Rethinking gene regulatory networks in light of alternative splicing, intrinsically disordered protein domains, and post-translational modifications

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Models for genetic regulation and cell fate specification characteristically assume that gene regulatory networks (GRNs) are essentially deterministic and exhibit multiple stable states specifying alternative, but pre-figured cell fates. Mounting evidence shows, however, that most eukaryotic precursor RNAs undergo alternative splicing (AS) and that the majority of transcription factors contain intrinsically disordered protein (IDP) domains whose functionalities are context dependent as well as subject to post-translational modification (PTM). Consequently, many transcription factors do not have fixed cis-acting regulatory targets, and developmental determination by GRNs alone is untenable. Modeling these phenomena requires a multi-scale approach to explain how GRNs operationally interact with the intra- and intercellular environments. Evidence shows that AS, IDP, and PTM complicate gene expression and act synergistically to facilitate and promote time- and cell-specific protein modifications involved in cell signaling and cell fate specification and thereby disrupt a strict deterministic GRN-phenotype mapping. The combined effects of AS, IDP, and PTM give proteomes physiological plasticity, adaptive responsiveness, and developmental versatility without inefficiently expanding genome size. They also help us understand how protein functionalities can undergo major evolutionary changes by buffering mutational consequences.

Keywords: cell fate specification, combinatorial transcription regulation, eukaryotes, evolution, development, gene regulatory networks, multi-stable dynamical systems, protein structure

INTRODUCTION

A fundamental assumption of contemporary developmental biology is that gene regulatory networks (GRNs, herein defined as circuits of interacting transcription factors and their cis-acting regulatory elements) are primary mechanisms controlling development. According to this assumption, at any time, the relative levels of transcription factors in an extended network determine the progress of development by regulating downstream genes (Carroll et al., 2004; Davidson and Erwin, 2006). This conception of gene control in multicellular organisms, which was formulated in several related versions a half century ago (Britten and Davidson, 1969; Kauffman, 1969; Britten, 1982; Davidson, 1982), proposes that GRNs are deterministic dynamical systems exhibiting multiple stable states. The theoretical foundations of this framework can be traced to studies of the bi-stable gene regulatory switch between the lytic and lysogenic states of the lambda phage in Escherichia coli (see Ptashne, 2004), and have been generalized and applied to the larger and more elaborate GRNs of eukaryotes in the form of models ranging from discrete Boolean networks to continuous systems of ordinary differential equations (Glass and Kauffman, 1973; Kauffman, 1974; Savageau, 1974; Glass, 1975; Lauffenburger, 2000; Jaeger and Monk, 2014). A commonality among these models describing cell differentiation is the assumption that gene products (i.e., proteins, particularly transcription factors) have specific identities and connectivity relationships to one another in the GRNs in which they function (Hasty and Collins, 2001; Forgacs and Newman, 2005). According to this view, variation in the outcome of the function of GRNs (e.g., alternative cell types) arises from nonlinearities and stochastic effects to which such complex, deterministic systems are subject.

This paradigm has been extended to other gene expression mechanisms that have been characterized since the GRN dynamics model was first proposed. Among these mechanisms is the alternative splicing (AS) of pre-mRNA exons and introns to assemble different proteins, a process that permits variation in the functionalities of subsets defining components of GRNs (e.g., transcription factors) at the level of RNA processing. Although
the mixing and matching of basic system components has no direct counterpart in a truly deterministic GRN model, the factors controlling AS can be viewed as having well defined functionalities, since the associated GRN dynamics permit different cell fates in a combinatorial deterministically prescribed manner. Likewise, the modulatory effects of microRNAs, riboswitches, and the enzymes that mediate post-translational modifications (and that can silence genes) can be viewed as adding a complicating, yet still deterministic set of regulatory mechanisms.

Here, we propose that this perspective must also contend with evidence that the majority of eukaryotic transcription factors contain intrinsically disordered protein (IDP) domains (Liu et al., 2006; Dunker et al., 2014) that comprise almost two-thirds of their sequences (Ward et al., 2004; Minezaki et al., 2006; Dunker et al., 2014) that comprise almost two-thirds of their sequences (Ward et al., 2004; Minezaki et al., 2006; Dunker et al., 2014) that comprise almost two-thirds of their sequences (Ward et al., 2004; Minezaki et al., 2006; Dunker et al., 2014). The effects of splicing factors are often position-dependent (i.e., U1, U2, U4–U6) and a range of assorted protein factors (Figure 1). Splicing is regulated by trans-acting repressor proteins within an RNA–protein complex, the spliceosome, which is canonically composed of five small nuclear RNAs (i.e., U1, U2, U4–U6) and a range of assorted protein factors (Figure 1). Splicing is regulated by trans-acting repressor-activator proteins and their corresponding cis-acting regulatory elements (Catania and Lynch, 2008), we suggest that the connection between cell type number and AS (given the association with IDP and PTM) is an inherent one that promoted occupation of new niches (see below “AS-IDP-PTM Phylogenetic Patterns”).

