens or staminoid organs. Homeoctically converted cauline leaves of these transgenic plants consist of two distinct regions whose epidermal cells exhibit a morphology similarity to that of anthers and filaments, respectively (Honma and Goto, 2001). Combination of genome-wide ChIP-seq and Dnase I-seq time-series experiments suggested that AP1 and SEP3 facilitate the opening of closed chromatin and promote gene activation in flower development (Pajoro et al., 2014), indicating roles as pioneer factors. Homeotic TFs are expressed throughout flower development. The analysis of target networks of homeotic TFs allows us to interrogate how homeotic TFs modulate organ growth and cellular morphology, and establish novel cell identities not found in vegetative leaves (Yan et al., 2016; Chen et al., 2018). For example, the homeotic gene AG acts in concert with the general organ polarity gene KANADI1 to suppress trichome initiation in the carpel epidermis (O’Maoléidigh et al., 2018).
et al., 2018). An example for a flower-specific gene activation is SPOROCYTELESS, which is activated by AG in early stages of floral organ development and plays an essential role in patterning processes related to sporogenesis (Schießlhalter et al., 1999; Yang et al., 1999; Ito et al., 2004). The finding that floral organs were derived from leaf-like organs during evolution can also explain the fact that many developmental TFs with roles in cellular patterning appear to act in more than one developmental process and display tissue- and organ-specific functions, since this could be explained by evolutionary co-option and diversification of ancestral regulatory programs. In fact, this complications analyses of cell identity in plants, since only a few genes are entirely characteristic to only one specific cell type. For example, TFs controlling abaxial and adaxial identity were recruited to control patterning in lateral organs, the stem and roots, and their activity can be modulated in a floral organ-specific manner (e.g., Siegfried et al., 1999; Yamaguchi et al., 2004).

The role of epigenetics in plant cell lineage specification

While cell identity can be programmed by cell type-specific TFs, the robustness of the acquired transcriptional status depends on the chromatin environment (Hennig and Derkacheva, 2009; Costa and Dean, 2019). Epigenetic memory can be established during developmental progression because of stable and heritable epigenetic modifications (Huang et al., 2013; Costa and Dean, 2019 and references therein). During cell division, the transcriptional status of genes can be recorded and transmitted to daughter cells via epigenetic regulation (Iwasaki and Paszkowski, 2014). To date, many epigenetic modifications have been found, such as DNA methylation, histone methylation, and histone acetylation. Chromatin modifications can be linked with gene activation or repression or a ‘poised’ state. For example, H3K4me3 and H3K36me3 are associated with gene activation, while H3K27me3 and H3K9me2 are commonly linked to transcriptional repression (Pikaard and Mittelsten Scheid, 2014).

How epigenetic modifications store and transmit ‘memory’ to daughter cells has been intensely studied. Inheritance of histone marks by daughter cells requires the collaboration between the DNA replication machinery, chromatin modifiers, and chromatin modifications themselves (Stewart-Morgan et al., 2020). Polycomb factors are known factors controlling epigenetic memory across cell division that mediate trimethylation of histone 3 Lys27 (H3K27me3). During the DNA replication at mitosis, parental nucleosomes with H3K27 tri-methylation recruit polycomb repressive complex 2 (PRC2) which catalyzes the trimethylation on daughter strand DNA (Jiang and Berger, 2017). This mediates the stability of the repressed status at many developmental gene loci, across developmental stages, such as in the case of FLOWERING LOCUS C (FLC) (reviewed in Costa and Dean, 2019), and in a tissue-specific manner, such as in the case of FLOWERING LOCUS T (FT) (Farrona et al., 2011). In more general terms, the phenotypes of Polycomb mutants strongly suggest broad roles in the mediation of developmental phase transitions and the commitment to cellular differentiation (Mozgova et al., 2015).

Epigenetic marks at specific genomic loci can be dynamically regulated, for example by TFs that interact with epigenetic factors. Marks can be erased, re-written, or diluted by cell division. For example, the B3 domain TFs LEAFY COTYLEDON 2 (LEC2) and FUSCA3 (FUS3) displace VAL1 and VAL2 (two key components for Polycomb-mediated FLC silencing by vernalization) during early embryogenesis from the cold memory cis-element of FLC to disrupt Polycomb silencing and thus prevent H3K27me3 maintenance at FLC during the rapid embryonic cell divisions (Tao et al., 2019). During flower initiation and morphogenesis, TFs such as LFY, MADS-box proteins, and ARF have been shown to modulate chromatin status by recruiting ATP-dependent nucleosome remodelers or general transcriptional co-regulators (Smaczniak et al., 2012; Wu et al., 2012, 2015). In sum, the current data suggest that epigenetic programming plays a role in cell lineage commitment in plants. The investigation of tissue- and stage-specific dynamics of epigenetic profiles can be expected to shed more light on the underlying molecular mechanisms.

