Morphological plant modeling: Unleashing geometric and topologic potential within the plant sciences

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HAL Id: hal-01412691
https://hal.inria.fr/hal-01412691

Preprint submitted on 8 Dec 2016

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58

59   **Keywords:** Plant biology, plant science, morphology, mathematics, topology, modeling

60   **Running title:** Plant morphological modeling

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71   **Abstract**
72
73   Plant morphology is inherently mathematical. The geometries of leaves and flowers and
74   intricate topologies of the root have fascinated plant biologists and mathematicians alike.
75   Beyond providing aesthetic inspiration, understanding plant morphology has become
76   pressing in an era of climate change and a growing population. Gaining an understanding of
77   how to modify plant architecture through molecular biology and breeding is critical to
78   improving agriculture, and the monitoring of ecosystems and global vegetation is vital to
79   modeling a future with fewer natural resources. In this white paper, we begin by
80   summarizing the rich history and state of the art in quantifying the form of plants,
81   mathematical models of patterning in plants, and how plant morphology manifests
82   dynamically across disparate scales of biological organization. We then explore the
83   fundamental challenges that remain unanswered concerning plant morphology, from the
84   barriers preventing the prediction of phenotype from genotype to modeling the fluttering
85   of leaves in a light breeze. We end with a discussion concerning the education of plant
86   morphology synthesizing biological and mathematical approaches and ways to facilitate
87   research advances through outreach, cross-disciplinary training, and open science. Never
88   has the need to model plant morphology been more imperative. Unleashing the potential of
89   geometric and topological approaches in the plant sciences promises to transform our
90   understanding of both plants and mathematics.
91
92   **I. Introduction**
93
94   *A. Plant morphology from the perspective of plant biology*
95
96   The basic morphology of most land plants—sessile, tubular and photosynthetic, with a
97   growing tip from which lateral organs originate and with finely-branched subterranean
98   organs that both anchor the plant and allow uptake of nutrients—has also evolved in other
99   multicellular eukaryotes. A broad definition of "plant" might include all eukaryotes capable
100   of photosynthesis (Niklas, 1997), which within green and red algae traces directly to the

101 single primary endosymbiosis event that resulted in the incorporation of a cyanobacterial
102 ancestor of the chloroplast into an ancient eukaryotic cell some one billion years ago.
103 Among the green plants (Viridiplantae), branching shoot-like architectures with leaf-like
104 appendages that are centers of photosynthesis have evolved in both land plants
105 (Embryophyta) and aquatic green algae (Charophyta and Chlorophyta) (Kaplan and
106 Hagemann, 1991). Algal holdfasts and vascular plant roots both anchor plants, but the
107 vascular plant root extends the plant body much deeper into the soil where it also functions
108 in nutrient uptake.
109
110 The study of plant morphology interfaces with all levels of biological organization (**Figure**
111 **1A**). Plant morphology can be descriptive and typological, as in systematics, focusing on
112 biological homologies to discern groups of organisms (Mayr, 1981; Wiens, 2000). These
113 studies additionally have important applications in agriculture and horticulture, beginning
114 with careful selection of ideal phenotypes. In plant ecology, the morphology of
115 communities, comprised of numerous individuals, defines vegetation types and biomes,
116 including their relationship to the environment. Plant morphology also informs plant
117 physiology, the study of the functions of plants, whereas plant genetics describes the
118 inheritance and, in modern times, the molecular basis of plant morphology (Kaplan, 2001).
119
120 Plant morphology is more, though, than an attribute affecting all levels of plant
121 organization; it is also dynamic. Developmentally, morphology reveals itself over the
122 lifetime of a plant through varying rates of cell division, cell expansion, and anisotropic
123 growth (Esau, 1960; Steeves and Sussex, 1989; Niklas, 1994) (**Figure 1B**). Responses to
124 changes in environmental conditions further modulate the abovementioned parameters.
125 Development is genetically programmed and driven by biochemical processes that are
126 responsible for physical forces that change the observed patterning and growth of organs
127 (Green, 1999; Peaucelle et al., 2011; Braybrook and Jönsson, 2016). Physical forces
128 continue to inform plant development, such as the growth of a root through heterogeneous
129 soil densities or the bending of branches and leaves to flows of air or water, long after the
130 initiation and patterning of young primordia (Niklas, 1992; Silk and Erickson, 1979; Vogel,

131   1984). Plant morphology evolves through inherited modifications of structure or

132   development, either incrementally or abruptly, over the generations (Niklas, 1997) (**Figure**

133   **1C**). A record of these changes over geologic time is preserved and correlates with the

134   paleoclimate, illuminating our understanding of the form and function of extant plants

135   today (Bailey and Sinnott, 1915). Development and evolution are the biological

136   mechanisms through which plant morphology arises, regardless of whether in a systematic,

137   ecological, physiological, or genetic context (**Figure 1**).

138

139   *B. Plant morphology from the perspective of mathematics*

140

141   In 1790, Johann Wolfgang von Goethe pioneered a perspective that transformed the way

142   we think about plant morphology: the idea that the essence of plant morphology is an

143   underlying process of deformation, and that seemingly differently shaped leaves and floral

144   organs within a single plant are homologously related (Goethe, 1790; Friedman and Diggle,

145   2011). The modern challenge that Goethe's paradigm presents is to quantitatively describe

146   deformations in phenotype by their underlying genetic, developmental, and environmental

147   causes. Thus, the fundamental challenge from a mathematical standpoint is how to define

148   shape descriptors to compare plant morphology in the context of a single deformation

149   process; or, how to remove all deformations from the shape that are not relevant to the

150   biological question of interest (Mio and Srivastava, 2004; Lie et al., 2010). A plethora of

151   mathematical approaches, utilizing techniques ranging from geometry and topology to

152   graph theory, exist to describe plant morphology and analyze it with statistical methods.

153

154   An important class of shape descriptors detail plant morphology from the perspective of

155   topology (**Figure 2A**). Topological analyses can compare shape characteristics

156   independently of events that deform and transform plant shape geometrically, providing a

157   framework by which plant morphology can be modeled. Algorithms that reduce the

158   branching architecture of plants to a set of intersecting lines or curve segments,

159   constituting a mathematical graph, are an intuitive representation of the plant form. Each

160   curve segment is denoted by an edge, and a vertex denotes a connection. Augmenting each

161   vertex with a 3D coordinate results in a skeletal 3D representation of plant architecture
162   derivable from imaging data (Bucksch et al., 2010; Bucksch, 2014a; Aiteanu and Klein,
163   2014). Such skeletal descriptions are used to derive quantitative measurements of lengths,
164   diameters, and angles (Bucksch and Fleck, 2011; Raumonen et al., 2013; Seidel et al., 2015).
165   Having an architectural description in place allows the definition of orders, either
166   biologically or in a more abstract sense. The relationships between orders, such as degree
167   of self-similarity (Prusinkiewicz, 2004) or nestedness (Godin and Ferraro, 2010) are used
168   to quantitatively summarize patterns of plant morphology.
169
170   Computer simulations use principles from graph theory, such as graph rewriting, to define
171   rules successively augmenting a graph with vertices and edges as plant development
172   unfolds. These rules recapitulate the differences observed in plant architectures across
173   plant species (Kurth, 1994; Prusinkiewicz et al., 2001). Morse theory (Milnor, 1963)
174   provides a rich toolset with which to model the morphology of plants. Roughly speaking, a
175   Morse function expands over the plant surface where the "intersection" or contour
176   between the Morse function and plant surface is evaluated during the expansion process.
177   Events such as merging, splitting, appearance, or disappearance of contours provide
178   avenues to quantify relationships between orders. Persistent homology (**Figure 2B-C**)
179   extends Morse theory with functions that deform a given plant shape gradually to define
180   self-similarity (MacPherson and Schweinhardt, 2012) and morphological properties
181   (Edelsbrunner and Harer, 2010) on the basis of event statistics.
182
183   In the 1980s, David Kendall defined an elegant alternative statistical framework to the
184   previously introduced topological descriptors that are capable of comparing outlines of
185   shapes independently of scale, rotation, and translation (Kendall, 1984). His idea to
186   compare the outline of shapes in a transformation-invariant fashion infused rapidly into
187   biology (Bookstein, 1997). Furthermore, Kendall's idea inspired the development of
188   methods such as elliptical Fourier descriptors (Kuhl and Giardina, 1982) and new trends
189   employing the Laplace Beltrami operator (Reuter et al., 2009), both relying on the spectral
190   decompositions of shapes (Chitwood et al., 2012a; Chitwood et al., 2012b; Laga et al. 2014).

191   From a mathematical point of view, developmental processes construct surfaces in a three-
192   dimensional space. Yet, this space in which development is embedded imposes constraints
193   on plant forms observed. Awareness of these constraints has led to new interpretations of
194   plant morphology (Prusinkiewicz and DeReuille, 2010; Bucksch et al., 2014b) and might
195   explain, at least in part, observations such as symmetry or asymmetry in leaf shape
196   (Martinez et al., 2016).
197
198   Parallel to strictly mathematical descriptions of plant morphology, Ronald Fisher
199   developed a statistical framework to partition variance into different sources of variability,
200   whether genetic, developmental, or environmental (Fisher, 1925). Specifically, with respect
201   to plant morphology, the iris flower dataset (Fisher, 1936) was used to develop novel
202   methods to differentiate three *Iris* species based on the length and width of sepals and
203   petals. The dataset is still used to this day as a benchmark to compare statistical methods[1].
204
205   In this white paper, we begin by summarizing the history and importance of the
206   inextricable link between mathematical approaches with advances in our understanding of
207   plant morphology. We proceed by outlining emerging questions at the interface of
208   mathematics and plant morphology, focusing on the acquisition of data, modeling, and
209   data-driven questions at all levels of plant organization. We conclude by describing
210   milestones that will signal the further infusion of mathematics into the plant sciences,
211   including integrated curriculums, outreach, and open science.
212
213   **II. A primer of the history, importance, and potential of mathematics in plant biology**
214
215   Perhaps because of their intricate topologies, geometries, and iterative growth, the
216   morphology of plants has inspired mathematicians; conversely, the biologist's conception
217   of the plant shape requires a mathematical basis. Below we briefly summarize the rich
218   history and state of the art in quantifying the plant form, mathematical models of

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[1] The Iris flower dataset, https://en.wikipedia.org/wiki/Iris_flower_data_set (retrieved September 24, 2016)

219   patterning in plants, and how plant morphology manifests dynamically across disparate

220   scales of biological organization and over time.

221

222   *A. Quantifying plant morphology*

223

224   *1. Capturing the dynamics of plant morphology across time.* Techniques to temporally

225   monitor plant development in 3D have been developed that are minimally invasive,

226   allowing the tracking of cell and whole organ morphology, and even plant behavior, over

227   timescales from seconds to days. Confocal microscopy, which eliminates out-of-focus light

228   using a pinhole, can be used to create optical sections of objects and reconstruct 3D models

229   (**Figure 3A**). Variations on confocal technology, such as Spinning Disk Confocal Microscopy

230   (SDCM; Oreopolous et al., 2013) and Light-Sheet Fluorescence Microscopy (LSFM; Maizel et

231   al., 2011; Sena et al., 2011; von Wangenheim et al., 2016), that minimize photo damage of

232   live samples, can be used to understand plant organogenesis using time-lapse imaging.

233

234   For example, by comparing cell proliferation rates of complex and simple leaf primordia

235   over time, the patterning of complex leaves in *Cardamine hirsuta* was demonstrated to

236   result from inhibition of growth at the boundary between leaflets (Vlad et al., 2014).