Alternative splicing produces protein isoforms from the same precursor mRNA by retaining or excluding different exons to achieve differential translation. First observed in the infectious adenovirus cycle (Berget et al., 1977; Chow et al., 1977) and subsequently in the transcripts of normal, endogenous genes (Leff and Rosenfeld, 1986), AS occurs in all eukaryotic lineages (Black, 2003) and becomes more prevalent as complexity, estimated by the number of different cell types, increases (Chen et al., 2014a). Although a number of scenarios have been advanced for the origin of AS, including a role in enabling the cell to filter out aberrant transcripts (Catania and Lynch, 2008), we suggest that the connection between cell type number and AS (given the association with IDP and PTM) is an inherent one that promoted occupation of new niches (see below “AS-IDP-PTM Phylogenetic Patterns”).

Five basic types of alternative splicing exist: alternative 3′ acceptor site, alternative 5′ donor splice site, intron retention, mutually exclusive exon splicing, and exon skipping (Black, 2003). The last is the most frequent. Regulation and selection of the splice sites are performed by trans-acting splicing activators and repressor proteins within an RNA–protein complex, the spliceosome, which is canonically composed of five small nuclear RNAs (i.e., U1, U2, U4–U6) and a range of assorted protein factors (Figure 1). Splicing is regulated by trans-acting repressor-activator proteins and their corresponding cis-acting regulatory elements and enhancers on the pre-mRNA (Matera and Wang, 2014). The effects of splicing factors are often position-dependent (Barash et al., 2010). A splicing factor that functions as an activator when bound to an intronic enhancer element may function as a repressor when bound to its splicing element in the context of an exon (Lim et al., 2011).

The secondary structure of the pre-mRNA transcript also determines which exons and introns will be spliced, e.g., by bringing together splicing elements or by masking a sequence that would otherwise serve as a binding element for a splicing factor. Consequently, activators, repressors, and secondary pre-mRNA structure constitute a splicing “code” that defines the protein isoforms produced under different cellular conditions. Additionally, the elements within this code function interdependently in ways
FIGURE 1 | Schematic of the structure and operation of a spliceosome to remove an intron flanked by exons on a pre-mRNA. (A) The splicing process is guided by a highly conserved 5′ splice site GU sequence, an A branch site near a pyrimidine-rich region, and a 3′ splice site AG sequence. The spliceosome protein complex contains RNA and protein components (i.e., small nuclear ribonucleoprotein or snRNPs, designated U1, U2, U4–U6) that recognize and bind to the pre-mRNA conserved sequences in a stepwise process. (B) The process begins with U1, which binds to the 5′ splice site, and U2, which binds to the A branch site. (C) U4, U6, and U5 subsequently bind the pre-mRNA transcript forming the mature spliceosome complex that configures the intron into loop bringing the 5′ and 3′ splice sites converge. (D) The mature spliceosome splices the 5′ first and the 5′ GU end second, creates a lariat by connecting the 5′ end to the A branch site. The U1 and U4 snRNPs are released and the 3′ splice site is cleaved. (E) The intron, U3 and the U5–U6 ensemble are released, and exons are attached. The intron will degrade and the snRNPs will be reused. (F) Schematic of alternative splicing of a pre-mRNA with four exons that can yield five different proteins.

that are context dependent, both intracellularly and extracellularly (Talavera et al., 2013). For example, cis-acting regulatory silencers and enhancers are influenced by the presence and relative position of other RNA sequence features, and the trans-acting context is affected by intracellular conditions that are in turn influenced by external conditions (Chen et al., 1999; Wang et al., 2004; Matlin et al., 2005) and other RNA sequence features. Furthermore, some cis-acting elements may reverse the effects on splicing if specific proteins are expressed in the cell (Boutz et al., 2007; Spellman et al., 2007). Indeed, the number of factors influencing AS is significantly large. A recent, empirically successful computer model for predicting the number and type of spliceoform in different human tissues depends on nearly 1400 exonic and intronic features and identifies more than 20,000 unique single-nucleotide variants that likely affect splicing (Xiong et al., 2015).

AS is adaptive and highly conserved. There is strong selection against mutations that alter splicing (Fairbrother et al., 2004; Ke et al., 2008). For example, Chang et al. (2014) report a conserved AS pattern for heat shock transcription factors in the moss Physcomitrella patens and the flowering plant Arabidopsis thaliana and show that the AS mechanism for heat regulation among land plants is an ancestral condition. Using mRNA sequence data, Pan et al. (2008) report that transcripts from ≈95% of human multi-exon genes undergo alternative splicing and that ≈100,000 intermediate to high abundance AS events occur in different tissue systems. Similar results are reported by Johnson et al. (2003) using microarray analyses of human tissues.