Synergistic action of position- and lineage-based cell fate control

The spatiotemporal expression pattern of floral homeotic TFs, and thereby the whorled organization of the flower, is facilitated by multiple factors, including epigenetic factors, positional signals, and regulatory feedback control (Alvarez-Buylla et al., 2010; Deny et al., 2017; Thomson and Wellmer, 2019). An example is provided by AG activity that is restricted to the inner whorls of the floral meristem giving rise to stamens and carpels. Besides being a PcG (Polycomb Group) target, AG expression is prevented in the outer floral whorls via the activity of histone deacetylases (Tian and Chen, 2001; Chen and Tian, 2007). Moreover, miR172 acts as a positional signal to restrict AG activity by regulating the spatiotemporal activity of AP2, which is a known repressor of AG (Aukerman and Sakai, 2003; Chen, 2004; Zhao et al., 2007; Wollmann et al., 2010). Activation of AG in the inner whorls is mediated by the combined activity of several factors including WUS, LFY, and PERIANTHIA (Lenhard et al., 2001; Lohmann et al., 2001; Maier et al., 2009). An autoregulatory feedback loop, possibly involving the interaction with SEP factors, contributes to the stable AG activity (Gomez-Mena et al., 2005; Kaufmann et al., 2009). This and other examples show that spatiotemporal gene expression determining cell identity requires combinatorial interplay of several factors (Fig. 2), and emphasizes the need for novel technological and computational approaches to understand the underlying cis-regulatory grammar.
Excellent examples for crosstalk of epigenetics and phytohormones have been described in controlling floral meristem specification. In general, some phytohormones, such as auxin, gibberellic acid, and brassinosteroids, have been shown to affect epigenetic modifications (Yamamuro et al., 2016). In the absence of auxin, Aux/IAA proteins repress ARF5 activity in the SAM by interacting with ARF5 and recruiting the transcriptional co-repressor TOPLESS (TPL). TPL in turn interacts with histone deacetylase HDA19, thus removing histone acetylation at ARF5 target loci, thereby preventing gene activation (Eberharter and Becker, 2002; Long et al., 2006; Szemenyei et al., 2008). Upon auxin sensing, Aux/IAA proteins are rapidly degraded, leading to the dissociation of TPL and HDA19, thereby freeing ARF5 to activate its targets (Wu et al., 2015; Lavy and Estelle, 2016). Furthermore, in the presence of auxin, ARF5 recruits ATP-dependent SWI/SNF remodeling complexes to its targets, including LFY and FILAMENTOUS FLOWER. This enhances chromatin accessibility at these loci and activates transcription linked with increased H3K9ac (Wu et al., 2015).

Stochasticity in cell fate determination

It is tempting to consider cell type specification as a fully determined process because of the highly reproducible tissue growth and organogenesis. However, the cellular and molecular behaviors underlying cell type specification are often stochastic. Scientists started to realize that stochasticity is needed to create small differences between identical cells, which are then amplified and stabilized by feedback loops to begin cell differentiation (Meyer and Roeder, 2014). Experimental confirmation of this theory is the study of the variable defects in the LEC2 mutant embryo, where FUS3 expression appears in randomly positioned patches. The explanation might be that residual ABSCISIC ACID INSSENSITIVE3 (ABI3) expression fails to induce the expression of FUS3 in some parts of the embryo while it succeeds in triggering FUS3 expression in other parts of the embryo, and the positive feedback loop can stabilize the expression of these two genes in these embryo parts (To et al., 2006).

Stochasticity happens at both cellular and molecular levels. For instance, during the growth of microtubule arrays, stochastic disassembly of individual microtubules allows them to go through various configurations and form optimal ones (Holy and Leibler, 1994; Allard et al., 2010; Eren et al., 2010). Studies have shown that the growth rates of leaf epidermal cells in Arabidopsis differ by several fold from each other, and change in time. This spatiotemporal variability is not related to the size of either the cell or the nucleus (Elsner et al., 2012).

Stochastic gene expression has also been described in various organisms, including plants (Elowitz et al., 2002; Paré et al., 2009; Dar et al., 2012; Ietswaart et al., 2017). Gene expression noise can be divided into two types: extrinsic noise which is due to fluctuations in the cellular or external environment that affect the overall expression in a cell, and intrinsic noise which...
is due to the inherent fluctuations of transcription and translation of a particular gene within a cell (Elowitz et al., 2002). The use of a dual reporter system in plants helped distinguish between extrinsic and intrinsic noise in Arabidopsis, and revealed that fluctuation in gene expression is coupled in neighboring cells in young leaves (Araújo et al., 2017). The trichome distribution pattern also emerges stochastically, and the variability in the trichome distribution pattern correlated with stochastic cell to cell variation in GL3 expression (Okamoto et al., 2020).

One of the most compelling examples of stochasticity is from cell type specification in the sepal epidermis. Sepals have both giant cells that are very long, usually stretching one-fifth the length of the sepal, and small cells that are much smaller in size (Figure 1; Roeder et al., 2010, 2012). The correct proportion of giant cells and small cells is required for the curvature of the sepal; with an altered proportion of giant cells, sepals are unable to enclose and protect the developing floral organs in the inner whorls (Roeder et al., 2010, 2012). In the early stage of sepal development, the levels of the epidermis regulator ARABIDOPSIS THALLANA MERISTEM LAYER1 (ATML1) fluctuate in sepal cells. When ATML1 reaches a high level, specifically at the time of cell division, that cell will be determined to become a giant cell, whereas if the level of ATML1 is low at this point in time, the cell will keep dividing and remain small. Thus, the stochastic fluctuations in the concentration of the TF ATML1 initiate the pattern of giant and small cells in the Arabidopsis sepal (Meyer et al., 2017). The examples presented above indicate that plants utilize stochastic mechanisms to establish robust and reproducible morphology.

How does single-cell omics contribute to understanding cell identity?

Single-cell omics technologies

Despite the limitations of the cell type concept, classifying cells can help to understand how cells or tissues function and interact, and to reveal specific mechanisms that govern processes that may influence a plant’s growth, development, and reproduction. Recent advances in profiling molecular features at single-cell resolution provide novel insights into the understanding of cell types (see, for example, Özel et al., 2021). Benefiting from the development of single-cell omics technologies, researchers can now study cellular heterogeneity at the levels of the transcriptome, epigenome, or proteome. Single-cell RNA-seq (scRNA-seq) (Tang et al., 2009), ATAC-seq (Buenrostro et al., 2015), ChIP-seq (Rotem et al., 2015), DNA methylation (Hui et al., 2018; Lee and Smallwood, 2018), metabolomics (Minakshi et al., 2019), and proteomics (Marx, 2019), have emerged and are used in animal and plant research, but scRNA-seq is still the most commonly used technique. For example, scRNA-seq permits analysis of the expression profiles of thousands of individual cells at the same time and can reveal the heterogeneity within a group of cells.