237   Similarly, early morphogenesis of lateral root primordia in *Arabidopsis* (**Figure 3A**) can be

238   temporally modeled by patterns of successive cell divisions (Maizel et al., 2011; von

239   Wangenheim et al. 2016). Cell lineage tracing during leaf organogenesis in *Arabidopsis*

240   allowed for the spatial mapping of variation in growth rate. The formulation of leaf growth

241   as a mathematical model incorporates local variations as a function of ontogenetic time

242   (Kuchen et al., 2012; Tauriello et al., 2015). Such studies demonstrate the power of

243   temporally tracking the origins of plant morphology down to the cellular level, revealing

244   the full, spatio-temporal morphology of plants. As we will discuss later, other technologies

245   can measure the consequences of plant development at grander, emergent scales.

246   Computerized tomography (CT) (**Figure 3B**) and Magnetic Resonance Imaging (MRI) can

247   be used to monitor the growth and morphology of mature organs, while radar, LIDAR, and

248   satellite imaging can monitor ecosystems and global vegetation (**Figures 3-4**).

249

250   *2. Morphometrics and allometry: the measurement of shape and size.* Once morphological

251   data has been collected, cells, young organ primordia, individual leaves, and roots—and by

252   extension canopies, root systems, and populations—can be described using morphometric

253   approaches. Shape features can be measured by hand, but increasingly machine vision

254   (Wilf et al., 2016) and other automated approaches are used to place landmarks or derive

255   elliptical Fourier descriptors from digital images of plant organs, allowing for statistical

256   and spectral decompositions of shapes, respectively (Iwata et al., 1998; Langlade et al.,

257   2005; Chitwood et al., 2012c; 2013; 2014a; 2014b; 2014c; Rellan-Alvarez et al., 2015).

258   Underlying the shapes of plant organs are differential rates of cell expansion. Cellular

259   expansion rates vary locally and can change rapidly deforming primordia into diverse

260   mature organ morphologies (Rolland-Lagan et al., 2003; Remmler and Rolland-Lagan,

261   2012; Das Gupta and Nath, 2015). Allometry, or the relationship between shape and size, is

262   a principal consideration when analyzing the factors linking the growth of an organ to its

263   resultant morphology. Just as allometry scales from the cellular to tissue and organ levels,

264   it scales from the individual to population levels, explaining intricate morphologies on a

265   global scale (Duncanson et al., 2015; Huang et al., 2016; Paul et al., 2015).

266

267   *B. Modeling pattern formation in plants*

268

269   *1. Turing, phyllotaxy, and reaction-diffusion modeling.* The beautiful patterns apparent in the

270   regular arrangement of leaves on stems and their connection to mathematics have inspired

271   scientists to create models explaining these phenomena. Alan Turing, a pioneering figure in

272   twentieth-century science, developed a longstanding interest in phyllotaxy, the

273   arrangement of leaves on a plant. Turing's approach to the problem was twofold: first, a

274   detailed geometrical analysis of the patterns (Turing, 1992); and second, an application of

275   his theory of morphogenesis through local activation and long-range inhibition (Turing,

276   1952). Combining physical experiments with computer simulations, Douady and Couder

277   (1996) subsequently modeled a diffusible chemical produced by developing primordia that

278   would inhibit the initiation of nearby primordia, successfully recapitulating known

279 phyllotactic patterns. Such reaction-diffusion systems, which model *de novo* pattern

280 formation as a process resulting from the spatial diffusion and interaction of substances

281 over space and time, can be applied to a wide variety of processes governing plant pattern

282 formation, including stem cell homeostasis and patterning of the shoot apical meristem

283 (Bernasconi, 1994; Meinhardt, 2004; Hohm et al., 2010; Fujita et al., 2011), floral organ

284 number (Kitazawa and Fujimoto, 2015), the regular spacing of root hairs (Meinhardt and

285 Gierer, 1974), and vascular patterning (Meinhardt, 1976). Studies on pattern formation in

286 plants, which originally inspired Turing and other reaction-diffusion modelers, have

287 contributed to universal theories of pattern formation throughout life (Gierer and

288 Meinhardt, 1972; Pearson, 1993).

289

290 *2. Auxin transport-feedback models of patterning.* The plant hormone auxin has been

291 determined (at least in part) to empirically contribute to many of the abovementioned

292 patterning events. Auxin distribution in plants is achieved through complex, transport-

293 feedback mechanisms. In order to understand how a single hormone could produce such

294 diverse patterns in a self-organizing manner, these feedback mechanisms have been

295 extensively modeled with dynamic system techniques (a complete review of auxin

296 transport mechanisms and their role in plant pattern formation is outside the scope of this

297 manuscript, but see Smith and Bayer, 2009). Auxin transport creates discrete patterning

298 events that are inconsistent with the gradients through which morphogens in animals

299 typically act, demonstrating that pattern formation in plants and animals is fundamentally

300 different (Bhalerao and Bennett, 2003). Elegant work combining modeling with

301 experimental verification has demonstrated the role of auxin in diverse plant patterning

302 processes including embryonic patterning (Moller and Weijers, 2009), patterning of the

303 female gametophyte (Pagnussat et al., 2009), vascular patterning (Scarpella et al., 2006;

304 Donner et al., 2009), the patterning of serrations and leaflets in complex leaves (Koenig et

305 al., 2009), lateral root formation (Casimiro et al., 2001), and phyllotaxy (Reinhardt et al.,

306 2003; Smith et al., 2006), among other processes too numerous to list here.

307

10

308 *3. Geometric models of cell division orientation.* Plant morphology is determined largely at
309 the organismal level, rather than an emergent property resulting from the collective
310 behavior of cells as in animals (Kaplan and Hagemann, 1991). Yet, as multicellular
311 organisms, cell division and expansion are inextricably linked to plant development (Alim
312 et al., 2012). Several attempts have been made to formulate universal rules predicting the
313 position of cell divisions. In the nineteenth century, Sachs proposed that plant cells divide
314 into two equally sized daughter cells with a division plane at right angles to the pre-existing
315 walls (Sachs, 1878). Leo Errera drew analogies between plant cells and soap bubbles and
316 posited that cells minimize their surface area. The predicted division plane among those
317 cells that respect Sachs' rules is thus the shortest possible (Errera, 1886). Yet, this rule fails
318 to fully account for observed patterns in cell division planes. Bessons and Dumais proposed
319 that the probability for a given plane of division is inversely proportional to the exponential
320 of its length (Besson and Dumais, 2011). Recently, for the *Arabidopsis* embryo, these
321 observations were extended into 3D (Yoshida et al., 2014), in which symmetric divisions
322 occur along a plane predicted by the Bessons-Dumais rule, but asymmetric divisions do not.
323 Inhibition of auxin response causes all cells to divide according to the "shortest wall" rule,
324 suggesting that auxin allows cells to deviate from the default rule.
325
326 *4. Tensor-based modeling of growth.* The pattern of cell walls in a root or shoot meristem
327 follows observable regularities. Two families of orthogonal lines can describe the patterns
328 observed in cell walls: periclines (divisions parallel to a local axis) and anticlines (divisions
329 normal to a local axis) (Sachs, 1878). Attempts to understand the origin of this regularity
330 led to the development of tensor-based models for growth and cell divisions. These models
331 share the underlying assumption of three mutually orthogonal principal directions of
332 growth (PDG) at each point of a growing meristem and the assumption that cells divide in
333 relation to PDGs (Hejnowicz and Romberger, 1984). Several tensor-based growth models
334 have been developed for the shoot apical meristem, the primary root meristem (Hejnowicz
335 and Karczewski, 1993; Nakielski, 2008), and the lateral root (Szymanowska-Pułka and

336 Nakielski, 2010; Szymanowska-Pulka et al., 2012). Such models provide a histogenetic link

337 between theories explaining cell division patterns with the morphogenesis of plant organs.

338

339 *5. Quantifying branching structures.* Maybe because of their prominent appearance in

340 everyday life, branching structures are one of the first geometrical forms recognized in

341 plants. Their representation is intuitive because they can easily be reduced to intersecting

342 line segments. A plethora of approaches to reduce imaged branching structures in 2D and

343 3D to lines or skeletons exist (see Bucksch 2011 and Bucksch, 2014a for an overview). In

344 principle, skeletons let us navigate through a branching structure and take measurements

345 of branch length and diameter. In tree crowns (Bucksch and Fleck 2011; Schilling et al.;

346 2012, Delagrange et al. 2014), skeletons are a fundamental descriptor towards

347 understanding self-shadowing in tree crowns and measuring biomass allocation over time.

348 Skeletonization methods are also a popular method to quantify the branching architecture

349 of roots, both statically at a single time point (Lobet et al., 2011; Galkovskyi et al., 2012)

350 and dynamically capturing root growth (Symonova et al., 2015).

351

352 The beauty and complexity of branch architectures has inspired a rich history of efforts to

353 quantitatively summarize them. One of the earliest efforts by Leonardo da Vinci described

354 allometric laws in the branching of tree crowns. Da Vinci quantified the hierarchical

355 relationships between diameters and length of branches (Long, 1994; Eloy, 2011), a

356 phenomenon later similarly observed in tree roots (Oppelt et al., 2001). Biologists visually

357 observed these underlying laws and created classes of tree architecture, noting, "trees in

358 the forest rarely exist in the ideal state …" (Halle et al., 1978). Quantitatively revealing the

359 "ideal state" of tree architecture has consumed researchers for over four decades. Early

360 fractal approaches to capture the repetitive and modular appearance of branching

361 structures are described in Henry Horn's pioneering book *The Adaptive Geometry of Trees*

362 (Horn, 1971), which inspired researchers for decades (Borchert and Slade, 1981; Zeide and

363 Pfeifer, 1991, Valladares, 1999; Godin and Ferraro, 2010). In turn, this view of a genetic

364 mechanism driving the developmental program of tree architecture has been realized

365 computationally (Palubicki, 2013), simulating the development of tree architectures

366 (Runions et al., 2007; Palubicki et al., 2009). Equivalently, functional-structural root models

367 simulate the efficiency of nutrient and water uptake following the development of roots

368 (Nielsen et al., 1994; Dunabin et al., 2013).

369

370 *C. Dynamic interactions of morphology between biological scales, with the environment, and*

371 *over time*

372

373 *1. Kinematics.* Although we have already discussed time-dynamic models of plant

374 morphogenesis and cellular division patterns above, the emergent morphologies that arise

375 from cells, tissues, and organs interact with the surrounding environment in functional

376 ways. For example, in roots there are three developmental zones starting at the tip that

377 propel cells through the soil medium: 1) the meristematic zone, where cells are actively

378 proliferating; 2) the elongation zone, where cells stop dividing and begin to expand; and 3)

379 the differentiation zone, where cells acquire their terminal cell fate (Barnes, 1898). Root

380 growth is controlled by both displacement via production of new cells (meristematic zone)

381 and expansion of existing cells (elongation zone). Describing a growth system with both

382 expanding and contracting components is a challenge. One approach to solve this problem,

383 pioneered by Wendy Silk (Silk and Erickson, 1979; Silk, 1984), is to adopt the formalism of

384 compressible fluid dynamics to describe plant growth in terms of kinematics. This

385 approach treats the plant "fluid" as a continuous medium, neglecting molecular-scale

386 heterogeneity. Considering the root as a homogenous fluid (which is theoretically

387 supported, if land plants are considered to be giant, single-celled organisms; Kaplan and

388 Hagemann, 1991; Coneva and Chitwood, 2015), this approach can be applied to questions

389 such as the relationship between water potentials and roots growth (Sharp et al., 1988;

390 1990) and the allometric growth of leaves (Bernstein et al., 1993; 1995).