In addition to producing protein isoforms, AS produces a disproportionate number of transcription factors with intrinsically disordered protein (IDP) domains, which leads to a synergistic expansion of functional and regulatory diversity (Liu et al., 2006; Vuzman and Levy, 2012). In the case of the Drosophila protein Ultrabithorax (Ubx), different spliceoforms have different affinities to common target sequences. Consequently, the isoforms are not interchangeable during development (Reed et al., 2010). More generally, when GRN-related transcription factors are alternatively spliced the function of the GRN may vary in a spatiotemporal fashion under the influence of physiological and physical factors external to the network.

INTRINSICALLY DISORDERED PROTEIN (IDP) DOMAINS

Intrinsically disordered protein (IDP) domains often form small interaction surfaces characterized by multiple specificity and modest affinity, an enhanced binding diversity, the ability to form large interaction surfaces with high affinity, fast association and dissociation rates, polymorphism in the bound state, and wide range of intracellular lifetimes (Dunker and
Uversky, 2010; Oldfield and Dunker, 2014). These traits make IDPs versatile signaling and regulatory molecules. Studies have identified intrinsically disordered domains as enriched in the non-constitutive exons, indicating that protein isoforms may display functional diversity due to the alteration of tissue-specific modules within these regions (Buljan et al., 2012). IDP domains can exist as molten globules with defined secondary structure or as unfolded chains that can function through transitions among different folded states. Their functional conformations can change by binding to other proteins and nucleic acids (Uversky, 2002; Oldfield et al., 2008; Hsu et al., 2013). IDPs also contribute to the process of alternative splicing: the RS-repeat domains of the conserved SR family of metazoan splicing factors are intrinsically disordered (Braunschweig et al., 2013). Post-translational modifications can also alter IDP functionalities (Iakoucheva et al., 2004; Dyson and Wright, 2005; Oldfield et al., 2008).

Numerous examples of IDP domains involved in transcriptional regulation are known (Campbell et al., 2000; Haynes and Iakoucheva, 2006; Liu et al., 2006; Sun et al., 2013). The C-terminal activation domain of the bZIP proto-oncoprotein c-Fos, which effectively suppresses transcription in vitro, is intrinsically disordered and highly mobile (Campbell et al., 2000). The C-terminal domain of the transcriptional co-repressor CtBP, which facilitates gene targeting and coordinated histone modifications in the multi-protein complex, is intrinsically disordered (Bhalla et al., 2006; Haynes and Iakoucheva, 2006; Sun et al., 2010). The unbound N-terminal domains of the DELLA proteins, which are central to the integration of plant developmental and environmental signaling, undergo disorder-order transitions upon binding to interacting proteins (Sun et al., 2010). The DELLA proteins are similar in their domain structures to the GRAS protein family, whose N-domains are intrinsically disordered (Sun et al., 2011) and are extensively involved in plant signaling by virtue of undergoing disorder-order transformations in interactions with a variety of molecular partners involved in development, light signaling, nodulation, and auxin signaling and transcription regulation to biotic and abiotic stresses.

Metazoans also carry out intercellular signaling via small molecules, called nuclear hormone receptors (NRHs), that bind to their cognate proteins. Following ligand binding, NRHs translocate to the nucleus where they act as transcription factors. In addition to the structured ligand and DNA binding domains, these NRHs have flanking and linking IDP domains that bind to large numbers of partners. These domains may be responsible for the variable or context dependent responses following hormone signaling (Simons and Kumar, 2013). Thus, NRHs use disorder to bind to many partners, and many partners use disorder to bind to structured docking sites on NRH ligand binding domains (Mohan et al., 2006; Dunker et al., 2014).

Another important example is provided by the Wnt pathway. This key signaling pathway, which is utilized in development from sponges to flies to mammals (Cadigan and Nusse, 1997; Nusse and Varmus, 2012), employs both IDPs and PTMs in fundamental ways. Briefly, β-catenin, a dual-function cofactor for adhesion and transcription, is phosphorylated at four nearby sites in a disordered tail by the destruction complex. This complex is held together by the disordered scaffold protein axin, which uses a long disordered region to flexibly tether β-catenin to two kinases, GSK3β and CK1α, thus speeding up the phosphorylation reactions by colocalization (Xue et al., 2012b; Dunker et al., 2014). These multiple phosphorylation events regulate both nuclear localization and proteasomal digestion of β-catenin. The activity of adenomatous polyposis coli (APC), a massively disordered member of the β-catenin destruction complex, is also regulated by phosphorylation (Minde et al., 2013). Thus, β-catenin accumulates, translocates to the nucleus, and turns on several genes that activate cell proliferation and polarity.

These examples illustrate that intrinsically disordered transcription factor domains are central to plant and animal development and homeostasis. They are by no means exceptional. Liu et al. (2006) found that 82.6–93.1% of the transcription factors in three databases contain extended regions of intrinsic disorder, in contrast to 18.6–54.5% of the proteins in two control datasets. Focusing on human transcription factors and using a disorder predictor and Hidden Markov Models to search for regions that are homologous to structured protein domains, Minezaki et al. (2006b) report that only 31% of the transcription factor residues align with known structured domains, which is only half of the 62% structurally aligned residues for E. coli proteins that regulate transcription.