An scRNA-seq experimental workflow usually begins with the dissociation and isolation of single cells from a tissue. However, in plants, the process of isolating single cells embedded in a rigid cell wall matrix is technically challenging, and it is usually achieved by incubating plant tissues with cell wall-digesting enzymes to release protoplasts. Protoplast response genes can cause artifacts in the downstream data analysis (Tucker et al., 2018; Denyer et al., 2019; Jean-Baptiste et al., 2019; Ryu et al., 2019; Shulse et al., 2019), and the time and harshness of the digestion that are required to digest the cell walls differ between tissues and organs. Thus, this method is not applicable to all tissues, and longer digestion times may aggregate artifacts. An alternative way to address this issue is to isolate nuclei, for example by tissue chopping or grinding and by cell membrane lysis (e.g. Thibivilliers et al., 2020 Deal and Henikoff, 2010; Kaufmann et al., 2010a). It has been shown that the composition of the RNA pool from plant nuclei is representative of that from the whole cell (Deal and Henikoff, 2010). However, the isolation procedure and the loss of mechanical connection with other cell components, particularly the cytoskeleton, may change the shape of the nucleus and impact gene expression (Goswami et al., 2020a). Additionally, dealing with the sparse RNA from nuclei is challenging for both the experimental and the computational parts of the work. Further optimizing tissue dissociation methods, especially for recalcitrant plant tissues is fundamentally critical to apply scRNA-seq to plant science.

What we learned from scRNA-seq is that no two cells are transcriptionally the same (Choi and Kim, 2019). Nevertheless, clustering of cells with similar expression or epigenetic profiles is often used to annotate cell types. Subclusters can reflect the variation of expression patterns among cells of the same tissue type, and may represent cell types. Taking data from roots as an example, in the stele cell cluster, there are protoxylem, phloem-like, meristematic xylem, and pericycle cells (Shulse et al., 2019). However, heterogeneity may also reflect differences in cellular states, or stochastic fluctuations in gene activity (Trapnell, 2015; Wimmers et al., 2018). Cells of different ontogenetic origins may have similar functions or ‘behaviors’ in terms of gene activity. In scRNA-seq clustering, cells are grouped based on the similarity of their transcriptome. For example, the lateral root cap (LRC) cells were found to cluster with the non-hair cells and columella cells (Shulse et al., 2019), which indicates that although they are different types of cells that originate from different initial cells surrounding the quiescent center (QC), they share a similar transcriptome that may provide them with the ability to protect the roots (Petricka et al., 2012). It has also been shown that meristematic cells cluster together independently of precise origin. The meristematic cell clusters are close to each other and consist of meristematic cells of different identities, such as cortex identity and trichoblast identity (Denyer et al., 2019). The reason for this may be because these cells share meristematic features such as a high division rate, although they have different ultimate cell fates.
The annotation of clusters in single-cell datasets other than roots—or in species other than Arabidopsis—is typically limited by the availability of tissue-specific reference datasets and specific marker genes. Such pre-knowledge of tissue-specific data can strongly enhance our capacity to annotate cells in single-cell omics datasets. This is also the case for the data specific data can strongly enhance our capacity to annotate cells and specific marker genes. Such pre-knowledge of tissue-dermis) or cells of a certain status of differentiation (e.g. floral stem cells).

Different computational approaches try to order the transcriptome of the cells obtained by single-cell omics in some type of differentiation trajectory. The earlier methods were based on ordering cells in a pseudotime defined by similarity, for example as implemented in Monocle or Palantir (Trapnell et al., 2014; Setty et al., 2019). In this way, cells with similar transcriptomes were ordered together in a computationally generated pseudotime. The main problem with these methods is the assumption that transcriptome similarity is related to a similar position in the differentiation pathway, because, as we stated before, cells even from different origins can have similar transcriptomes. New approaches to infer lineage decisions are based on estimating the dynamic ratios of spliced and unspliced transcripts, for example as utilized in velo or sevelo (La Manno et al., 2018; Bergen et al., 2020). We can infer reaction rates of transcription, splicing, and degradation by modeling the abundance of spliced and unspliced transcripts, therefore providing an estimation of the latent time behind these dynamics.

Towards a virtual flower: understanding cell identity in its positional context

Single-cell omics procedures are associated with the loss of positional information of plant cells. However, as discussed in the previous sections, positional information is vital for morphogenesis and cell identity in plants. By combining high-resolution imaging of marker gene activity with single-cell omics, the position of cells in their original tissue context can be predicted (Satija et al., 2015; Halpern et al., 2017; Cang and Nieuwenhuizen, 2020). It is also possible to map scATAC-seq data to spatial maps of gene activity (Bravo González-Blas et al., 2020). A computational framework called novoSpaRc was developed to this aim, which, in theory, can be used to de novo reconstruct single-cell spatial gene expression without prior spatial information, although the use of prior spatial information enhances its performance (Nitzan et al., 2019). Attempts to map expression of selected regulatory genes to a virtual 3D floral meristem based on reporter gene expression and in situ hybridization provide a resource for this kind of computational technologies (Refahi et al., 2021), and could be expanded to comprehensively cover all tissue types in the developing flower.

Although these computational tools can regain the positional information of dissociated cells, the dissociation procedure itself may cause plant cells to alter their identity, as already discussed above. So, many efforts have been made to retain tissue spatial context by using fluorescence in situ hybridization (FISH)-based methods, such as multiplexed error-robust FISH (MERFISH), spatially resolved transcript amplicon readout mapping (STARmap), and sequential fluorescence in situ hybridization (seqFISH+) (Shah et al., 2016; Wang et al., 2018; Peng et al., 2019). SeqFISH+ can image mRNAs from up to 10 000 genes in single cells with high resolution, allowing identification of cell types based on both transcriptional profile and their spatial organization in situ (Peng et al., 2019). Besides, seqFISH+ can also reveal subcellular mRNA localization in single cells. However, the drawback is that these kinds of methods only allow targeted studies and lack unbiased examination of the whole transcriptome. A recently published technology, expansion sequencing (ExSeq), combined expansion microscopy with long-read in situ RNA sequencing, resulting in a nanoscale visualization of the position of transcripts in intact tissues (Alon et al., 2021). ExSeq does not need target genes, so it is unbiased compared with other in situ sequencing methods, as mentioned above (Alon et al., 2021).