391

392 *2. Plant-wind interactions.* While conceptualizing the root as a homogenous fluid can be a

393 useful simplification, the reality is that plants—the terrestrial Embryophyta and aquatic

394 Charophyta and Chlorophyta—live within currents of air and water. Work from Steven

395 Vogel and Laura Miller views the architecture of plants from the perspective of their fluid

13

396　dynamic properties (Vogel, 1970; 1989; 1992; 2009; Miller et al., 2012). Plants that live in

397　mechanically extreme environments, such as those with frequent hurricanes, strong winds,

398　and tornados, appear to be better adapted to resist wind forces. Plant architecture is

399　modified at all levels to reduce the drag that acts on a plant and avoid resonant forcing in

400　addition to structural reinforcements of the trunk and roots. Even at low wind speeds,

401　exchange through leaf and petiole designs that augment passive movements in the wind,

402　such as leaf flutter, may be physiologically relevant (Schuepp, 1972; Grace, 1978; Roden

403　and Pearcy, 1993; Roden, 2003). Regardless of whether a plant lives in environments with

404　frequent winds, the morphology of plants must reckon with other fluid dynamic effects,

405　such as the boundary layer (Schuepp, 1993). The effects of leaf size on the boundary layer

406　are well-known, influencing transpiration losses, the diffusion of photosynthetic gases, and

407　heat dissipation (Parkhurst and Loucks, 1972). In addition to shape and size, plant

408　morphology at finer scales, including epidermal features such as trichomes and wax

409　deposition, may influence the boundary layer and therefore plant fitness. Later, we detail

410　models of plant-fluid interactions in depth.

411

412　*3. Leaf shape and the paleorecord.* Plant morphology responds dynamically to turbulent

413　environments and can plastically respond, developmentally, to subtler changes. But what is

414　the relationship between plant morphology and geologic time? Ultimately, the shape of

415　leaves is subject to selection pressures from a variety of sources, whether the influence of

416　high winds, herbivory, thermoregulation, or light interception. Plant morphology may be

417　constrained by hydraulic, biomechanical, developmental, or phylogenetic effects (Nicotra et

418　al., 2011). Any one of these factors—and other unknowns—can create patterns observed in

419　both extant and fossil leaf assemblages (Chitwood and Sinha, 2016).

420

421　Larger leaves, with fewer and less pronounced serrations, predominate in warmer, wetter

422　climates whereas smaller leaves, with larger and more numerous serrations, tend to be

423　found in colder, drier temperate environments (Bailey and Sinnott, 1915). This

424　fundamental relationship between leaf dissection with temperature and precipitation has

425　been demonstrated across geographies and in different phylogenetic groups, although

14

426   often the correlation differs or is more or less robust depending on context (Wolfe, 1979;

427   1993; Wing and Greenwood, 1993; Wilf, 1997; Wilf et al., 1998; Huff et al., 2003; Jacobs and

428   Herendeen, 2004; Royer et al., 2005; Spicer et al., 2009; Peppe et al., 2011). Plasticity in leaf

429   shape during the lifetime of an individual plant correlates with temperature and

430   precipitation similarly to the fossil record, demonstrating parallels in the morphological

431   response of plants to changes in environmental conditions over developmental and

432   evolutionary timescales (Royer et al., 2009; Chitwood et al., 2016). Measuring plant

433   morphology in evolving populations provides insights into adaptive versus neutral features

434   (Moyle, 2008; Nordborg and Weigel, 2008; Brachi et al., 2011), revealing insights into the

435   relationship between morphology and function (Wright et al., 2004).

436

437   **III. Emerging questions and barriers in the mathematical analysis of plant**

438   **morphology**

439

440   A true synthesis of plant morphology, comprehensively modeling observed biological

441   phenomena and incorporating a mathematical perspective, remains out of reach. In this

442   section we highlight the challenges facing the study of plant morphology, including the

443   limits of acquiring morphological data, phenotype prediction, responses of plants to the

444   environment, integrating models across biological scales, and the modeling of complex

445   phenomena that remain elusive, such as the reconfiguration of plant architecture to gale

446   force winds.

447

448   *A. Technological limits acquiring plant morphological data*

449

450   Before the invention of digital photography in 1969 (Smith, 2010), the measurement of

451   plant architecture was a manual process performed directly on the plant or using

452   traditional photographs (Atkins, 1843; 1853). Describing morphology at smaller

453   anatomical or cellular scales involved qualitative descriptions, sketches, and the use of

454   camera lucida to faithfully record the perceptions of the researcher (von Nägeli, 1863).

455   Digital imaging devices enabled the shape of plants to be captured on all biological scales in

456   2D and 3D. The most accessible scale is plant architecture. In general, technology to record

457   plant architecture produces two kinds of data: surface samples and volumetric data.

458   Stereovision data are generated from different views of 2D digital images (Boyde, 1973;

459   Dumais and Kwiatkowska, 2002). Through matching of identical features, a spatial

460   alignment of the images is computed that allows the calculation of depth information.

461

462   Time of flight scanners, such as terrestrial laser scanning, overcome unit-less measurement

463   by recording the round-trip time of hundreds of thousands of laser beams sent at different

464   angles from the scanner to the first plant surface within the line of sight (Vosselman and

465   Maas, 2010) (**Figure 4**). The constant speed of light allows the calculation of the distance

466   between the point hit on the plant surface and the laser scanner. Both stereovision and

467   laser scanning produce surface samples or point clouds as output. However, both face

468   algorithmic challenges if plant parts occlude each other, as both rely on the reflection of

469   waves from the plant surface (Bucksch, 2011).

470

471   Penetrating the plant surface to resolve occlusions is possible with X-ray and magnetic

472   resonance imaging (MRI). While both technologies circumvent occlusions and can even

473   penetrate soil, their limitation is the requirement of a closed imaging volume. X-ray

474   systems rely on the principle that different tissues and materials absorb different amounts

475   of X-ray radiation (Kumi et al., 2015.) (**Figure 3B**). A detector plate records the remaining

476   X-ray radiation. If X-ray images are taken from various angles around the plant, then a 3D

477   model can be reconstructed from the absorption levels. The field produced with two

478   cylindrical magnets defines the imaging volume of an MRI (van Dusschoten et al., 2016).

479   The hydrogen molecules of the plant align with the field of the first magnet. Creating a

480   second magnetic field causes a detectable change in hydrogen alignment, from which the

481   3D structure of a plant can be reconstructed. The resulting data are voxels that encode the

482   local hydrogen density as an intensity value. Both surface-penetrating imaging technologies

483   intrinsically limit the imaging volume, either by the need of a detector plate or the closed

484   cylinder needed to establish a magnetic field. Thus, MRI and X-ray are destructive if applied

16

485   to mature plant organs such as roots in the field or tree crowns that are larger than the

486   imaging volume (Fiorani et al., 2012).

487

488   Radar provides another non-invasive technique to study individual tree and forest

489   structures over wide areas. Radar pulses can either penetrate or reflect from foliage,

490   depending on the selected wavelength (Kaasalainen et al., 2015). Most radar applications

491   occur in forestry and are being operated from satellites or airplanes, although more

492   compact and agile systems are being developed for precision forestry above- and below-

493   ground (Feng et al., 2016).

494

495   *B. The genetic basis of plant morphology*

496

497   One of the outstanding challenges in plant biology is to link the inheritance and activity of

498   genes with observed phenotypes. This is particularly challenging for the study of plant

499   morphology, as both the genetic landscape and morphospaces are complex: modeling each

500   of these phenomena alone is difficult, let alone trying to model morphology as a result of

501   genetic phenomena (Benfey and Mitchell-Olds, 2008; Lynch and Brown, 2012; Chitwood

502   and Topp, 2015). Although classic examples exist in which plant morphology is radically

503   altered by the effects of a few genes (Doebley, 2004; Clark et al., 2006; Kimura et al., 2008),

504   many morphological traits have a polygenic basis (Langlade et al., 2005; Tian et al., 2011;

505   Chitwood et al., 2013; 2014b). Two approaches to link genotype to phenotype for complex

506   traits are quantitative trait loci (QTL) analysis and genome wide association studies

507   (GWAS). Both approaches identify statistical associations between genetic variants and

508   phenotype, the former usually through a controlled cross and the latter often relying on

509   historical recombination and linkage disequilibrium (Flint-Garcia et al., 2003; Meyer and

510   Purugganan, 2013).

511

512   QTL analyses for plant morphology span scales from the cellular to the whole organ level.

513   At the cellular level, root cortex cell number (Ron et al., 2013), the cellular basis of carpel

514   size (Frary et al., 2000), and epidermal cell area and number (Tisne et al., 2008) have been

515   analyzed. The genetic basis of cellular architecture ultimately affects organ morphology,

516   and quantitative genetic bases for fruit shape (Monforte, et al., 2014; Paran and van der

517   Knaap, 2007), root architecture (Zhu et al., 2005; Clark et al., 2011; Topp et al., 2013; Zurek,

518   et al., 2015) and leaf shape (Langlade et al., 2005; Ku et al., 2010; Tian et al., 2011;

519   Chitwood et al., 2013; 2014a; 2014b; Zhang et al., 2014; Truong et al., 2015) have been

520   described. Increasingly, studies are linking the effects of cell shape and organ morphology

521   to pleiotropic effects on plant architecture. Thompson et al. (2015) link shoot apical

522   meristem (SAM) shape to adult plant traits such as leaf shape, flowering time, and yield.

523   Similarly, high-throughput image processing of the SAM in a diverse panel of maize inbreds

524   has uncovered candidate genes involved in hormone transport, cell division, and cell size

525   that correlate the morphology of the meristem with SNPs residing within candidate genes

526   (Leiboff et al., 2015).

527

528   Natural variation in cell, tissue, or organ morphology ultimately impacts plant physiology.

529   For example, root cortical aerenchyma formation reduces the metabolic costs of soil

530   exploration, thereby improving plant growth under conditions of suboptimal availability of

531   water and nutrients (Zhu et al. 2010; Postma and Lynch, 2011; York et al., 2013; Lynch et

532   al., 2014; Lynch, 2013; 2015, Chimungu et al. 2015). Maize genotypes with greater root

533   cortical cell size or reduced root cortical cell file number also have reduced metabolic costs,

534   and therefore deeper rooting and increased water capture under drought (Chimungu et al.,

535   2014a; 2014b). The radial distribution of auxin in the rice root leads to differential cell

536   expansion and more vertical root angles, resulting in greater water capture in drying soils.

537   The genetic basis of this effect was mapped to *DRO1* (Uga et al., 2013), and exemplifies how

538   the control of cellular organization by a single gene can have profound effects at the

539   organismal level and the interaction of plants with their environments.