Since protein-DNA recognition and protein-protein recognition are central transcription factor functionalities, these and other studies illustrate the extent to which eukaryotic transcription factors manifest extensive flexibility as a consequence of disorder-associated signaling and transcriptional regulation (Dunker et al., 2014). This permits them to bind to a greater array of partners that in turn can induce conformational changes in bound protein and DNA substrates (Oldfield et al., 2005). A well-studied example of this is the isoforms of the Drosophila Ubx transcription factor described above. Here, two intrinsically disordered domains modulate the binding affinity of the structured DNA-binding homeodomain to its target sequence (Liu et al., 2008) and to other transcription factors (Johnson et al., 1999; Bondos et al., 2006; Hsiao et al., 2014). The C-terminal IDP region, which is alternatively spliced, alters the relative affinity of Ubx for different DNA sequences (Liu et al., 2009).

**POST-TRANSLATIONAL MODIFICATIONS (PTMs)**

Post-translational modifications (PTMs) alter the regulatory interfaces of proteins so as to induce gain, loss, or exchange of binding partners, thereby affecting function at many levels (Van Roey et al., 2013). Significantly, the structure of chromatin, the mechanochemical medium within which eukaryotic transcription occurs, is regulated by PTM of histone proteins. Mediator, a multi-protein complex involved in RNA Pol II-regulated transcription, is both positively and negatively regulated by phosphorylation (Gonzalez et al., 2014). Combinatorial PTMs of the C-terminal domain of RNA Polymerase II regulate multiple stages of transcription initiation and coordinate transcription with mRNA processing (Yogesh et al., 2014).

However, our focus here is on the effect of PTMs on specific transcription factors. In transcriptional regulation, each
transcription factor must participate in many different macromolecular recognition/binding events (Bondos and Tan, 2001; Sun et al., 2013; Abdel-Hafiz and Horwitz, 2014). Transcription factor binding to DNA often occurs in conjunction with other specific transcription factors, requiring tissue-specific protein-protein interactions as well. Transcription factors must interact with Mediator or other components of the general transcription apparatus to elicit their function. Many transcription factors that are active in developmental processes also bind histone acetylases and de-acetylases. Phosphorylation can regulate each of these events. For example, DNA binding by the transcription factor Ets-1 is allosterically coupled to a serine-rich region (Lee et al., 2008; Mooney et al., 2014). Ca2+ signaling induces phosphorylation of this region, which modulates DNA binding by Ets-1. Phosphorylation of the intrinsically disordered PAGE4 protein (as part of the stress-response pathway) causes PAGE4 to release the transcription factor c-Jun, enabling its activity in transcription regulation (Mooney et al., 2014). Phosphorylation can also increase interactions among cofactors. For example, the cytokines TNF and IL-1 induce phosphorylation of the p65 subunit of NF-kB, which in turn induces a conformational change that allows p65 ubiquitination and interaction with transcriptional cofactors (Milanovic et al., 2014). Association of Elk-1 and ETS domain transcription factors with Mediator and histone acetyltransferases is dependent on Elk-1 phosphorylation (Galbraith et al., 2013).

As a final example, we again turn to the Drosophila Hox protein Ubx (Ronshaugen et al., 2002). This transcription factor is multiply phosphorylated (Gavis and Hogness, 1991), including at sites within its intrinsically disordered domain, which regulates DNA binding, protein-protein interaction and transcription activation (Tan et al., 2002; Bondos et al., 2006; Liu et al., 2008, 2009). Given that phosphorylation has the potential to regulate as well as coordinate multiple transcription factor functions, it is not surprising that this mechanism is widely used. Indeed, transcription factors are disproportionately phosphorylated compared to other classes of proteins (Kaganovich and Snyder, 2012). Furthermore, the divergence of the sequence and function of transcription factor paralogs created by whole genome duplication events correlates positively with the extent to which the transcription factor is phosphorylated (Kaganovich and Snyder, 2012).

SYNERGISTIC EFFECTS OF AS-IDP-PTM

Importantly, although AS, IDP, and PTM can operate independently of one another, they are often more co-localized to operate synergistically. The co-localization of AS, IDP, and PTM is apparent in many ways. For example, pre-mRNA segments undergoing AS are far more likely to code for IDP domains than for structured domains. These AS-associated IDP domains also frequently contain binding sites for protein or nucleic acid partners such that they operate together to “rewire” GRNs (Romero et al., 2006; Dunker et al., 2014). The AS–IDP collaboration to rewire GRNs is commonly observed at the tissue-specific level and is well conserved over evolutionary time (Builjan et al., 2012, 2013; Ellis et al., 2012; Colak et al., 2013).

IDP domains are also far more likely than structured regions to undergo PTMs, especially the phosphorylation of serines and threonines (Iakoucheva et al., 2004; Gao et al., 2010; Gao and Xu, 2012). These IDP-associated PTMs are often observed to alter partner choice for IDP-based protein-protein interactions (Oldfield et al., 2008; Hsu et al., 2013), which can further rewire GRNs. In addition, different patterns of multiple PTMs in localized protein regions have been shown to signal different downstream results, leading to their designation as a histone or PTM “code” (Strahl and Allis, 2000; Lothrop et al., 2013). Finally, “constellations” of multiple PTMs generally occur in IDP regions, (Pejaver et al., 2014), some examples of which have been shown to be further modified by AS (Dunker et al., 2008).

