

1 **Advancing root developmental research through single-cell technologies**

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11 **Abstract**

12 Single-cell RNA-sequencing has greatly increased the spatiotemporal resolution of root
13 transcriptomics data, but we are still only scratching the surface of its full potential. Despite
14 the challenges that remain in the field, the orderly aligned structure of the Arabidopsis root
15 meristem makes it specifically suitable for lineage tracing and trajectory analysis. These
16 methods will become even more potent by increasing resolution and specificity using tissue
17 specific scRNA-seq and spatial transcriptomics. Feeding multiple single-cell omics datasets
18 into single-cell gene regulatory networks will accelerate the discovery of regulators of root
19 development in multiple species. By providing transcriptome atlases for virtually any species,
20 single-cell technologies could tempt many root developmental biologists to move beyond the
21 comfort of the well-known Arabidopsis root meristem.

22

23 **Highlights**

- 24 ○ Plant root meristems are uniquely suitable for lineage tracing and trajectory analysis.
- 25 ○ Spatial transcriptomics will initially assist in validation of scRNA-seq data but might soon
26 become the main tool for transcriptomic profiling in plants.
- 27 ○ Gene regulatory networks obtained at single-cell level in multiple species will be an
28 invaluable tool to identify conserved regulators of root development.

29

30 **Keywords:** Root development, single-cell transcriptomics, single-nucleus transcriptomics,
31 spatial transcriptomics

32 **Abbreviations**

33	CITE-seq	Cellular Indexing of Transcriptomes and Epitopes by Sequencing
34	CRE	Cis-Regulatory Element
35	FACS	Fluorescence-Activated Cell Sorting
36	FANS	Fluorescence-Activated Nuclei Sorting
37	GRN	Gene Regulatory Network
38	GWAS	Genome-Wide Association Study
39	IACS	Intelligent Image-Activated Cell Sorting
40	ISH	<i>In Situ</i> Hybridization
41	LR	Lateral Root
42	scATAC-seq	Single-cell Assay for Transposase-Accessible Chromatin sequencing
43	scRNA-seq	Single-cell RNA-sequencing
44	SNP	Single-Nucleotide Polymorphism
45	snRNA-seq	Single-nucleus RNA-sequencing
46	TF	Transcription factor
47		

48 **Introduction**

49 Despite having joined the single-cell RNA-sequencing (scRNA-seq) party somewhat later
50 compared to mammalian and medical research colleagues, the plant field pioneered in
51 generating tissue specific transcriptomic data sets on Arabidopsis root apical meristems by
52 combining fluorescence-activated cell sorting (FACS) and microarray analysis or bulk RNA-
53 seq as early as 2003. These atlases became increasingly more detailed over the years as
54 technology advanced and even included cell-type level responses to a spectrum of abiotic and
55 biotic stresses [1–7]. Although the importance of these datasets for the entire plant community
56 cannot be stressed enough, the introduction of droplet-based scRNA-seq has undoubtedly
57 provided a massive increase in resolution of transcriptome maps in the Arabidopsis root apical
58 meristem [8–17]; and in other organs [18–24]. As this technology is not based on the
59 availability of tissue specific marker lines, it is quickly becoming a very important technology
60 in other plant species as well [25–35]. Despite being fully embraced by the plant community,
61 scRNA-seq technology is mostly being used to query gene expression in a spatiotemporal way,
62 similar to the FACS based data that has been around for almost 20 years [1–4]. There are
63 however clear examples of how the increase in spatiotemporal resolution has advanced our
64 understanding of root development [10,15,23,27], but scRNA-seq technology and the available
65 datasets have much more potential (Fig. 1). We are thus only scratching the proverbial surface
66 of what is already possible and will become possible in the very near future.

67 **The blessing and the curse of using plants for single-cell analyses**

68 It does not come as a surprise that the Arabidopsis root apical meristem was the first organ to
69 be studied using scRNA-seq, as individual cells can easily be generated by enzymatic digestion
70 of the cell wall in a process called protoplasting [1,36]. The capacity to generate single cells
71 has however been and will continue to be a main bottleneck for the plant community [37–40].
72 Besides potentially introducing an unwanted transcriptional response while generating
73 protoplasts, commercial systems based on microfluidics technology limit the size of captured
74 cells to about 30-50 μm . This poses a real problem as plant cells range from 10 μm to 100 μm
75 in size [41]. Although larger cells up to 125 μm can be captured using specific assays, this was
76 shown to introduce a bias in the relative abundance of subpopulations of large cells [42,43].
77 Furthermore, certain cell types might fail to be digested or specific cells might burst during the
78 procedure. Because of these reasons, some studies have resorted to nuclei isolation instead of
79 whole cells [29,31,44–51]. This approach is theoretically applicable to any organ and any plant
80 species, including frozen samples, and might prevent inherent capture biases associated with
81 generating single cells [47]. The main disadvantage is capturing fewer mRNAs compared to
82 whole cells, although single-nucleus RNA-sequencing (snRNA-seq) generally performs
83 equally well for sensitivity and classification of cell types [47,49,52–54].