Another experimental approach to retrieve cell positional information is spatial RNA-seq. Researchers have generated high-quality RNA-seq data with maintained two-dimensional positional information by lysing histological sections on arrayed reverse transcription primers with unique positional barcodes (Stahl et al., 2016). A similar method, Slide-seq, transfers RNA from tissue sections onto DNA-barcoded drop-seq beads arrayed on a surface with known positions, allowing whole-genome sequencing of RNA with inferred locations (Rodrigues et al., 2019). However, these technologies can only capture tissues in a thin section, and each bead is not strictly capturing RNA from a single isolated cell. Combined with scRNA-seq, these approaches may help to map or assign single-cell transcriptomics data back into a tissue context, overcoming the need for targeted spatial expression analyses of marker genes.

In sum, parallel imaging of the expression of multiple regulatory genes or spatial omics approaches present promising avenues for mapping the expression and regulatory programs of each individual cell in a developing flower, thus taking into account position and lineage. To trace plant cell lineages, it would be interesting to test the applicability of CRISPR/Cas9 [clustered regularly interspaced palindromic repeats / CRISPR-associated protein 9]-based lineage tracing in plants (Spanjaard et al., 2018).
The promises of single-cell omics

Single-cell level transcriptomics can define cell types and identify marker genes (Trapnell, 2015). Although the clustering of cells based on ‘similarity’ has limitations (explained in the next section) in cell type identification, the techniques provide us with new insights into cellular heterogeneity: (i) single-cell technology enables the discovery of novel cell types; (ii) detects subtypes or cell states in a single cell type; and (iii) orders cells in ‘time’ along a trajectory makes it possible to infer, or at least predict, differentiation pathways (Shekhar and Menon, 2019; Torii et al., 2019; Rich-Griffin et al., 2020).

Single-cell omics can also provide new ways to study the function of positional signals in cell identity determination or tissue patterning. For example, a combination of single-cell omics and genetic perturbation would allow us to decipher cell type-specific auxin response pathways and dosage-dependent mechanisms in the flower. Another interesting application of scRNA-seq is to detect gradients of regulatory molecules within tissues. For example, scRNA-seq with roots has shown that some genes, such as SCARECROW and UPBEAT, represent concentration gradients along the clusters (Ryu et al., 2019). Underlying mechanisms, such as mobility or gradients of regulation, will require combination of different experimental set-ups.

Since every cell of a plant is exposed to an environment that is inherently heterogeneous, single-cell omics may also help us understand cell type-specific environmental responses in the future.

Current limitations of single-cell omics in exploring cell identities

Although single-cell omics techniques are gaining more and more popularity, we have to be aware of their limitations. For example, the power of scRNA-seq is limited because of its inability to sensitively capture all transcripts, leaving false-negative ‘zeros’ in gene expression (Dal Molin and Di Camillo, 2019). In addition, clustering methods are typically based on the assumption that cells with similar transcriptional features are ontogenetically closely related. However, the actual relationships among the profiled cells are not known because expression only represents one layer of cellular regulation, and transcriptomic similarity may not always reflect ontogenetic origin. Integrated profiling with other molecular features and spatial reconstruction will overcome these limitations (see e.g. Macaulay et al., 2015; Angermueller et al., 2016; Cao et al., 2018).

One of the uses of scRNA-seq is to define subtypes or states within a cell type. However, fluctuations in gene expression can be caused by oscillatory cell behavior, linked to cell division, apoptosis, the circadian clock, and stochastic or bursty transcription (Stegle et al., 2015; Dal Molin and Di Camillo, 2019). Nevertheless, it is possible to correct the expression noise with computational methods if these cell behaviors are not of interest (Leng et al., 2015; Barron and Li, 2016).

Conclusion and outlook

The difficulty of defining cell type conceptionally reflects the complex and dynamic nature of cells in plants. Researchers over the past decades have proposed genetic mechanisms and models to elucidate how plants build up their bodies with diverse cell types. Both the position and ‘history’ of a plant cell are important for its identity. Positional signals such as auxin, small peptides, miRNAs, and mobile TFs, as well as mechanical forces, have been shown to contribute to cell type specification and patterning in plants. TFs, including pioneer TFs associated with epigenetic modifications that evoke or consolidate cellular ‘history’, reflect the role of cell lineage in plant cell fate determination. It is the synergistic action of position- and lineage-based cell fate controlling factors, together with unregulated factors (stochasticity) that eventually determines cell identity.

Single-cell technology brings new insight into the understanding of cell identity by its advantage of studying gene expression, chromatin status, and other cellular features at the single-cell level. Despite the room for improvement, it allows us to dissect cellular heterogeneity by defining novel cell types or states that may have been neglected by classical studies. Single-cell technology can also order cells along a trajectory and make it possible to infer the origins and consequences of differentiation.

The combination of genetic analyses with single-cell technologies, reporter gene analyses, and spatial omics can be expected to deepen our knowledge on mechanisms and consequences of cell identity specification and organ patterning in plants in the future.

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Author contributions

XX: drafting and correcting the manuscript; KK, JM, and CS: editing and improving the manuscript, and contributing to specific sections.

References

Allard JF, Wasteneys GO, Cytrynbaum EN. 2010. Mechanisms of self-organization of cortical microtubules in plants revealed by computational simulations. Molecular Biology of the Cell 21, 278–286.

Alon S, Goodwin DR, Sinha A, et al. 2021. Expansion sequencing: spatially precise in situ transcriptomics in intact biological systems. Science 371, eaae2656.
Alvarez-Buylla ER, Benitez M, Corvera-Poire A, et al. 2010. Flower development. The Arabidopsis Book 8, e0127.