540

541   High-throughput phenotyping techniques are increasingly used to reveal the genetic basis

542   of natural variation. In doing so, phenotyping techniques complement classic approaches of

543   reverse genetics and often lead to novel insights even in a well-studied species like

544   *Arabidopsis*. Such techniques reveal a genetic basis for such dynamic traits as root growth

545   (Slovack et al., 2014) and by studying the genetic basis of cellular traits can reveal the

546   developmental genetic basis of root architecture (Meijón et al., 2014). Similarly, high-

547   resolution sampling of root gravitropism has led to an unprecedented understanding of the

548   dynamics of the genetic basis of plasticity (Miller et al., 2007; Brooks et al., 2010; Spalding

549   and Miller, 2013). Molecular technologies, such as next-generation sequencing, also

550   provide a mean for the high-throughput determination of the genetic basis of molecular

551   traits. eQTL analysis (gene expression QTL) reveals the genetic basis of tens of thousands of

552   transcript abundance levels (Jansen and Nap, 2001; Druka et al., 2010). Such an approach

553   demonstrates that single loci with profound pleiotropic effects on plant architecture, such

554   as *ERECTA* in *Arabidopsis*, can influence the expression of numerous genes in the genome

555   (Keurentjes et al., 2007). The genetic basis of gene expression can be linked to QTL

556   affecting plant morphology (Majewski and Pastinen, 2011; Chitwood et al., 2013; 2014b;

557   Anderson et al., 2014; Ranjan et al., 2016), providing yet another genetic link between

558   emergent levels of organization.

559

560   *C. The environmental basis of plant morphology*

561

562   Phenotypic plasticity is defined as the ability of one genotype to produce different

563   phenotypes based on environment (Bradshaw 1965; DeWitt and Scheiner, 2004) and adds

564   to the phenotypic complexity created by genetics and development. This produces both

565   challenges and opportunities for mathematical modeling. Trait variation in response to the

566   environment has been defined classically using reaction norms (originally

567   "Reaktionsnorm") where the value of a certain trait is plotted against different

568   environments (Woltereck, 1909). If the reaction norm line is flat, the trait is not plastic; if

569   the reaction norm varies across the environment the trait is plastic and the slope of the

570   reaction norm line will be a measure of the plasticity. Significant differences in slopes

571   among genotypes indicate a GxE interaction (Via and Lande, 1985). If a certain trait in one

572   genotype remains constant in contrasting environments, the trait is said to be canalized

573   (Waddington, 1953).

574

575   Seminal work by Clausen, Keck, and Hiesey (1941) demonstrated using several clonal

576   species in a series of reciprocal transplants that although heredity exerts the most

577   measureable effects on plant morphology, environment is also a major source of

578   phenotypic variance. Research continues to explore the range of phenotypic variation

579   expressed by a given genotype (phenotypic plasticity + developmental noise) in the context

580   of different environments, as this has important implications for many fields, such as

581   conservation, evolution, and agriculture (Nicotra et al., 2010; DeWitt, 2016). Many studies

582   examine phenotypes across latitudinal or altitudinal gradients, or other environmental

583   clines, to characterize the range of variation possible and its relationship to the

584   environment (Cordell et al. 1998; Díaz et al., 2016). Community effects are also essential to

585   understanding the origins of plant morphology, and there has been increased interest in

586   the role of biotic effects such as hybridization, pathogens, shading, and crowding (Nicotra

587   et al., 2010; Holeski et al., 2012; Atlan et al., 2015), as well as abiotic contributors such as

588   nutrient availability (Dorken and Barrett, 2004), to plasticity.

589

590   Many plastic responses by plants to the environment are conserved and have been studied

591   in detail from the molecular to ecological levels. One such example is phototropism: plants

592   can sense the direction of illumination and grow towards it (Briggs, 2014). Another

593   example of shoot environmental plasticity is neighbor detection and shade avoidance

594   (Casal, 2013; Schmitt et al., 2003). Light transmitted through or reflected from leaves has a

595   unique spectral signature that plants use as a cue for incipient or ongoing photosynthetic

596   competition. In response to neighbor shade, some plants make striking morphological

597   changes to compete for light. Elongation of stems and petioles can increase two-fold or

598   more. This increase in stem elongation comes at the expense of allocation to laminar

599   outgrowth of leaves and fruit and seed set. Additionally, plants undergoing shade

600   avoidance show increased apical dominance (reduced branching) as growth upwards is

601   favored over growth outwards to reach light before competitors. Understanding the

602   molecular basis of the shade avoidance response has implications for agriculture.

603   Improvements are anticipated through manipulation of resource allocation in competition

604   scenarios of neighboring plants at high planting densities versus crop production standards
605   (Carriedo et al., 2016).
606
607   Below ground, plants encounter diverse sources of environmental variability, including
608   water availability, soil chemistry, and physical properties like soil hardness and movement.
609   These factors vary not only between individual plants (Razak et al., 2013), but within an
610   individual root system, to which plants respond in diverse ways (Drew, 1975; Robbins and
611   Dinneny, 2015). Plasticity at a micro-environmental scale has been linked to developmental
612   and molecular mechanisms (Bao et al., 2014). The scientific challenge is here to integrate
613   these effects at a whole root system level to understand the optimal acquisition in resource
614   limited conditions (Rellan-Alvarez, et al., 2016) (**Figure 5**).
615
616   *D. Integrating models from different levels of organization*
617
618   Most of the mathematical models highlighted in the previous sections have been designed
619   independently to study plant development at specific spatial and temporal scales. Since it is
620   extremely difficult to examine complex interdependent processes occurring at multiple
621   spatio-temporal scales, mathematical modeling can be used as a complementary tool with
622   which to disentangle component processes and investigate how their coupling may lead to
623   emergent patterns at a systems level (Hamant, 2008; Band and King, 2012; Jenzen and
624   Fozard 2015; Band et al. 2012).
625
626   A detailed multiscale model may involve hundreds of variables and parameters. It is likely
627   that the exact values of most parameters will not be known, nor even identifiable. Thus, to
628   be practical, a multiscale model should generate well-constrained predictions despite
629   significant parameter uncertainty (Gutenkunst et al., 2007, Hofhuis et al., 2016). It is
630   desirable that a multiscale model has certain modularity in its design such that individual
631   modules are responsible for modeling specific spatial aspects of the system (Baldazzi et al.,
632   2012). Global sensitivity analysis can be applied to reveal how individual modules function
633   when other modules are perturbed (Sudret, 2008). Most importantly, a multiscale model

634   must be tested against available data. An important future problem in plant mathematical

635   biology is how to integrate existing mathematical models that describe plant

636   morphogenesis at different spatial scales (Hill et al., 2013).

637

638   It is a useful exercise to illustrate the challenges of integrating models at the molecular and

639   cellular levels with an example. At the molecular scale, mathematical models describe the

640   spatiotemporal dynamics of key biomolecules and phytohormones controlling the function

641   and fate of individual cells. Some of these biomolecules can be diffusive, but others, such as

642   membrane-bound receptors, may be spatially localized (Battogtokh and Tyson, 2016). The

643   distribution of molecules in the cell i (i=1..N) can be described by a system of partial

644   differential equations, $\frac{dX_i}{dt} = f(X,\mu) + D_X \sum_{j=neighbors}(X_j - X_i)$, where the components of

645   X are the levels of biomolecules in a given cell, the nonlinear functions f describe the

646   kinetics of the reactions between the molecules, μ is the set of model parameters, DX is a

647   diagonal matrix of diffusion coefficients, N is the number of the cells, and the sum in the last

648   term is for the diffusion of the biomolecules between the cells (Fujita et al., 2011).

649

650   At the cellular scale, mathematical models describe dynamics of cell networks where the

651   mechanical pressures exerted on the cell walls are important factors for cell growth and

652   division (Jensen and Fozard, 2015) (**Figure 6A**). In models describing plant development in

653   a two-dimensional cross-section geometry, cells are often modeled as polygons defined by

654   walls between neighboring cells. The spatial position of a vertex, where the cell walls of

655   three neighboring cells coalesce, is a convenient variable for mathematical modeling of the

656   dynamics of cellular networks (Prusinkiewicz and Lindenmayer, 2012). A given vertex is

657   driven by the sum of forces acting on it, the turgor pressures from the neighboring cells,

658   and the elastic forces exerted on the cell walls. Thus, the time evolution of the vertex j

659   located at the position x can be described by the dynamic equation: $\frac{d^2X_j}{dt^2} = F_j$, where Fj is

660   the total force acting on the vertex j. By numerically integrating the dynamic equations for

661   the positions of all the vertexes in the polygonal cells, the dynamics of the cell network can

662   be simulated. Additionally, certain rules can be adopted for cell division (Besson and

663   Dumais, 2011), which may depend on the size of a cell (Fujita et al., 2011), on its location in

664   the network (Prusinkiewicz et al., 2001), and on the local levels of hormones and

665   biomolecules (Dello Ioio et al., 2008).

666

667   A multiscale model can be assembled by combining the models at molecular and cellular

668   levels. Such a model allows the computation of spatiotemporal dynamics of biomolecules X

669   for realistic initial and boundary conditions on the dynamic cellular networks. Mutations

670   and deletions of the genes encoding the biomolecules can be modeled by changing the

671   parameters, or by modifying the reaction terms in the function f. By inspecting the effects

672   of such modifications on the dynamics of cellular networks, the relationship between

673   genotypes and phenotypes can be predicted. In the aforementioned model by Fujita et al.

674   (2011), the biomolecules X describe the dynamics of the proteins involved in the stem cell

675   regulation of the SAM in *Arabidopsis*. It can simulate SAM development in wild type and

676   mutant plants, demonstrating how integrative models can be valuable tools in predictive

677   modeling of plant development (**Figure 6B**).

678

679   *E. Modeling the interaction between plant morphology and fluid dynamics*

680

681   Modeling plant morphology is a complicated endeavor, but the forces that interact with

682   plants, such as currents of wind and water, are sometimes even more complex and affect

683   the development and evolution of plant architecture. Smaller scale airflows around plants

684   can affect plant health as much as gale force winds, but are determined by plant

685   morphology at an exquisite level of detail. For example, how does vegetation density affect

686   average flow speeds within complex canopies and how do the dynamics of the airflow

687   affect the movement of the trunks, stems, branches, and individual leaves? Simulating flow

688   on scales that span the individual leaf to the entire canopy are not currently feasible but

689   nonetheless critical for understanding plant morphology. Below, we highlight efforts to

690   model the interaction of currents with plant morphology at two extreme scales—light

691   breezes and gale force winds.

692

693   *1. The role of light breezes in cooling and gas exchange.* Leaf shape and material properties

694   that alter the boundary layer of the fluid over the surface of the leaf or enhance passive

695   movement can potentially augment gas and heat exchange. For example, it has been

696   proposed that the broad leaves of some trees flutter for the purpose of convective and

697   evaporative heat transfer (Thom, 1968; Grant, 1983). Fluttering may also allow more light

698   to penetrate the canopy (Roden and Pearcy, 1993).

699

700   One way to quantify the relative importance of advective and diffusive transport rates is by

701   using the Peclet number. The Peclet number is given as Pe = UL/D, where U is a velocity,

702   such as wind speed, L is a length, such as the length of the leaf, and D is the diffusivity of the

703   quantity of interest (heat, O_2, CO_2, etc.). If Pe << 1, then the transport of the heat or

704   chemical gradient is dominated by diffusion. If Pe>>1, then the movement of chemicals and

705   the dissipation of heat is dominated by advection. Note that diffusive transport alone is a

706   relatively slow process.

707

708   The morphology and movement of leaves change the boundary layer near the surface, and

709   subsequently alter the relative importance of advective and diffusive transport represented

710   by the Peclet number. Boundary layers describe the fluid layers close to surfaces in which

711   the flow of air or water approaches zero. Note that for any viscous fluid, the velocity of air

712   at a surface relative to the velocity of the surface itself is always zero. This is called the no-

713   slip condition. Diffusive transport is typically dominant very close to surfaces, and

714   advective transport dominates farther from surfaces.