84 Despite these and other pitfalls in applying single-cell technologies to plant samples, the plant
85 field also has distinct advantages compared to other fields of research which merit more
86 exploitation. For example, linking cell states across periods of time in mammalian systems is
87 very challenging and is currently approached by tracking cell clones via sequencing of inherited
88 barcode sequences [55]. In contrast, every root meristem contains cells at all differentiation

89 stages, orderly aligned in cell files and fixed within their tissue context. The Arabidopsis root
90 apical meristem is thus highly suitable for lineage tracing and developmental trajectory
91 analysis, eliminating the need for time-series experiments in whole tissues and the associated
92 batch effects, though care must be taken to avoid possible confounding effects due to positional
93 information [56]. So far, pseudotime trajectories have mostly been used to interpret
94 developmental time in scRNA-seq datasets and affirm known cell lineages [8,15,24]. They can
95 however be used to address more complex developmental processes in root biology such as
96 cell ontogeny and specification events. Indeed, trajectory analysis of protophloem cells
97 revealed a differentiation gradient that mediates cellular specification [17] and analysis of the
98 first stages of lateral root (LR) formation led to the identification of a group of precursor cells
99 that rapidly reprograms and splits into various LR cell fates [23]. Furthermore, trajectory
100 analysis can provide insights into the mode and speed of cell state transitions (gradual or
101 switch-like), reveal bifurcations in ontogeny, and discover new regulators of these processes.
102 As the Arabidopsis root meristem is very well studied, trajectory analysis is most likely to
103 reveal novel insights in less characterized species [27,33]. Although its potential is clear,
104 trajectory analysis and gene discovery require sufficient cells at each step of the pseudotime
105 and high data content per cell. At the moment, achieving such high-resolution data using whole
106 organ datasets presents a major financial burden. Thus, until sequencing technologies becomes
107 more affordable, dedicated tissue specific data sets, which contain much fewer cells but with
108 higher sequence coverage per cell will prove useful in studying root development and might
109 allow re-visiting of the text-book concept of tissue identities by e.g. defining functional units
110 of cells that span different tissue types.

111 **Increasing specificity in single-cell experiments**

112 Although scRNA-seq is capable of capturing rare cells or cell types, their occurrence in whole
113 organ atlases might still be insufficient to infer good statistical power or advance to gene
114 discovery and functional characterization studies. Although this issue can be partially solved
115 by profiling a larger number of cells [14], this comes with an unrealistic financial cost if high
116 data content per cell is required or many samples are involved. As mentioned above, this issue
117 can be resolved by enriching these rare cell states or tissues, resulting in more specific datasets.
118 In mammalian systems, this is achieved by using combinatorial antibody staining in e.g. CITE-
119 seq approaches [57–60]. Although large collections of antibodies and tissue-specific epitopes
120 are not readily available in the plant field, increased specificity can be facilitated by manually
121 removing unwanted tissues [20,21] or specific cell enrichment using FACS/Fluorescence-
122 activated nuclei sorting (FANS) on fluorescent protein tagged reporter lines whose expression
123 represent a spatiotemporal domain within the tissue of interest. The fact that increased
124 specificity leads to novel biological insights was elegantly shown by profiling the sieve element
125 lineage from cell birth to terminal differentiation [17], the Arabidopsis inflorescence [61], and
126 the first four stages of LR formation [23]. The major drawback of this method is obviously its
127 reliance on *a priori* knowledge and the availability of specific marker lines, which are rare or
128 absent for most plant species. Alternatives to purify cell types without resorting to specific
129 antibodies or transgenes have been suggested in the form of intelligent image-activated cell
130 sorting (IACS), which performs real-time high-throughput cell microscopy analysis prior to

131 sorting based on a range of morphological features [62]. Other computational methods combine
132 single-cell transcriptomics with FACS index sorting to set non intuitive sorting gates to purify
133 cell types based on scRNA-seq data [63]. Successful application of these technologies would
134 allow for smooth integration of other single-cell omics and streamline the identification of
135 molecular morphometric phenotypes. However, these methods require accurate training of
136 deep learning algorithms, so other unbiased methods that also do not require markers or
137 antibodies will be needed.