Angermueller C, Clark SJ, Lee HJ, et al. 2016. Parallel single-cell sequencing links transcriptional and epigenetic heterogeneity. Nature Methods 13, 229–232.

Araújo IS, Pietsch JM, Keizer EM, et al. 2017. Stochastic gene expression in Arabidopsis thaliana. Nature Communications 8, 2132.

Arber A. 1950. The natural philosophy of plant form. Cambridge: Cambridge University Press.

Aukerman MJ, Sakai H. 2003. Regulation of flowering time and floral organ identity by a MicroRNA and its APETAL2-like target genes. The Plant Cell 15, 2730–2741.

Balanzà V, Balleser F, Colombo M, Fourquin C, Martínez-Fernández I, Ferrándiz C. 2014. Genetic and phenotypic analyses of carpel development in Arabidopsis. Methods in Molecular Biology 1110, 231–249.

Barron M, Li J. 2016. Identifying and removing the cell-cycle effect from single-cell RNA-Sequencing data. Scientific Reports 6, 33892.

Benková E, Michniewicz M, Sauer M, et al. 2003. Local, efflux-dependent auxin gradients as a common module for plant organ formation. Cell 115, 591–602.

Bergen V, Lange M, Peidli S, Wolf FA, Theis FJ. 2020. Generalizing RNA velocity to transient cell states through dynamical modeling. Nature Biotechnology 38, 1408–1414.

Berger F, Haseloff J, Schiefelbein J, Dolan L. 1998. Positional information in root epidermis is defined during embryogenesis and acts in domains with strict boundaries. Current Biology 8, 421–430.

Bowman JL, Smyth DR, Meyerowitz EM. 1991. Genetic interactions among floral homeotic genes of Arabidopsis. Development 112, 1–20.

Brady SM, Orlando DA, Lee JY, et al. 2007. A high-resolution root spatiotemporal map reveals dominant expression patterns. Science 318, 801–806.

Bravo González-Blas C, Quan XJ, Durán-Romáñ R, et al. 2020. Identification of genomic enhancers through spatial integration of single-cell transcriptomics and epigenomics. Molecular Systems Biology 16, e9438.

 Bueno Rojo JD, Wu B, Litzenburger UM, et al. 2015. Single-cell chromatin accessibility reveals principles of regulatory variation. Nature 523, 486–490.

Cang Z, Nie Q. 2020. Inferring spatial and signaling relationships between cells from single cell transcriptomic data. Nature Communications 11, 2084.

Cao J, Cusanovich DA, Ramani V, et al. 2020. Architecture of gene regulation in Arabidopsis. Scientific Reports 6, 33892.

Costa S, Shaw P. 2006. Chromatin organization and cell fate switch respond to positional information in Arabidopsis. Nature 439, 493–496.

Dal Molin A, Di Camillo B. 2019. How to design a single-cell RNA-sequencing experiment: pitfalls, challenges and perspectives. Briefings in Bioinformatics 20, 1384–1394.

Dar RD, Razooky BS, Singh A, et al. 2012. Transcriptional burst frequency and burst size are equally modulated across the human genome. Proceedings of the National Academy of Sciences, USA 109, 17454–17459.

D’Ario M, Griffiths-Jones S, Kim M. 2017. Small RNAs: big impact on plant development. Trends in Plant Science 22, 1056–1068.

Dawe RK, Freeing M. 1991. Cell lineage and its consequences in higher plants. The Plant Journal 1, 3–8.

Deal RB, Henikoff S. 2010. A simple method for gene expression and chromatin profiling of individual cell types within a tissue. Development 13, 1030–1040.

Denay G, Chahtane H, Tichtinsky G, Parcy F. 2017. A flower is born: an update on Arabidopsis floral meristem formation. Current Opinion in Plant Biology 35, 15–22.

Denyer T, Ma X, Klesen S, Scacchi E, Nieselt K, Timmermans MCP. 2019. Spatiotemporal developmental trajectories in the arabidopsis root revealed using high-throughput single-cell RNA sequencing. Developmental Cell 48, 840–852.e5.

Eberharder A, Becker PB. 2002. Histone acetylation: a switch between repressive and permissive chromatin. EMBO Reports 3, 224–229.

Eloffitz MB, Levine AJ, Siggia ED, Swain PS. 2002. Stochastic gene expression in a single cell. Science 297, 1183–1186.

Elsner J, Michalski M, Kwiatkowska D. 2012. Spatiotemporal variation of leaf epidermal cell growth: a quantitative analysis of Arabidopsis thaliana wild-type and triple cycld3 mutant plants. Annals of Botany 109, 897–910.

Eren EC, Dixit R, Gautham N. 2010. A three-dimensional computer simulation model reveals the mechanisms for self-organization of plant cortical microtubules into oblique arrays. Molecular Biology of the Cell 21, 2674–2684.

Fal K, Landrein B, Hamant O. 2016. Interplay between miRNA regulation and mechanical stress for CUC gene expression at the shoot apical meristem. Plant Signaling & Behavior 11, e127497.

Farrarona S, Thorpe FL, Engelnhor J, et al. 2011. Tissue-specific expression of FLOWERING LOCUS T in Arabidopsis is maintained independently of polycistron group protein repression. The Plant Cell 23, 3204–3214.

Finnan JD, Guilak F. 2010. The effects of osmotic stress on the structure and function of the cell nucleus. Journal of Cellular Biochemistry 109, 460–467.

Fletcher JC, Brand U, Running MP, Simon R, Meyerowitz EM. 1999. Signaling of cell fate decisions by CLAVATA3 in Arabidopsis shoot meristems. Science 283, 1911–1914.

Gallos JL, Nora FR, Mizukami Y, Sablowski R. 2004. WUSCHEL induces shoot stem cell activity and developmental plasticity in the root meristem. Genes & Development 18, 375–380.

Glover B. 2000. Differentiation in plant epidermal cells. Journal of Experimental Botany 51, 497–505.