AS–IDP–PTM PHYLOGENETIC PATTERNS

Evidence drawn from phylogenetically different lineages indicates that AS–IDP–PTM is ancient and has undergone significant amplifications during the prokaryotic-to-eukaryotic transition. For example, analyses of the Viridiplantae (the green and chlorophyan algae, and the land plants) show that early divergent unicellular chlorophytes employ AS extensively and that the frequency of AS in the unicellular green alga Chlamydomonas reinhardtii is comparable to that of the flowering plant A. thaliana (Labadorf et al., 2010). Many of the ancient physiological processes in the Viridiplantae rely on IDPs, e.g., the extensively disordered N-terminal region of the CP12 protein regulates two critical (and extremely ancient) enzymes in the Calvin cycle (glyceraldehyde-3-phosphate dehydrogenase and phosphoribulokinase) (Mileo et al., 2013). More generally, in a comprehensive study using over 39 million expressed sequence tags available for 47 eukaryotic species with fully sequenced genomes, Chen et al. (2014b) found that the occurrence of AS has increased steadily over the last 1.4 billion years of eukaryotic evolution. The frequency of AS is not due to covariance with other factors proposed to account for organismic complexity, e.g., genome size, protein interactivity, and proteome disorder. These authors conclude that organismic complexity, as gauged by the number of different cell types, has increased as a result of AS driven transcript diversification that has increased the information content of cells (Chen et al., 2014b).

Less is known about the extent to which IDP–PTM has changed over evolutionary history. Quantitative measures of proteome intrinsic disorder are only recently becoming available. However, a positive relationship between a large number of proteins with intrinsically disordered domains and the extent to which species are evolutionarily derived has been noted. This relationship appears to be step-wise rather than continuous, which likely reflects major evolutionary transitions. Xue et al. (2012a) examined 3484 viral, bacterial, and eukaryotic proteomes and found that the largest variance of intrinsically disordered content occurred among the viruses (i.e., 7.3–77.3%), whereas only a weak correlation between complexity as gauged by the number of different cell types and overall ID domain content was observed within the eukaryotes. These authors also report that the ID domain content is generally independent of proteome size for both the prokaryotes and eukaryotes, but that it is significantly higher for eukaryotic compared to prokaryotic species and possibly correlated with the more elaborate signaling systems eukaryotes use to coordinate their intracellular functions (Xue et al., 2012a). Schad et al. (2011) report that complexity (as
gauged by the number of cell types) and proteome size (measured as the total number of amino acids) correlate positively across diverse organisms, and that the fraction of ID domains increases significantly from prokaryotes to eukaryotes, but does not increase further within the eukaryotes.

However, in contrast to the aforementioned study, which did not delve into a species-level analysis of the data, Niklas et al. (2014) have uncovered a statistically robust ($r^2 = 0.721, P < 0.0001, F = 44.0$) log-log linear relationship between the number of different cell types and the fraction of ID residues in the proteomes reported for a diverse group of unicellular and multicellular algae, land plants, invertebrates, and vertebrates (spanning genera such as Chlamydomonas, Volvox, Arabidopsis, Hydra, Caenorhabditis, Drosophila, and Homo sapiens). Perhaps more significant, the slope of this log-log linear relationship numerically significantly exceeds unity, which indicates that a small increase in the fraction of proteomic ID residues is correlated with disproportionately large increases in the diversity of cell types. As in the Schad et al. (2011) study, Niklas et al. (2014) found that the slope for the log-log linear relationship between the number of different cell types and genome size (as gauged by base-pair numbers) is less than unity, which is consistent with the so-called C paradox. Clearly, as noted by many workers, statistically robust correlations between any two variables of interest are not evidence for cause-effect relationships. Nevertheless, strong correlations can be taken as evidence for consistency between empirical observations and theoretical expectations.

**RESOLVING STOCHASTIC DEVELOPMENTAL EFFECTS**

Many developmental processes appear initially disorganized but subsequently produce an ordered, patterned structure. Live images of Drosophila embryos provide striking examples of this phenomenology (Bothma et al., 2014). For example, fluorescent imaging has been used to monitor the genesis of the second stripe of eve expression in Drosophila, a critical step in the segmentation stage of development (Figures 2A–C). Eve is initially activated in a broad stripe, in which cells expressing eve are mixed with cells lacking eve expression. This observation reflects an initial randomness of the initial decision to transcribe (or not) the eve gene. However, over time, the eve expression domain is refined to a narrow stripe of homogeneous eve-expressing cells. The initial dynamic behavior is due in part to short bursts of eve transcription characterized by a range of Pol II loading rates, indicative of non-deterministic behavior. Although deterministic systems