715

716   The morphology and mechanical properties of leaves can alter the boundary layer. For

717   example, trichomes, the hair-like protrusions on the surfaces of leaves, can effectively

718   thicken the boundary layer around a leaf under some conditions (Benz and Martin, 2006).

719   Denser or longer trichomes that increase the boundary layer may decrease rates of heat

720   dissipation, evaporation, and gas exchange. Other configurations of trichomes may increase

721   turbulence in the boundary layer and increase gas exchange and heat dissipation

722   (Schreuder et al., 2001). Furthermore, any movement of the leaf relative to the movement

723   of the air or water may, in some cases, act to decrease the boundary layer and increase gas

724   exchange, evaporation, and heat dissipation (Roden and Pearcy, 1993). Each of these

725   parameters may be altered by the plant to improve the overall function of the leaf (Vogel,

726   2012).

727

728   *2. Effect of extreme winds and currents on broad leaves.* Plants use flexibility as one strategy

729   to reduce the drag acting upon them while simultaneously avoiding the violent oscillations

730   observed in many bluff structures subjected to strong winds. Vogel (1989) was the first to

731   provide quantitative data on drag reduction in plants. He found that single broad leaves

732   reconfigure at high flow velocities into cone shapes that reduce flutter and drag when

733   compared to paper cut-outs of similar shape and flexibility (**Figure 7A-B**). He later found

734   that a similar phenomenon occurs in water for herbaceous plants and proposed that this

735   mechanism increases survival rates in flash floods (Vogel, 2006). Subsequent experimental

736   studies on broad leaves and flowers also support rapid repositioning in response to strong

737   wind as a general mechanism to reduce drag (Niklas, 1992; Ennos, 1997; Etnier and Vogel,

738   2000), and simple mathematical models of a flexible beam immersed in a two-dimensional

739   flow exhibit similar behavior (Alben et al., 2002). At the larger scale, Vogel (1989) noticed

740   that leaf clusters and leaflets tend to reconfigure into larger drag reducing structures and

741   speculated that a similar phenomenon is also true for groups of branches (**Figure 7C**).

742

743   Passive reconfiguration of flexible structures in strong flows are prevalent throughout the

744   natural world (Vogel, 1994). One key feature that is important to drag and flutter reduction

745   is the shape of the reconfiguration. A wide range of broad leaves reconfigure into cone

746   shapes as both single leaves and clusters (Vogel, 1989; 2006). The branches and fronds of

747   flexible trees such as palms and willows appear to reconfigure into grouped conical shapes

748   in strong winds and hurricanes. The daffodil *Narcissus spp.* (Etnier, 2000) and sessile

749   marine organisms such as sea anemones (Koehl; 1977a; 1977b) reconfigure into cone

750   shapes through deformations of the petals or tentacles, respectively. Previous work has

751   shown that the cone shape is significantly more stable than other reconfigurations such as

752   U-shapes (Miller et al., 2012). The folding of the leaves and the interactions of surfaces are

753   critical to determining the shape of the reconfiguration and its stability (**Figure 7**).

754

755   Of course there are likely other strategies that may be used in place of, or in addition to,

756   conical reconfigurations. For example, Gosselin and de Langre (2011) showed how plants

757   such as pines may act as a poroelastic structures to reduce drag. As mentioned in the

758   previous section, there may also be situations in which flutter reduction is not desirable. In

759   these cases, conical reconfiguration may not be present at wind speeds relevant to cooling

760   and light transfer (0–4.5 m/s) when flutter could enhance these mechanisms.

761

762   Some plants also use active reconfiguration in flow. Plant responses to a mechanical

763   stimulus, known as a thigmonastic movement, are thought to be regulated by electrical

764   signal transduction. These electrical signals are action potentials that share many of the

765   same properties as those that occur in animals (Bose, 1926; 1928; Volkov, 2000; 2006). In

766   the case of the touch-me-not, *Mimosa pudica*, action potentials are generated as a result of

767   various stimuli such as touch, heat, electrical voltage, or wind. The electrical signal is

768   spread from the point of stimulus to the pinnae, petiole or stem (Bose, 1926; 1928) and will

769   result in the folding of the pinnae and the collapse of the petiole or stem. The action

770   potential generated from a weak stimulus will stop at the base of a single pinna so that the

771   leaflets from neighboring pinna remain unfolded (Volkov et al., 2010). These movements

772   are generated by motor organs called pulvini that consist of swollen joints.

773

774   Although it is now clear that flexible plants can reduce drag through reconfiguration, it is

775   not at all obvious how leaves and leaf clusters minimize the effects of vortex induced

776   oscillations. Previous work suggests that structural damping and stiffness nonlinearities

777   (Niklas, 1992; Bruchert et al., 2003; Miller, 2005) may reduce these destructive oscillations,

778   but this complex fluid-structure interaction problem remains relatively unexplored. The

779   study of vortex induced oscillations is also a significant area of research in structural

780   dynamics, and it seems likely that work in both fields may inform the other. In human-

781   made structures, improvements have been made by modifying the cross sectional shapes

782 and material properties of cables and other support structures to avoid self-excitation

783 through vortex shedding (Wilson, 2003; Tomita et al., 1988; Luongo and Piccardo, 1998).

784 Improving the stability of large flexible structures, such as banners and flags, in strong

785 winds remains problematic.

786

787 **IV. Milestones to accelerate the infusion of math into the plant sciences**

788

789 Despite the rich history of integration between mathematics and plant biology, sadly many

790 members of each discipline remain uninformed about the other. Mathematical modeling

791 has provided foundational insights into plant biology, but until recently, only a minority of

792 plant biologists have become versed in math, and even if quantitative, the expertise of plant

793 biologists is often limited to programming, computation, and statistics. Reciprocally,

794 mathematicians are often unaware of the tremendous empirical theory that plant biologists

795 have developed and lack the perspective of their biological training, limiting their

796 contributions to the plant sciences. Both mathematics and plant biology are timely

797 disciplines, especially in a world ever increasingly geared towards a quantitative mindset

798 and with dwindling natural resources. Both disciplines need to be emphasized more in

799 early education and come together through opportunities to interact, including workshops,

800 meetings, and funding opportunities. Both fields can immediately benefit from more open

801 approaches to science.

802

803 *A. Education*

804

805 The accuracy of mathematics in helping to understand the "reality" of biological processes

806 depends on how well the hypotheses supporting the definition of models fit with the real

807 world (May, 2004). A modeling perspective is itself one of the principal advantages that

808 integrating mathematics into plant science curriculums has to offer. Mathematics has been

809 likened to "biology's next microscope", because of the insights into an otherwise invisible

810 world it has to offer. Conversely, biology has been described as "mathematics' next

811 physics", stimulating novel mathematical approaches because of the hitherto unrealized

812 phenomena that it studies (Cohen, 2004). The scale of the needed interplay between
813 mathematics and plant biology is enormous and may lead to new science disciplines at the
814 interface of both: ranging from the cellular, tissue, organismal, and community levels to the
815 global; touching upon genetic, transcriptional, proteomic, metabolite, and morphological
816 data; studying the dynamic interactions of plants with the environment or the evolution of
817 new forms over geologic time; and spanning quantification, statistics, and mechanistic
818 mathematical models. It is important to remember that teaching this new synthesis is not
819 as simple as educating students with "naïve" minds. Rather, curriculums must be tailored in
820 such a way that students are competent to understand the skillsets they lack to tackle
821 specific problems or attain more nuanced understanding of phenomena (Gross, 2004). Or
822 to put this another way, students need to be taught how they can understand the
823 limitations of their own knowledge.
824
825 The new age of research is becoming increasingly interdisciplinary and undergraduate,
826 graduate, and post-graduate groups are actively trying to fix the archaic separation
827 between mathematics and biology skillsets. While many graduate programs have
828 specialization tracks under the umbrella of mathematics or biology-specific programs,
829 more frequently departments are forming specially designed graduate groups for
830 mathematical biology. Plant biology professors and researchers at the postdoctoral level
831 need to reach out to these programs at their schools. We encourage researchers at the
832 principal investigator and post-doctoral level to become more active in departments with
833 their university that house young researchers being trained in these programs.
834 Recruitment of students from these programs takes advantage of the inherent desire to
835 pursue mathematics and biology-focused research. Since one of the main problems
836 separating fields is each field learning the jargon of the other (Oleana, 2014), guest lectures
837 in courses within these new graduate programs would introduce how each field
838 approaches research. The most powerful way to connect mathematics and plant biology is
839 to first acknowledge our presence to this emerging pool of young researchers.
840
841 *B. Citizen science, the maker movement, and public outreach*

842

843   Citizen science, which is a method to make the general public aware of scientific problems

844   and employ their help in solving them[2], is an ideal platform to initiate a synthesis between

845   plant biology and mathematics because of the relatively low cost and accessibility of each

846   field. A variety of research fields have successfully implemented citizen science and

847   gamification approaches, including astronomy[3], animal biology[4], and cell biology[5].

848

849   Arguably, using citizen science to collect plant morphological diversity has already been

850   achieved, but has yet to be analyzed fully. In total, it is estimated that the herbaria of the

851   world possess greater than 207 million voucher specimens[6], representing the diverse

852   lineages of land plants (and often algae as well) collected over their respective

853   biogeographies over a timespan of centuries. Both herbaria and living collections of botanic

854   gardens are dynamic. Although plant collecting is in decline, it is an active discipline and

855   voucher submissions to herbaria should be encouraged (Prather et al., 2004). Botanic

856   gardens not only possess living collections, but more than ever they spearhead bio-

857   conservation efforts to ensure plant diversity exists in the future to be studied (Schatz,

858   2002). Digital documentation of the millions of vouchers held by the world's botanic

859   gardens is actively underway, allowing for researchers and citizens alike to access and

860   study for themselves the wealth of plant diversity across the globe and centuries (Smith et

861   al., 2003; Corney et al., 2012; Ryan, 2013).

862

863   Beyond herbaria and botanic gardens, the developmental changes in plants responding to

864   environmental variability and microclimatic changes over the course of a growing season

865   can be analyzed by studying phenology. Citizen science projects such as the USA National

866   Phenology Network[7] or Earthwatch[8] and associated programs such as My Tree Tracker[9]

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[2] For example, see the White Paper on Citizen Science for Europe, http://www.socientize.eu/sites/default/files/white-paper_0.pdf (retrieved May 29, 2016)

[3] Galaxy Zoo, http://www.galaxyzoo.org/ (retrieved May 29, 2016)

[4] Wisconsin Wildlife Watch, https://www.zooniverse.org/projects/zooniverse/wisconsin-wildlife-watch (retrieved May 29, 2016)

[5] Cell slider, http://www.cellslider.net/#/ (retrieved May 29, 2016)

[6] List of herbaria, https://en.wikipedia.org/wiki/List_of_herbaria (retrieved May 29, 2016)

[7] https://www.usanpn.org/# (retrieved May 29, 2016)

[8] http://earthwatch.org/scientific-research/special-initiatives/urban-resiliency (retrieved May 29, 2016)

867   document populations and individual plants over seasons and years, providing a

868   distributed, decentralized network of scientific measurements. Integrating phenological

869   data—essentially plant morphological measurements over time—with weather modeling

870   and satellite data promises to provide meaningful predictions of the ecological

871   consequences of climate change.