Glover BJ, Martin C. 1998. The role of petal cell shape and pigmentation in pollination success in Antirrhinum majus. Heredity 80, 778–784.

Gomez-Mena C, de Folter S, Costa MM, Angenent GC, Sablowski R. 2005. Transcriptional program controlled by the floral homeotic gene AGAMOUS during early organogenesis. Development 132, 429–438.
Iwafuchi-Doi M, Zaret KS. 2014. Pioneer transcription factors in cell reprogramming. Genes & Development 28, 2679–2692.

Iwafuchi-Doi M, Zaret KS. 2016. Cell fate control by pioneer transcription factors. Development 143, 1833–1837.

Iwasaki M, Paszkowski J. 2014. Epigenetic memory in plants. The EMBO Journal 33, 1987–1998.

Jacques E, VerbeLEN JP, Vissenberg K. 2013. Mechanical stress in Arabidopsis leaves orients microtubules in a ‘continuous’ supracellular pattern. BMC Plant Biology 13, 163.

Jean-Baptiste K, McFaline-Figueroa JL, Alexandre CM, et al. 2019. Dynamics of gene expression in single root cells of Arabidopsis thaliana. The Plant Cell 31, 993–1011.

Jiang D, Berger F. 2017. DNA replication-coupled histone modification maintains Polycomb gene silencing in plants. Science 357, 1146–1149.

Jiao Y, Meyerowitz EM. 2010. Cell-type specific analysis of translating RNAs in developing flowers reveals new levels of control. Molecular Systems Biology 6, 419.

Jin R, Klasfeld S, Zhu Y, et al. 2021. LEAFY is a pioneer transcription factor and licenses cell reprogramming to floral fate. Nature Communications 12, 626.

Kaufmann K, Airoldi CA. 2018. Master regulatory transcription factors in plant development: a blooming perspective. Methods in Molecular Biology 1830, 3–22.

Kaufmann K, Muño JM, Jauregui R, et al. 2009. Target genes of the MADS transcription factor SEPALLATA3: integration of development and hormonal pathways in the Arabidopsis flower. PLoS Biology 7, e1000390.

Kaufmann K, Muño JM, Osteràs M, Farinelli L, Krajewski P, Angenent GC. 2010a. Chromatin immunoprecipitation (ChIP) of plant transcription factors followed by sequencing (ChIP-SEQ) or hybridization to whole genome arrays (ChIP-CHIP). Nature Protocols 5, 457–472.

Kaufmann K, Wellmer F, Muño JM, et al. 2010b. Orchestration of floral initiation by APETALA1. Science 328, 85–89.

Kidner C, Sundaresan V, Roberts K, Dolan L. 2000. Clonal analysis of the Arabidopsis root confirms that position, not lineage, determines cell fate. Planta 211, 191–199.

Kinosita A, Betsuyaku S, Osakabe Y, et al. 2010. RPK2 is an essential receptor-like kinase that transmits the CLV3 signal in Arabidopsis. Development 137, 3911–3920.

Kondo T, Sawa S, Kinosita A, et al. 2006. A plant peptide encoded by CLV3 identified by in situ MALDI-TOF MS analysis. Science 313, 845–848.

Kretzschmar K, Watt FM. 2010. Lineage between animals and plants—colonizing chromatin for gene regulation. Molecules 15, 3226–3229.

Lai X, Mathieu R, GrandVuillemin L, et al. 2021. The LEAFY floral regulator displays pioneer transcription factor properties. Molecular Plant 5, 829–837.

Lai X, Verhage L, Hougouviex V, Zubieta C. 2018. Pioneer factors in animals and plants—colonizing chromatin for gene regulation. Molecules 23, 1914.

La Manno G, Soldatov R, Zeisel A, et al. 2018. RNA velocity of single cells. Nature 560, 494–498.

Landrein B, Kiss A, Sassi M, et al. 2015. Mechanical stress contributes to the expression of the STM homeobox gene in Arabidopsis shoot meristems. eLife 4, e07811.

Larkin JC, Marks MD, Nadeau J, Sack F. 1997. Epidermal cell fate and patterning in leaves. The Plant Cell 9, 1109–1120.

Lavy M, Estelle M. 2016. Mechanisms of auxin signaling. Development 143, 3226–3229.

Lee HJ, Smallwood SA. 2018. Genome-wide analysis of DNA methylation in single cells using a post-bisulfite adapter tagging approach. In: Head SR, Ordoukhian P, Salomon DR, eds. Next generation sequencing: methods and protocols. New York: Springer New York, 87–95.
Lee JS, Hnilova M, Maes M, et al. 2015. Competitive binding of antagonistic peptides fine-tunes stomatal patterning. Nature 522, 439–443.

Leng N, Chu LF, Barry C, et al. 2015. Oscpe identifies oscillatory genes in unsynchronized single-cell RNA-seq experiments. Nature Methods 12, 947–950.

Lenhard M, Bohnert A, Jürgens G, Laux T. 2001. Termination of stem cell maintenance in Arabidopsis floral meristems by interactions between WUSCHEL and AGAMOUS. Cell 105, 805–814.

Leysser O. 2018. Auxin signaling. Plant Physiology 176, 465–479.

Lin G, Zhang L, Han Z, et al. 2017. A receptor-like protein acts as a specificity switch for the regulation of stomatal development. Genes & Development 31, 927–938.

Lohmann JU, Hong RL, Hobe M, et al. 2001. A molecular link between stem cell regulation and floral patterning in Arabidopsis. Cell 105, 793–803.

Long JA, Ohno C, Smith ZR, Meyerowitz EM. 2006. TOPLESS regulates apical embryonic fate in Arabidopsis. Science 312, 1520–1523.

Lovett DB, Shekhar N, Nickerson JA, Roux KJ, Lele TP. 2013. Modulation of nuclear shape by substrate rigidity. Cellular and Molecular Bioengineering 6, 230–238.