138 **Seeing is believing**

139 In all cases, predictions derived from scRNA-seq data should be validated experimentally, as
140 conclusions drawn from scRNA-seq data analyses can be skewed by biases generated during
141 sample or library preparation and the downstream computational analysis. This can be achieved
142 by generating reporter lines [8,15,17,20] or by performing *in situ* hybridization (ISH)
143 [24,27,33]. However, constructing reporter lines is limited to species that are amenable to
144 transformation and *in situ* hybridization is labour intensive and can be hindered by the lack of
145 robust probes in many plant species [64]. Fortunately, rapid progress is being made in the
146 development of spatial transcriptomics [65–70]. This technology theoretically enables all genes
147 at low spatial resolution (untargeted) or a subset of genes at high spatial resolution (targeted)
148 to be visualized *in situ*, without the need of marker genes or reporter lines and is applicable to
149 any species and tissue. Nevertheless, its application to the plant field is currently still hampered
150 by technical difficulties (see also [71]).

151 Combining the complementary advantages of single-cell and spatial transcriptomics will
152 revolutionize both fundamental and applied research. Indeed, linking scRNA-seq data with its
153 natural spatial context enables instant cell identity mapping, which is of particular use in non-
154 model organisms. Moreover, it provides unprecedented resolution to study structure-function
155 relationship, cell-cell interactions, plant pathogen interactions and environmental responses in
156 general. If the resolution and accessibility of untargeted spatial transcriptomics increases
157 further, it can even be envisioned that this technology will largely replace scRNA-seq in plant
158 research. Besides advancing basic root biology research in Arabidopsis, spatial transcriptomics
159 is expected to accelerate the establishment of new species for molecular biology applications.
160 No doubt, this technology will also be rapidly adopted to study crops and species which are
161 difficult to transform.

162 **From off-the-shelf to out-of-the-box**

163 Coordinated growth and development in a changing environment requires interplay among
164 many components in complex gene regulatory networks (GRNs), where transcription factors
165 (TFs) and non-coding functional *cis*-regulatory elements (CREs) cooperatively regulate gene
166 expression and as such determine the final cell differentiation start and phenotypical response.
167 Due to the high spatiotemporal resolution, single-cell data is able to deconstruct tissue
168 heterogeneity, making it highly suited for GRN analysis [72]. For example, environmental
169 GRNs have been constructed for Arabidopsis roots where heat-shock treatment led to drastic
170 transcriptional changes in the three outermost cell layers of the root [9]. In another study,

171 sucrose induced enrichment of root hairs and gene expression changes were highly tissue-
172 specific [11]. However, precise GRN predictions require CREs or TF-binding site information
173 with matching spatiotemporal resolution. This can be achieved by complementing scRNA-seq
174 data with profiling accessible chromatin regions through scATAC-seq [27,50,73]. As such, our
175 understanding of the regulatory networks governing root growth and development can be
176 achieved by pairing scRNA-seq with scATAC-seq across critical growth and developmental
177 transition stages or under a spectrum of environmental stresses.

178 Apart from answering fundamental research questions using off-the-shelf single-cell
179 applications, the high spatiotemporal resolution embedded in scRNA-seq data can be used in
180 more surprising ways to modulate complex traits in crop species. For example, scRNA-seq
181 data was linked with genome-wide association (GWAS) data in developing maize ears where
182 significant single-nucleotide polymorphisms (SNPs) were found within scRNA-seq marker
183 genes that are associated with yield-related traits [35]. As such, the high-resolution GRNs
184 constructed through scRNA-seq and scATAC-seq will help to pinpoint key regulators
185 underlying traits of interest together with corresponding CREs. SNPs within these CREs can
186 then potentially serve as targets for genome editing to precisely deliver targeted phenotypic
187 changes. Also, such high-resolution SNPs could facilitate applications like GWAS and marker-
188 assisted selection in plant breeding.

189 **Conclusion**

190 In the few years since they have been adopted by the plant field, single cell applications are
191 revolutionizing the way we study root development. Although they are still mostly increasing
192 our spatiotemporal resolution and identifying specific developmental regulators; soon, they
193 might tempt many root biologists to move beyond the well-studied Arabidopsis root
194 meristems and quickly prepare other species for molecular biology applications by providing
195 fully annotated transcriptome atlases. The applications are perhaps even more promising for
196 studying complex systems such as plant-pathogen interactions or cell-cell interactions, but
197 will for sure also be readily adopted in crop species.