872

873   Citizen science is also enabled by low-cost, specialized equipment. Whether programing a

874   camera to automatically take pictures at specific times or automating a watering schedule

875   for a garden, the maker movement—a do-it-yourself cultural phenomenon that intersects

876   with hacker culture—focuses on building custom, programmable hardware, whether via

877   electronics, robotics, 3D-printing, or time-honored skills such as metal- and woodworking.

878   The focus on programming is especially relevant for integrating mathematical approaches

879   with plant science experiments. Originally built for education, single-board computers like

880   Raspberry Pi, Hummingboard, or Cubieboard, are microcomputers that are widely adopted

881   by educators, hobbyists and researchers that have enabled the maker culture. Single-board

882   computers are used for diverse tasks, from logging and distributing data from

883   environmental sensors to playing MineCraft. The low-cost of single-board computers

884   computers ($5.00- $35.00, depending on model) makes tinkering more permissive for a

885   greater population of citizen science than previously feasible. In classrooms,

886   microcomputers can be used simply as a computer station, or incorporated into lesson

887   plans using a variety of available teaching resources[10]. Single-board computers enable the

888   early cultivation of interest in computational science through educational modules that can

889   grow in complexity with student ability.

890

891   *C. Workshops and funding opportunities*

892

893   Simply bringing mathematicians and plant biologists together to interact, to learn about

894   tools, approaches, and opportunities in each discipline that researchers may not be aware

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9 http://www.mytreetracker.org/cwis438/websites/MyTreeTracker/About.php?WebSiteID=23 (retrieved May 29, 2016)

10 https://www.raspberrypi.org/resources/ (retrieved May 29, 2016)

895   of, is a major barrier preventing the full integration of these two disciplines. This white

896   paper itself is a testament to the power of bringing mathematicians and biologists together,

897   resulting from a National Institute for Mathematical and Biological Synthesis (NIMBioS)

898   workshop titled "Morphological Plant Modeling: Unleashing Geometric and Topologic

899   Potential within the Plant Sciences", held at the University of Tennessee, Knoxville

900   September 2-4, 2015[11] (**Figure 8**). Other mathematical institutes such as the Mathematical

901   Biology Institute (MBI) at Ohio State University[12], the Statistical and Applied Mathematical

902   Sciences Institute (SAMSI) in Research Triangle Park[13], the Institute for Mathematics and

903   Its Applications at University of Minnesota[14], and the Centre for Plant Integrative Biology at

904   the University of Nottingham[15] have also hosted workshops for mathematical and

905   quantitative biologists from the undergraduate student to the faculty level.

906   There are efforts to unite biologists and mathematics through initiatives brought forth

907   from The National Science Foundation, including Mathematical Biology Programs[16] and the

908   Joint DMS/NIGMS Initiative to Support Research at the Interface of the Biological and

909   Mathematical Sciences[17] (DMS/NIGMS). Outside of the Mathematics and Life Sciences

910   Divisions, the Division of Physics houses a program on the Physics of Living Systems.

911   Societies such as The Society for Mathematical Biology and the Society for Industrial and

912   Applied Mathematics (SIAM) Life Science Activity Group[18] are focused on the dissemination

913   of research at the intersection of math and biology, creating many opportunities to present

914   research and provide funding. One of the problems in the creation of an active intersection

915   of mathematicians and plant biologists is that scientists often describe themselves as one

916   or the other. Ways to actively unite these fields would be to incorporate cross topic

917   sessions, inviting scientists from mathematics and plant biology to speak in an effort to

918   inform each discipline and initiatives for cross-trained scientists that bridge both

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[11] http://www.nimbios.org/workshops/WS_plantmorph (retrieved May 29, 2016)

[12] https://mbi.osu.edu/ (retrieved May 29, 2016)

[13] http://www.samsi.info/ (retrieved May 29, 2016)

[14] https://www.ima.umn.edu/ (retrieved May 29, 2016)

[15] https://www.cpib.ac.uk/outreach/cpib-summer-school/ (retrieved May 29, 2016)

[16] https://www.nsf.gov/funding/pgm_summ.jsp?pims_id=5690 (retrieved May 29, 2016)

[17] http://www.nsf.gov/funding/pgm_summ.jsp?pims_id=5300&org=DMS (retrieved May 29, 2016)

[18] https://www.siam.org/activity/life-sciences/ (retrieved May 29, 2016)

919   disciplines. Funding initiatives to help promote these sessions and provide travel stipends

920   to attend conferences for this specific purpose could greatly enhance collaboration of the

921   plant sciences and mathematics. Another possibility is funding for a joint meeting where

922   mathematicians and plant biologists are paired together to work on a dataset together. This

923   has worked well in computer science contexts and reaches a broad audience[19].

924

925   *D. Open Science*

926

927   Ultimately, mathematicians and plant biology must unite at the level of jointly collecting

928   data, analyzing it, and doing science together. Sharing data and code is perhaps the easiest

929   way to unite these two disciplines, reducing redundancy and providing benchmarked

930   datasets from which radically different interpretation of plant morphology can be

931   compared. Great strides have already been made in creating an open science environment,

932   but much work remains to increase the number of accessible datasets and code and to use

933   such open access to facilitate communication between plant biologists and mathematicians

934   that might not otherwise interact.

935

936   A number of platforms provide open, public access to datasets, figures, and code that can be

937   shared, including Dryad[20], Dataverse[21], and Figshare[22]. These services are integrated to

938   journals and the publication of articles to varying degrees, and have different limitations on

939   the amount of data that can be stored and contrasting financial models. Beyond the ability

940   to share data is the question of open data formats and accessibility. For example, in remote

941   sensing research it is unfortunately common that proprietary data formats are shared,

942   which prevents their use without specific software. ASCII format is universal, but it has

943   issues with performance and storage size that can become an obstacle with large datasets.

944   Open binary data formats are needed to ensure data availability and accessibility in

945   interdisciplinary research. Hierarchical Data Formats (HDF) 4 and 5 are a good example of

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[19] Hack the Dinos, American Museum of Natural History, http://www.amnh.org/calendar/hack-the-dinos (retrieved May 29, 2016)

[20] http://datadryad.org/ (retrieved May 29, 2016)

[21] http://dataverse.org/ (retrieved May 29, 2016)

[22] https://figshare.com/ (retrieved May 29, 2016)

946   this[23]. For point cloud data describing 3D structures, ASPRS .las[24] and .e57 (Huber, 2011)

947   formats have been designed with distributability in mind. Phenopackets is a new format

948   that aims to handle a wide range of phenotypic and environmental data in big datasets[25].

949

950   Beyond datasets, making code openly available, citable and user friendly is a means to

951   share methods to analyze data. Places to easily share code include web-based version

952   controlled platforms like Bitbucket[26] or Github[27] and software repositories like

953   Sourceforge[28]. Most of these platforms allow the easy generation of DOI associated with the

954   dataset or code. For example, with Github you can automatically obtain a DOI for your

955   repository through Zenodo[29]. Initiatives like Depsy[30] are trying to put into value the work

956   of research software developers that most of the times goes unnoticed. Also, these

957   repositories allow for distributed version control, which makes it easier for multiple

958   developers to work and collaborate on the same code.

959

960   The most powerful open science tools bring datasets, code, and analysis together in unique,

961   accessible ways. Meta-analysis datasets provide curated resources where numerous

962   published and unpublished datasets related to a specific problem (or many problems) can

963   be accessed by researchers[31]. The portability and interoperability of disparate datasets

964   between software tools can be overcome with standardized formats. The crucial element is

965   that such formats are somehow reflective of universal plant morphological features,

966   bridging the gap between programming languages and biology, as seen in the Root System

967   Mark-up Language (Lobet et al., 2015) and OpenAlea (Pradal et al., 2008). Bisque is a

968   versatile platform to store, organize, and analyze image data, providing simultaneously

969   open access to data and analyses as well as the requisite computation (Kvilekval et al.,

23 http://www.hdfgroup.org/ (retrieved May 29, 2016)

24 http://www.asprs.org/a/society/committees/standards/LAS_1_4_r13.pdf (retrieved May 29, 2016)

25 https://github.com/phenopackets (retrieved September 24, 2016)

26 https://bitbucket.org/ (retrieved May 29, 2016)

27 https://github.com/ (retrieved May 29, 2016)

28 https://sourceforge.net/ (retrieved May 29, 2016)

29 http://zenodo.org/ (retrieved May 29, 2016)

30 http://depsy.org/ (retrieved May 29, 2016)

31 BAAD: a Biomass And Allometry Database for woody plants, https://github.com/dfalster/baad (retrieved May 29, 2016)

970   2010). CyVerse[32] (formerly iPlant) is a similar platform, on which academic users get 100

971   GB storage for free and can create analysis pipelines that can be shared and reused (Goff et

972   al., 2011). For example, DIRT[33] is an automatic, high throughput computing platform

973   (Bucksch et al., 2014c; Das et al., 2015) that the public can use hosted on CyVerse using the

974   Texas Advanced Computing Center[34] (TACC) resources at UT Austin that robustly extracts

975   root traits from digital images.

976

977   Regardless of the specific purpose of an open science dataset, analysis, or platform, the key

978   point is that the data be accessible and generalizable to the widest audience as possible. For

979   both plant biologists and mathematicians, formatting open data can take inordinate

980   amounts of time. Effective data sharing can be accomplished by making raw data available

981   as well as processed data, and documenting the steps that were used to arrived at the

982   processed data[35]. Data should always be accompanied by a codebook, metadata describing

983   each variable used, and when at all possible, be "tidy data" (that is, a data frame with

984   columns for each variable and rows for each observation that is immediately accessible for

985   data analysis and visualization, Wickham, 2014). Beyond datasets, there are

986   recommendations for organizing file structures and documenting progress (Noble, 2009)

987   as well as programming itself (Wilson et al., 2014). Further, the discoverability of biological

988   data to mathematicians is key: ideally public data would be accompanied by information

989   including outstanding questions that mathematics may be able to solve or reasons why the

990   dataset should be explored in alternative ways to how it has already been analyzed. Well-

991   formatted data can be made public easily enough, but experimental metadata alone may

992   not be sufficient to generate further interest. While the above are attempts to make

993   biological data more accessible to mathematicians, the reciprocal direction of sharing is

994   less well represented. In fact, evidence suggests that using equations (one of the primary

995   means by which mathematicians share their work) impedes communication with biologists

32 http://www.cyverse.org/ (retrieved August 20, 2016)

33 http://dirt.iplantcollaborative.org/ (retrieved August 20, 2016)

34 https://www.tacc.utexas.edu/ (retrieved August 20, 2016)

35 How to share data with a statistician, https://github.com/jtleek/datasharing (retrieved May 29, 2016)

996   (Fawcett and Higginson, 2012). Providing context and narrative, in addition to the

997   interactions between plant biology and mathematics outlined in this paper, should help the

998   accessibility of math and its contingent equations to biologists.

999

1000 **V. Conclusion: Unleashing geometric and topologic potential within the plant**

1001 **sciences**

1002

1003 The plant form is inherently architectural, from the shapes of leaves to the hierarchies in

1004 the branching patterns of shoots and roots. Plant morphology has served as an inspiration

1005 to mathematicians, who innovated new methods to quantify and model the plant form as a

1006 result of evolutionary, developmental and environmental responses (**Figures 1-2**). Plant

1007 morphology is an unresolved mystery to plant biologists, who seek to understand the

1008 molecular mechanisms by which such predetermined, yet seemingly endless variations of

1009 organizational patterns emerge.