MacAllister CA, Ohashi-Ito K, Bergmann DC. 2007. Transcription factor asymmetry controls asymmetric division cells that establish the stomatal lineage. Nature 445, 537–540.

Macaulay IC, Haerty W, Kumar P, et al. 2015. G&T-seq: parallel sequencing of single-cell genomes and transcriptomes. Nature Methods 12, 519–522.

Maier AT, Stehling-Sun S, Wollmann H, et al. 2009. Dual roles of the bZIP transcription factor PERIANTHA in the control of floral architecture and homeotic gene expression. Development 136, 1613–1620.

Marhava P, Hoermayer L, Yoshida S, Marhavý P, Benková E, Friml J. 2019. Re-activation of stem cell pathways for pattern restoration in plant wound healing. Cell 177, 957–969.e13.

Marx V. 2019. A dream of single-cell proteomics. Nature Methods 16, 809–812.

Mayer KF, Schoof H, Haecker A, Lenhard M, Jürgens G, Laux T. 1998. Role of WUSCHEL in regulating stem cell fate in the Arabidopsis shoot meristem. Cell 95, 805–815.

Mazzarello P. 1999. A unifying concept: the history of cell theory. Nature Cell Biology 1, E13–E15.

Meier I, Griffiths AHN, Groves NR, Wagner A. 2016. Regulation of nuclear shape and size in plants. Current Opinion in Cell Biology 40, 114–123.

Meyer HM, Roeder AH. 2014. Stochasticity in plant cellular growth and patterning. Frontiers in Plant Science 5, 420.

Meyer HM, Teles J, Formosa-Jordan P, et al. 2017. Fluctuations of the transcription factor ATML1 generate the pattern of giant cells in the Arabidopsis sepal. eLife 6, e19131.

Minakshi P, Ghosh M, Kumar R, et al. 2019. Single-cell metabolomics: technology and applications. In: Barth D, Azevedo V, eds. Single-cell omics. London: Academic Press, 319–353.

Morris SA. 2019. The evolving concept of cell identity in the single cell era. Development 146, dev169748.

Mozgova I, Köhler C, Hennig L. 2015. Keeping the gate closed: functions of the polycomb repressive complex PRC2 in development. The Plant Journal 83, 121–132.

Murphy E, Smith S, De Smet I. 2012. Small signaling peptides in Arabidopsis development: how cells communicate over a short distance. The Plant Cell 24, 3198–3217.

Nadeau JA, Sack FD. 2002. Stomatal development in Arabidopsis. The Arabidopsis Book 1, e0066.

Nagpal P, Ellis CM, Weber H, et al. 2005. Auxin response factors ARF6 and ARF8 promote jasmonic acid production and flower maturation. Development 132, 4107–4118.

Nitzan M, Karaiskos N, Friedman N, Rajewsky N. 2019. Gene expression cartography. Nature 576, 132–137.

Ohashi-Ito K, Bergmann DC. 2006. Arabidopsis FAMA controls the final proliferation/differentiation switch during stomatal development. The Plant Cell 18, 2493–2505.

Okada K, Ueda J, Komaki MK, Bell CJ, Shimura Y. 1991. Requirement of the auxin polar transport system in early stages of Arabidopsis floral bud formation. The Plant Cell 3, 677–684.

Okamoto S, Negishi K, Toyama Y, Ushijima T, Morohashi K. 2020. Leaf trichome distribution pattern in arabidopsis reveals gene expression variation associated with environmental adaptation. Plants 9, 908.

Ó’Maoléidigh DS, Graciet E, Wellmer F. 2014. Gene networks controlling Arabidopsis thaliana flower development. New Phytologist 201, 16–30.

Ó’Maoléidigh DS, Stewart D, Zheng B, Coupland G, Wellmer F. 2018. Floral homosolc proteins modulate the genetic program for leaf development to suppress trichome formation in flowers. Development 145, dev157784.

Özel MN, Simon F, Jafari S, et al. 2021. Neuronal diversity and convergence in a visual system developmental atlas. Nature 589, 88–95.

Pajoro A, Madrigal P, Muino JM, et al. 2014. Dynamics of chromatin accessibility and gene regulation by MADS-domain transcription factors in flower development. Genome Biol 15, R41.

Parcy F, Nilsson O, Busch MA, Lee I, Weigel D. 1998. A genetic framework for floral patterning. Nature 395, 561–566.

Paré A, Lemons D, Kosman D, Weaver B, Freund Y, McGinnis W. 2009. Visualization of individual Sc mRNAs during Drosophila embryogenesis yields evidence for transcriptional bursting. Current Biology 19, 2037–2042.

Parry G, Calderon-Villalobos LI, Prigge M, et al. 2009. Complex regulation of the TIR1/AFB family of auxin receptors. Proceedings of the National Academy of Sciences, USA 106, 22540–22545.

Pattanaik S, Patra B, Simcox SK, Youn J. 2014. An overview of the gene regulatory network controlling trichome development in the model plant, Arabidopsis. Frontiers in Plant Science 5, 259.

Pelaz S, Ditta GS, Baumann E, Wisman E, Yanofsky MF. 2000. B and C floral organ identity functions require SEPALLATA MADS-box genes. Nature 405, 200–203.

Pelaz S, Tapia-López R, Alvarez-Buylla ER, Yanofsky MF. 2001. Conversion of leaves into petals in Arabidopsis. Current Biology 11, 182–184.

Peng PC, Khoeirei P, Girardot C, et al. 2019. The role of chromatin accessibility in cis-regulatory evolution. Genome Biology and Evolution 11, 1813–1828.

Petricka JJ, Winter CM, Benfey PN. 2012. Control of Arabidopsis root development. Annual Review of Plant Biology 63, 563–590.

Pikaard CS, Mittelsten Scheid O. 2014. Epigenetic regulation in plants. Cold Spring Harbor Perspectives in Biology 6, a019315.

Pillitteri LJ, Boogenschutz NL, Torii KU. 2008. The bHLH protein, MUTE, controls differentiation of stomata and the hydathode pore in Arabidopsis. Plant & Cell Physiology 49, 934–943.