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**FIGURE 2** | Diagrams of a section of Drosophila embryo expressing the eve–MS2 reporter at three different times in nc14 centered at ≈37% embryo length (A–C) and POND scores for order vs. disorder in four transcription factors (D). Nuclei that show foci of active transcription are depicted in green. (A–C) Redrawn from Bothma et al. (2014, see their Figures 2A–B).
often exhibit transient behavior en route to achieving their stable states, the eve patterning mechanism is not conventionally deterministic. Whereas the maternally deposited transcription factor Bicoid is essential for the correct spatial positioning of eve stripe 2, and the latter’s target CREs in the stripe 2 eve enhancer are well-characterized (Ludwig et al., 2011), studies have shown that disrupted, unstable, and highly abnormal Bicoid gradients fail to disturb the precision of this process (Luccetta et al., 2008).

We propose that the initial heterogeneity in the eve patterning system is related to the effects of AS-IDP-PTM on the transcription factors regulating this gene, which is activated by Hunchback and repressed by Giant and Krüppel. According to a recent computational analysis (Ilsley et al., 2013), Bicoid has a dual context-dependent activator/inhibitor role in eve 2 expression, although the mechanism for this is unknown. All four transcription factors contain large intrinsically disordered domains (Figure 2D). In addition, all four transcription factors are phosphorylated and bicoid is alternatively spliced (Ollo and Maniatis, 1987; Capovilla et al., 1992). The variation in transcription levels during these burst phases could be the result of regulation by different spliceoforms or phosphoforms of these proteins.

These issues have been explored in greater detail for Ubx, where the nature of intrinsically disordered domains provides a basis for both this early stochastic behavior, and mechanisms to resolve such behavior into an ordered response. The alternative splicing and phosphorylation of Ubx are tissue-specific, creating different dominant forms of the transcription factor in each tissue with tissue-specific capacities for protein interactions, DNA binding, and transcription regulation (Gavis and Hogness, 1991; Liu et al., 2008; Kim et al., 2010; Reed et al., 2010; Fuxreiter et al., 2011; de Navas et al., 2011). However, in any given cell minor forms created by splicing and phosphorylation are also present, yielding a mixture of Ubx functional states (O’Connor et al., 1988; Gavis and Hogness, 1991; Lopez and Hogness, 1991). In our model, the form of the Ubx protein that first binds a newly available gene target is expected to determine the initial response, creating an initial variation in transcription levels and stochastic phenotypes.

As in the case of Bicoid (Ilsley et al., 2013), Ubx has context-dependent dual activator/inhibitor roles and its “collaboration” with other transcription factors, which is regulated in a spatiotemporal fashion, can determine the “sign” (positive or negative) of its regulatory role (Walsh and Carroll, 2007) and thus profoundly influence GRN logic. Such modulation, canalization, and refinement of the Ubx response is likely to depend on post-translational modification or protein interactions mediated by the intrinsically disordered regions of this protein. Like most transcription factors in development, Ubx (i) regulates genes encoding cell signaling proteins (Pearson et al., 2005; Bondos et al., 2006), (ii) is regulated (phosphorylated) by cell signaling proteins (Gavis and Hogness, 1991; Taghli-Lamallem et al., 2008), and (iii) binds cell signaling proteins and cell signaling-regulated transcription factors (Liu et al., 2008). These mechanisms enable the community of cells to make a collective decision regarding gene regulation. Binding by the form of Ubx that is supposed to regulate a specific target gene enhancer will be supported by the presence of other factors that cooperatively regulate this gene in conjunction with Ubx. Downstream cell-signaling events could further reinforce this decision within a neighboring group of cells. In contrast, binding by the incorrect form of Ubx may lack the requisite co-factors and signaling to stabilize the bound complex, ultimately resulting in dissociation of the protein and providing a second opportunity for the correct Ubx form to bind. In this paradigm, AS–IDP–PTM–protein interactions (i) generate the initial stochastic behavior, (ii) are required to reinforce the correct cell decisions, and (iii) mediate the rectifying response (Figure 3).

The described behaviors appear to differ from the transients exhibited by deterministic dynamical systems (such as GRNs in the standard model), as they evolve toward their “attractors,” i.e., stationary points and orbits, or in the case of Turing-type reaction-diffusion systems (reviewed in Forgacs and Newman, 2005), stationary non-uniform spatial patterns. The temporal evolution, and the distribution and stability of attractors, are
strongly dependent on the network topology of such systems. In contrast, the rewiring of Bicoid and Ubx regulatory circuits en route to the biologically functional patterns in which they function suggests that, rather than determining the end-states of the respective systems, the GRNs are actually subordinate to them.