1010

1011 Never have the resources to study plant morphology been more plentiful. Burgeoning

1012 imaging technologies—innovative confocal microscopy, X-ray imaging, MRI, radar,

1013 terrestrial laser scanning, among many others—have made detailed 3D models of plants

1014 feasible (**Figures 3-4**). Imaging across time and scales looks behind the curtains from the

1015 sub-cellular scale to the entirety of the earth surface. The mathematical methods to model

1016 plant morphology, from Turing's reaction-diffusion processes to innovative shape

1017 descriptors, quantify the plant form to be used in a wide-variety of biological applications

1018 (**Figure 6**).

1019

1020 Yet, all models are idealizations of reality and seemingly simple problems remain

1021 unresolved in the study of plant morphology. Integrating observations at different scales is

1022 a persistent challenge. We discussed for example that the ability to model the fluttering of

1023 leaves within a tree canopy remains out of reach (**Figure 7**). Interest in the hidden half of

1024 plant morphology—the root system—has only recently seen a renaissance with

1025 technologies capable of penetrating soil and visualizing roots *in situ* (**Figure 5**). Not only is

35

1026 describing phenotype problematic, but linking it to a genetic and (especially in plants)

1027 environmental basis, within the context of plant evolution, is still an unattained goal. Never

1028 have the answers to these questions been more vital and timely. Modifying plant

1029 architecture through molecular biology and breeding is key to develop agricultural outputs

1030 and sustainability. Monitoring the morphology of plants in response to a shifting

1031 environment is necessary to model global responses to climate change. Cross-disciplinary

1032 training of scientists, citizen science, and open science are all necessary components to

1033 address these needs (**Figure 8**). Unleashing the potential of geometric and topological

1034 approaches in the plant sciences promises to transform our understanding of both plants

1035 and mathematics, and to meet the challenges posed by a future with dwindling and

1036 uncertain natural resources.

1037

1038 **Funding**

1039

1040 This manuscript resulted from the National Institute for Mathematical and Biological

1041 Synthesis (NIMBioS) workshop on plant morphological models at the University of

1042 Tennessee, Knoxville, held September 2-4, 2015, resulting from a proposal submitted by

1043 Alex Bucksch and Dan Chitwood.

1044

1045 **Acknowledgements**

1046

1047 The authors are grateful to the National Institute for Mathematical and Biological Synthesis

1048 (NIMBioS, University of Tennessee, Knoxville) for hosting and funding the workshop

1049 "Morphological Plant Modeling: Unleashing geometric and topological potential within the

1050 plant sciences" that inspired this manuscript.

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Figure Legends

Figure 1: Plant morphology from the perspective of biology. A) Adapted from Kaplan (2001). Plant morphology interfaces with all disciplines of plant biology—plant physiology, plant genetics, plant systematics, and plant ecology—influenced by both developmental and evolutionary forces. **B)** Artistic rendering of Goethe's *Urpflanze* concept (Goethe, 1837; Friedman and Diggle, 2011), demonstrating the importance of development in modulating the plant form. Goethe viewed the iterative development of plants as a transformation of lateral organ identities, from juvenile to adult leaves to reproductive structures. **C)** Evolutionary and ecological effects also shape plant morphology. Alexander von Humboldt's distribution of plants across the elevations of Chimborazo (Ecuador)

2121   demonstrates the effects of evolution and environment on plant morphology (Black and

2122   Black, 1839).

2123

2124   **Figure 2: Plant morphology from the perspective of mathematics. A)** The topological

2125   complexity of plants requires a mathematical framework to describe plant morphology.

2126   Shown is the top of a maize crown root 42 days after planting. Color represents root

2127   diameter. Image provided by Jonathan Lynch and Johannes Postma (Pennsylvania State

2128   University). **B)** Persistent homology deforms a given plant morphology using functions to

2129   define self-similarity in a structure. In this example, a color map of geodesic distance

2130   functions to the ground level of a tree (blue indicates smaller values near the proximal base

2131   and red larger values at the distal tips) is recorded as **C)** an H_0 barcode, in which "births"

2132   and "deaths" of connected components as a function of the distance function are recorded.

2133   Images provided by Mao Li (Danforth Plant Science Center).

2134

2135   **Figure 3: Imaging techniques to capture plant morphology. A)** Confocal sections of an

2136   Arabidopsis root. The upper panel shows a new lateral root primordium at an early stage of

2137   development (highlighted in yellow). At regular intervals new roots branch from the

2138   primary root. The lower panel shows the primary root meristem and the stem cell niche

2139   (highlighted in yellow) from which all cells derive. Scale bars: 100µm. Images provided by

2140   Alexis Maizel (Heidelberg University). **B)** Computational tomographic (CT) x-ray sections

2141   through a reconstructed maize ear (left and middle) and kernel (right). Images provided by

2142   Chris Topp (Donald Danforth Plant Science Center). **C)** Laser ablation tomography (LAT)

2143   image of a nodal root from a mature, field-grown maize plant, with color segmentation

2144   showing definition of cortical cells, aeraenchyma lacunae, and metaxylem vessels. Image

2145   provided by Jennifer Yang (Penn State).

2146

2147   **Figure 4: Terrestrial laser scanning creates a point cloud reconstruction of a Finnish**

2148   **forest. A)** Structure of a boreal forest site in Finland as seen with airborne (ALS) and

2149   terrestrial (TLS) laser scanning point clouds. The red (ground) and green (above ground)

2150   points are obtained from National Land Survey of Finland national ALS point clouds that

2151   cover hundreds of thousands of square kilometers with about 1 point per square meter

2152   resolution. The blue and magenta point clouds are results of two individual TLS

2153   measurements and have over 20 million points each within an area of about 500 square

2154   meters. TLS point density varies with range but can be thousands of points per square

2155   meter up to tens of meters away from the scanner position. **B)** An excerpt from a single TLS

2156   point cloud (blue). The TLS point cloud is so dense that individual tree point clouds

2157   (orange) and parts from them (yellow) can be selected for detailed analysis. **C)** A detail

2158   from a single TLS point cloud. Individual branches (yellow) 20 meters above ground can be

2159   inspected from the point cloud with centimeter level resolution to estimate their length

2160   and thickness. Images provided by Eetu Puttonen (Finnish Geospatial Research Institute in

2161   the National Land Survey of Finland). ALS data was obtained from the National Land

2162   Survey of Finland Topographic Database, 08/2012 (National Land Survey of Finland open

2163   data licence, version 1.0).

2164

2165   **Figure 5: The environmental basis of plant morphology.** Root system architecture of

2166   Arabidopsis Col-0 plants expressing ProUBQ10:LUC2o growing in **A)** control and **B)** water-

2167   deficient conditions using the GLO-Roots system (Rellán-Álvarez et al., 2015). Images

2168   provided by Ruben Rellán-Álvarez (Laboratorio Nacional de Genómica para la

2169   Biodiversidad, CINVESTAV).

2170

2171   **Figure 6: Integration of tissue growth and reaction-diffusion models. A)** Vertex model

2172   of cellular layers (Prusinjiewicz and Lindenmayer, 2012). K, l_a, and l_0 are the spring

2173   constant, current length, and rest length for wall a. K_P is a constant and S_A is the size of cell

2174 A. Δt is time step. Shown is a simulation of cell network growth. **B)** Reaction diffusion

2175   model of the shoot apical meristem for WUSCHEL and CLAVATA interactions (Fujita et al.,

2176   2011). u=WUS, v=CLV, i=cell index, Φ is a sigmoid function. E, B, A_S, A_d, C, D, u_m, D_u, D_v are

2177   positive constants. Shown are the distributions of WUS and CLV levels within a dynamic

2178   cell network. Images provided by Dorjsuren Battogtokh (Virginia Tech).
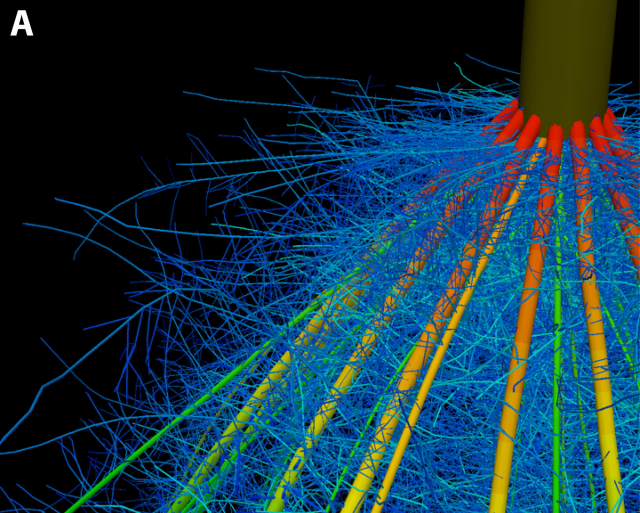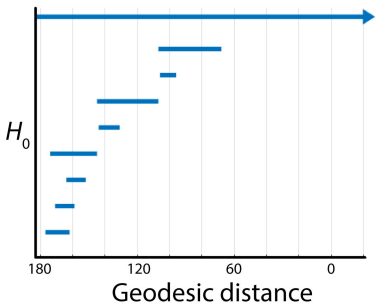
2179

2180   **Figure 7: Modeling the interaction between plant morphology and fluid dynamics. A)**

2181   3D immersed boundary simulations of flow past a flexible rectangular sheet (left) and disk

2182   with a cut from the center to edge (right). Both structures are attached to a flexible petiole,

2183   and the flow is from left to right. The contours show the magnitude of vorticity (the

2184   rotation in the air). The circular disk reconfigures into a cone shape, similar to many broad

2185   leaves. **B)** Reconfiguration of tulip poplar leaves in 3 m/s (left) and 15 m/s flow (right). The

2186   leaves typically flutter at lower wind speeds and reconfigure into stable cones at high wind

2187   speeds. **C)** A cluster of redbud leaves in wind moving from right to left. The wind speed is

2188   increased from 3 m/s (left) to 6 m/s (middle) and 12 m/s (right). Note that the entire

2189   cluster reconfigures into a cone shape. This is different from the case of tulip poplars and

2190   maples where each leaf individually reconfigures into a conic shape. Images provided by