Poethig RS. 1987. Cional analysis of cell lineage patterns in plant development. American Journal of Botany 74, 581.

Poethig S. 1989. Genetic mosaics and cell lineage analysis in plants. Trends in Genetics 5, 273–277.

Przemek GK, Mattsson J, Hardtke CS, Sung ZR, Berleth T. 1996. Studies on the role of the Arabidopsis gene MONOPTEROS in vascular development and plant cell axialization. Planta 200, 229–237.

Rast MI, Simon R. 2008. The meristem-to-organ boundary: more than an extremity of anything, Curr Opin Genet Dev 18, 287–294.

Refahi Y, Zardilis A, Michelin G, et al. 2021. A multiscale analysis of early flower development in Arabidopsis provides an integrated view of molecular regulation and growth control. Developmental Cell 56, P540–P556.

Reinhardt D, Mandel T, Kuhlemeier C. 2000. Auxin regulates the initiation and radial position of plant lateral organs. The Plant Cell 12, 507–518.

Rich-Griffin G, Stechemesser A, Finch J, Lucas E, Ott S, Schäfer P. 2020. Single-cell transcriptomics: a high-resolution avenue for plant functional genomics. Trends in Plant Science 25, 186–197.
2016. What is the problem? Development, 342–357.

Shah S, Lubeck E, Zhou W, Cai L. 2016. In situ transcription profiling of single cells reveals spatial organization of cells in the mouse hippocampus. Nature Biotechnology, 451–460.

Characterization of cell fate probabilities in single-cell data with Palantir. Setty M, Kiseliovas V, Levine J, Gayoso A, Mazutis L, Pe’er D. 2016. Reproductive organs. Development, 124, 689–698.

Schleiden MJ. 1838. Beiträge zur phytogenesis. Berlin: Veit et Comp.

Schleiden M. 1992. The origin of pattern and polarity in the Drosophila embryo. Cell, 68, 201–219.

Stähli PL, Salmén F, Vickovic S, et al. 2016. Visualization and analysis of gene expression in tissue sections by spatial transcriptomics. Science, 353, 78–82.

Steeves TA, Sussex IM. 1989. Patterns in plant development. Cambridge: Cambridge University Press.

Stegle O, Teichmann SA, Marioni JC. 2015. Computational and analytical challenges in single-cell transcriptomics. Nature Reviews. Genetics, 16, 133–145.

Stent GS. 1985. The role of cell lineage in development. Philosophical Transactions of the Royal Society of London. Series B, Biological Sciences, 312(1153), 3–19.

Stewart R, Burk L. 1970. Independence of tissues derived from apical layers in ontogeny of the tobacco leaf and ovary. American Journal of Botany, 57, 1010–1016.

Stewart R, Demchen H. 1975. Flexibility in ontogeny as shown by the contribution of the shoot apical layers to leaves of periclinal chimeras. American Journal of Botany, 62, 935–947.

Stewart-Morgan KR, Petryk N, Groth A. 2020. Chorimat replication and epigenetic cell memory. Nature Cell Biology, 22, 361–371.

Szemenyei H, Hannon M, Long JA. 2008. TOPLESS mediates auxin-dependent transcriptional repression during Arabidopsis embryogenesis. Science, 319, 1384–1386.

Szymkowiak EJ, Sussex IM. 1996. What chimeras can tell us about plant development. Annual Review of Plant Physiology and Plant Molecular Biology, 47, 351–376.

Tang F, Barbacioru C, Wang Y, et al. 2009. mRNA-Seq whole-transcriptome analysis of a single cell. Nature Methods, 6, 377–382.

Tao Z, Hu H, Luo X, Jia B, Du J, He Y. 2019. Embryonic resetting of the parental vernalized state by two B3 domain transcription factors in Arabidopsis. Nature Plants, 5, 424–435.

Tao Z, Shen L, Gu X, Wang Y, Yu H, He Y. 2017. Embryonic epigenetic reprogramming by a pioneer transcription factor in plants. Nature, 551, 125–128.

Tauriello G, Meyer HM, Smith RS, Kounoutsakos P, Roeder AH. 2015. Variability and constancy in cellular growth of arabidopsis sepal epidermal cells. Plant Physiology, 169, 2342–2358.

Theissen G. 2001. Development of floral organ identity: stories from the MADS house. Current Opinion in Plant Biology, 4, 75–85.

Theißen G, Saedler H. 2001. Floral quartets. Nature, 409, 469.

Thibivilli S, Anderson D, Libault M. 2020. Isolation of Plant Root Nuclei for Single Cell RNA Sequencing. Current Protocols in Plant Biology, e20120. doi:10.1002/cppb.20120.

Thomson B, Wellmer F. 2019. Molecular regulation of flower development. Current Topics in Developmental Biology, 131, 185–210.

Tian L, Chen ZJ. 2001. Blocking histone deacetylation in Arabidopsis induces pleiotropic effects on plant gene regulation and development. Proceedings of the National Academy of Sciences, USA, 98, 200–205.
Barbeau F, Moreau J, Lalande D, et al. 2014. Establishment of floral meristem competence by PLETHORA-dependent LEAFY expression. Development 141, 1549–1559.

Barbeau F, Moreau J, Lalande D, et al. 2014. Establishment of floral meristem competence by PLETHORA-dependent LEAFY expression. Development 141, 1549–1559.

Barbeau F, Moreau J, Lalande D, et al. 2014. Establishment of floral meristem competence by PLETHORA-dependent LEAFY expression. Development 141, 1549–1559.

Barbeau F, Moreau J, Lalande D, et al. 2014. Establishment of floral meristem competence by PLETHORA-dependent LEAFY expression. Development 141, 1549–1559.

Barbeau F, Moreau J, Lalande D, et al. 2014. Establishment of floral meristem competence by PLETHORA-dependent LEAFY expression. Development 141, 1549–1559.