**DISORDER FROM ORDER**

According to the most mathematically sophisticated deterministic GRN dynamics models (e.g., Foster et al., 2009; Jaeger and Monk, 2014), each cell type is an attractor. That is, if a cell’s state at a given time is represented by a point in a multidimensional “state space” whose axes are the concentration ranges of key transcription factors, the point’s position will change until it settles stably at one of a finite number of discrete sub-regions within the space (Figure 4). These sub-regions (i.e., system attractors) can be stationary points, periodic orbits, or a mixture of these behaviors, depending on the subset of the components involved in the system. Deterministic systems of sufficient complexity can also exhibit the so-called “butterfly effect,” in which an infinitesimal displacement of the system point can take it along widely divergent trajectories, as well as chaotic behaviors, characterized by “strange attractors,” i.e., regions within the state space in which a point remains bounded but wanders in an unpredictable fashion (Strogatz, 2001; Kaneko, 2006). Each attractor in a deterministic dynamical system is surrounded by a “basin of attraction” toward which a system point gravitates. Importantly, the number of attractors within a deterministic dynamical system is, in principle, a predictable function of its network topology and rate constants, and is always much smaller than the number of basic interacting components. The rationale for applying this mathematical formalism to GRNs and cell differentiation thus arises from observations like the fact that the human genome specifies more than 1300 transcription factors (Vaquerizas et al., 2009) but the human body contains only about 244 cell types (Niklas et al., 2014).

If a system is less than fully deterministic, the dynamics become much more complicated. One example is “noisy systems” in which conventional network topologies and interactions are in place, but the values of the variables (such as concentrations of transcription factors) are perturbed by extrinsic or intrinsic factors, as a consequence of some key proteins and other biomolecules being present in small numbers in individual cells and thus varying in a stochastic fashion (Elowitz et al., 2002; Bhalla, 2004; Gomez et al., 2014). Indeed, mathematics has shown that the notion of an “attractor” still applies, but that their properties and thus behavior are less predictable than in systems without noise (Jacobs and Schreiber, 2006; Zhao and Li, 2011).

The issue of noise is well recognized to confound the biological effects of even those gene regulatory systems that are formally deterministic. For example, Rosenfeld and coworkers examined the bacteriophage lambda promoter $P_R$ in *E. coli* and found...
that protein production rates fluctuated over the time scale of one cell cycle, with intrinsic noise levels of the circuit decaying rapidly within each cycle (Rosenfeld et al., 2005). Nonetheless, the aggregate effect of fluctuations in other cellular components undermined accuracy in transcriptional responses for time-scales longer than a single cell cycle. Thus, although individual circuits can be demonstrated to behave according to a deterministic GRN dynamics model over short time intervals (Rosenfeld et al., 2007), GRN-level determinism breaks down when embedded in the wider network of real complexity and temporal scale. It is our contention that the gene regulatory indeterminism produced by AS-IDP-PTM is different in kind from that of deterministic dynamical systems operating (as described above) under nonlinear, chaotic, or stochastic regimes. GRNs acted upon by AS-IDP-PTM are inherently non-deterministic, since network logic (i.e., connectivity relationships), strengths of interactions, and even the identities of the transcription factors as regulatory molecules are constantly subject to change due to internal fluctuations and external influences. As a consequence, the properties of the transcription factor components of GRNs that have been thought to make them suitable to be represented as nodes in discrete networks or variables in systems of ordinary differential equations are actually subject to change from point to point and moment to moment during development. Although GRNs have been modeled as Boolean networks acting “at the border between order and chaos” (Shmulevich et al., 2005), the presence of AS-IDP-PTM in actual GRNs raises questions about whether any strictly deterministic models can adequately capture the behavior of these systems (Figure 4).

ORDER OUT OF DISORDER

As noted earlier, GRNs are frequently not deterministic due to the independent and interdependent actions of AS-IDP-PTM. Nevertheless, this is in no way equivalent to the counterfactual assertion that embryonic development is itself non-deterministic. Rather, it is our hypothesis that deterministic GRN dynamics are not a sufficient causal basis for developmental regularities. Although a GRN might provide a rough template for a cellular function (particularly if the GRN was established concurrently with the evolutionary origination of that function), remodeling of the GRN by AS-IDP-PTM will have rendered cell phenotype identity increasingly dependent on internal (i.e., cell physiological) and external (e.g., microenvironmental and extraorganisinal) conditionalties beyond the GRN itself. This assertion is consistent with, if not confirmed by, somatic stem cell production and subsequent differentiation as well as examples of dedifferentiation (e.g., Sprecher and Desplan, 2008).

The conservation of a useful cell function or morphological phenotype over the course of evolution accompanied by an unmooring from its originating GRN appears to be a common scenario in the history of multicellular plants and animals, reflected in what has been termed “developmental system drift” (True and Haag, 2001; Haag, 2014). The inability to consistently pin heritable variation in diseases and other traits to particular genes (Zaitlen and Kraft, 2012), may also plausibly be a manifestation of the operation of AS-IDP-PTM over evolutionary time. AS-IDP-PTM may also provide flexibility and adaptability even within a single tissue type. For example, in a study of lineage commitment among the eight progenitor populations of the major myeloid and lymphoid elements of human blood, Chen et al. (2014b) identified cell type-specific expression changes during early differentiation stages encompassing 6711 genes and 10,724 transcripts. They also detected 7881 novel splice junctions and 2301 differentially used AS events, enriched in genes involved in regulatory processes. Although only AS was considered, the authors concluded that “a previously undetected layer of regulation affecting cell fate... involves transcriptional isoforms switching without noticeable changes at the gene level” (Chen et al., 2014b).