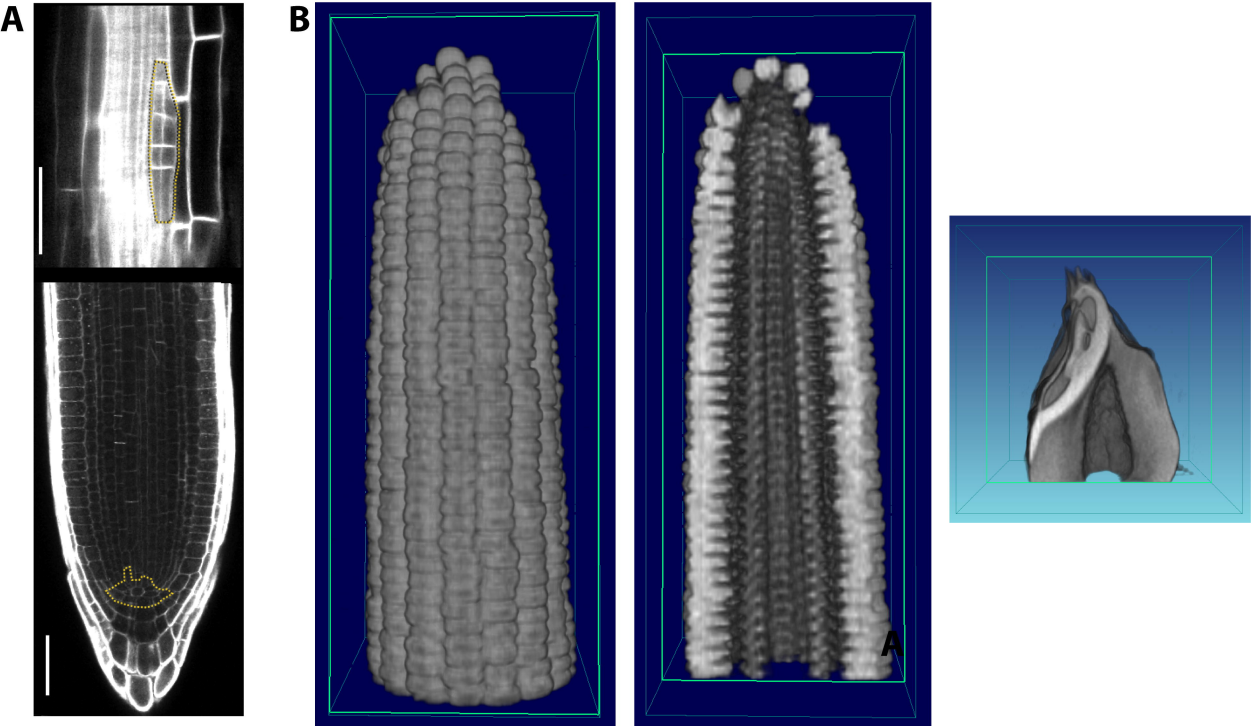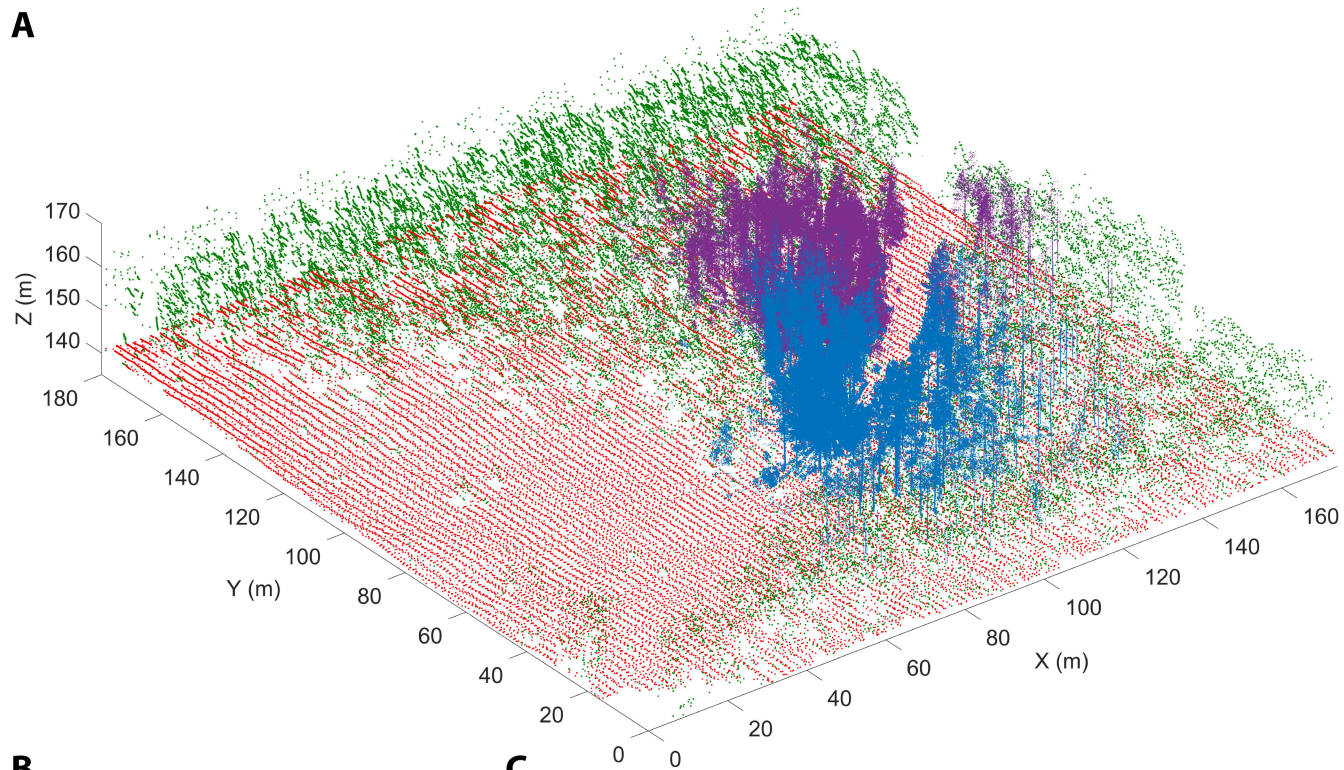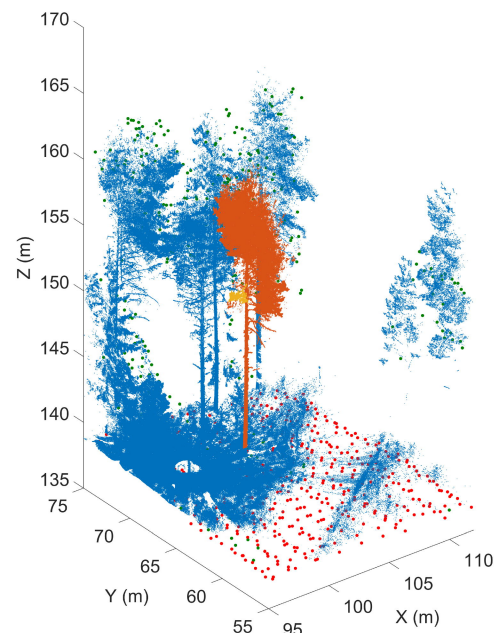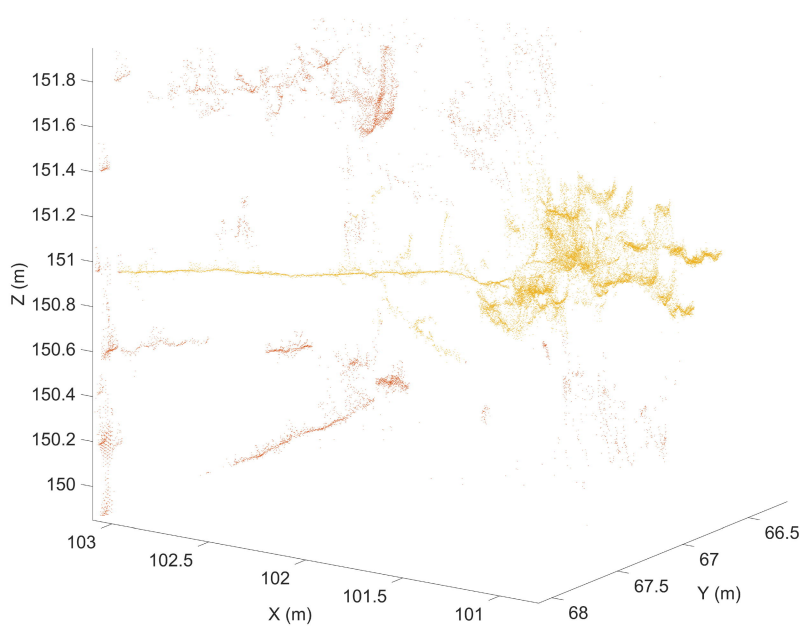2191   Laura Miller (University of North Carolina, Chapel Hill).

2192

2193   **Figure 8: Milestones to accelerate the infusion of math into the plant sciences.** Group

2194   photo of the authors from the National Institute for Mathematical and Biological Synthesis

2195   (NIMBioS) meeting on plant morphological models (University of Tennessee, Knoxville,

2196   September 2-4, 2015) that inspired this manuscript. Workshops such as these, bringing

2197   mathematicians and plant biologists together, will be necessary to create a new synthesis
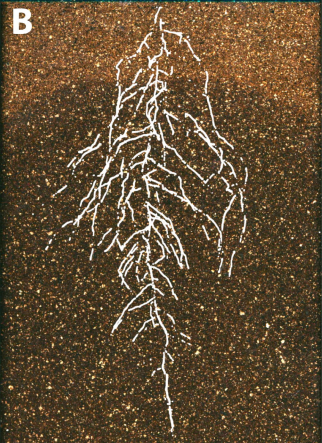
2198   of plant morphology.

A

PLANT
PHYSIOLOGY
emphasis on
function

PLANT
SYSTEMATICS
emphasis on
homologies

DEVELOPMENT

PLANT
MORPHOLOGY
emphasis on
analogies

EVOLUTION

PLANT
GENETICS
emphasis on
heredity

PLANT
ECOLOGY
includes
vegetation types

B

C

HUMBOLDT'S DISTRIBUTION OF PLANTS IN EQUINOCTIAL AMERICA, ACCORDING TO ELEVATION ABOVE THE LEVEL OF THE SEA.

A

B

C

H_0

Geodesic distance

180   120   60   0

A

Total force on a vertex i

$$\vec{F}_i^T = \sum_{n \in a,b,c} \vec{F}_s^n + \frac{1}{2} \sum_{w \in a,b,c;\ m \in A,B,C} \vec{P}_w^m$$
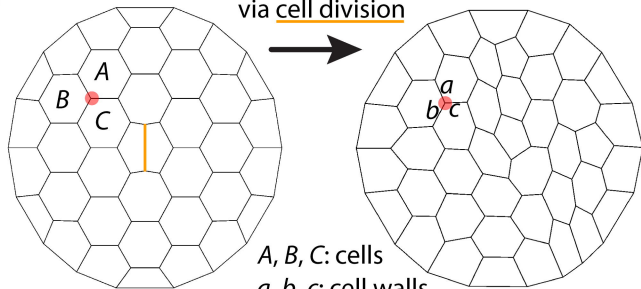
Spring force on a wall a

$$\vec{F}_s^a = -K(l_a - l_0)$$

Turgor pressur on a from A

$$\vec{P}_A^a = \frac{K_P}{S_A}; \vec{l}_a \perp \vec{P}_A^a; \vec{l}_a \perp \vec{P}_B^a$$

Dynamics of vertex i

$$\vec{x}_i(t + \Delta t) = \vec{x}_i(t) + \vec{F}_i^T \Delta t$$

Tissue growth via <u>cell division</u>

A, B, C: cells
a, b, c: cell walls
● vertex: i

B
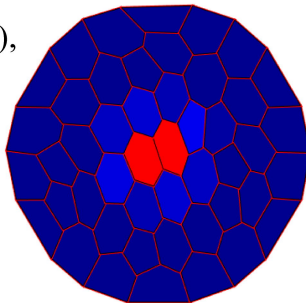
$$\frac{\partial u_i}{\partial t} = \Phi(E - B \cdot v_i + A_S \cdot u_i) - A_d \cdot u_i + D_u \sum_{j=neighbors}(u_j - u_i),$$

$$\frac{\partial v_i}{\partial t} = C \cdot u_i - D \cdot v_i + D_v \sum_{j=neighbors}(v_j - v_i),$$

$$\Phi(x) = \frac{A_d u_m}{2}\left(1 + \frac{2x/(A_d u_m) - 1}{\sqrt[n]{1 + |\ 2x/(A_d u_m) - 1\ |^n}}\right).$$

WUSCHEL

CLAVATA

A

B

3 m/s   15 m/s

C