198 **Declaration of competing interests**

199 The authors declare no competing interests

200

201 **Author contributions**

202 M.M., Y.K., M.S.S. and B.D.R. conceptualised and wrote the manuscript.

203

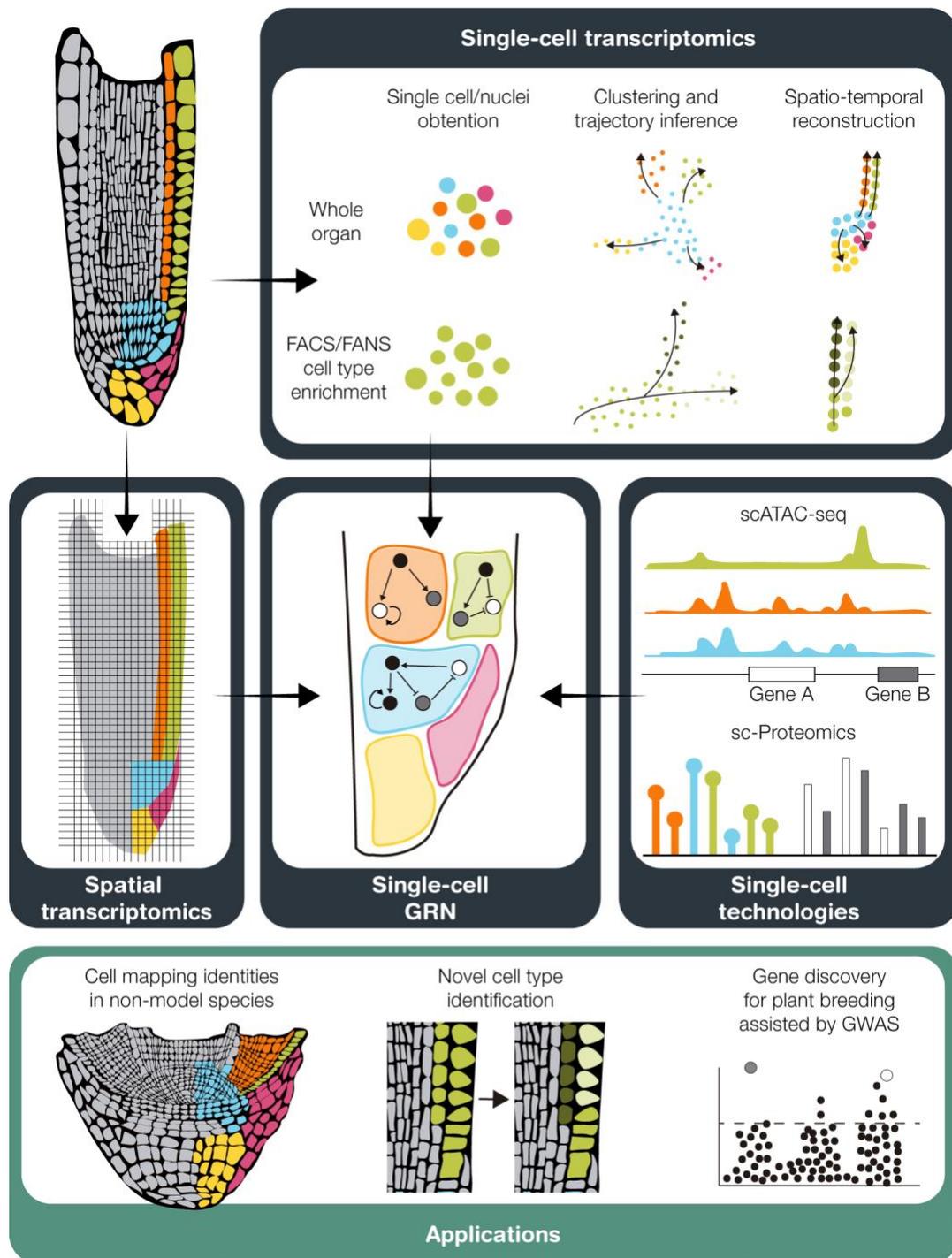
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212

213 **Figure 1. New opportunities for understanding root development using single-cell**
 214 **technologies.** The use of RNA-seq applied to individual cells or nuclei allows reconstructing
 215 developmental trajectories in root tissues. The resolution can be increased using FACS/FANS
 216 to study a specific tissue or cell type of interest. scRNA-seq data will be soon complemented
 217 with spatial transcriptomics technology. Combining the transcriptome information with the
 218 chromatin accessibility and proteome studies at single cell level could be used to create gene
 219 regulatory networks (GRN) with an unprecedented spatiotemporal detail. Root development
 220 research will benefit from these technologies to e.g. study non-model plant species, find new
 221 functional cell types, or discover target genes for plant breeding.

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