Finally, AS-IDP-PTM and its synergies provide a context for understanding how the functionalities of ancient proteins and regulatory networks can be stably modified over the course of evolution to adapt to changing external conditions. Target sequence recognition and selectivity by a transcription factor are subtle properties of the latter’s structure (De Masi et al., 2011). It is well documented, for example, that novel relationships between protein structure and PTM educed by mutation can lead to altered protein-protein interactions resulting in dramatic changes in transcription factor function (Brayer et al., 2011; Lynch et al., 2011). However, synergy with AS and IDP provides an even greater multiplicity of functional states that can be explored ecologically and physiologically ahead of any mutational change.

Furthermore, nascent potentially adaptive mutations can be retained within (and subsequently integrated into) GRNs by virtue of AS-IDP-PTM modifications that can buffer GRNs from the immediate consequences of such mutations. In this scenario, a mutated GRN could survive by virtue of AS-IDP-PTM adaptive modifications that would permit the GRN time to adaptively reorganize. In this way, evolutionary changes would involve an interactive “genome ↔ AS-IDP-PTM” feedback loop. Consider the transcription factor AkUbx, a homolog of Ubx in the velvet worm Acanthokara kaputensis, an invertebrate with a simple body plan. AkUbx has very little intrinsic disorder and is not alternatively spliced (Grenier and Carroll, 2000; Galant and Carroll, 2002). Ubx, in contrast, which participates in the development of the later-evolving, more complex body plans of Drosophila melanogaster has considerable disorder content as well as undergoing AS and PTM (Gavis and Hogan, 1991; Liu et al., 2008; Reed et al., 2010). The synergistic effects of AS-IDP-PTM ensured that once it had arisen in the earliest multicellular GRNs it would have promoted its own elaboration, as well as generation of new developmental contexts that would eventually be reflected in greater anatomical and physiological complexity.

CONCLUSIONS

The association of alternative splicing (AS) with intrinsically disordered protein (IDP) domains and post-transcriptional modifications (PTMs) is a core functional complex that mediates the modifications of protein functionalities required for context dependent cell signaling, regulation, and differentiation. The combined effects of AS-IDP-PTM also likely buffer genomes...
from mutations (some of which can subsequently become adaptive to new conditions) and contributes to the evolvability of GRNs (see for example, Masel and Trotter, 2010; Steiner, 2012; Albergante et al., 2014). AS–IDP–PTM is ancient and likely promoted variability and thus adaptive evolution to support more complex intracellular signaling processes coordinating the activities of functionally interdependent discretized organelles, cells, tissues, and organs.

Unlike promoter activity, which primarily regulates the amount of transcripts, AS changes the structure of transcripts and their encoded proteins. The ability of IDP domains to assume different conformations expands the functional repertoire of proteins assembled by AS from a pre-mRNA to diversify the phenotypic domain that a single genome can provide. This repertoire is yet again increased by PTMs, which generate additional functionalities. Thus, AS–IDP–PTM can yield virtually limitless combinatorial possibilities, which can be adaptively sifted over the course of evolution.

Consequently, GRNs are inherently plastic and therefore adaptive. Moreover, they function in a noisy cellular milieu owing to the operation of AS–IDP–PTM in a multitude of other biochemical pathways as well as the effects of mutations and variations in gene and protein copy number (Richard and Yvert, 2014). (Note that this noisiness is over and above the described intrinsic indeterminacy.) The evolution of cell differentiation may indeed have depended on such stochastic effects (Kupiec, 2009). However, heterogeneity at both the molecular and cell phenotypic levels must be suppressed for reliable development to occur. This is accomplished by a variety of “scaffolding” effects (Caporael et al., 2014) at multiple scales, including consistency of external cues from neighboring cells and the physical environment (Braendle and Félix, 2008), and the stabilizing effects of natural selection (Richard and Yvert, 2014).

The multiscale nature of developmental processes is increasingly acknowledged (see, for example, Schnell et al., 2008). In particular, tissue morphogenesis and cellular pattern formation involves the mobilization, by key gene products of the developmental “toolkit,” of mechanical, electrical and other physical phenomena external to the genome (Forgacs and Newman, 2005; Newman and Bhat, 2009; Hernández–Hernández et al., 2012). It is therefore unsurprising that the determination of cell type identity does not reside at the single scale occupied by GRNs, but rather draws on factors at several causal levels, as described above, among the most important of which are the mechanistic aspects of chromatin reorganization associated with changes in gene expression (Amendola and van Steensel, 2014; Lavelle, 2014).

We do not suggest that deterministic mathematical and computational modeling of GRNs has nothing to contribute to understanding cell fate determination. However, this perspective must acknowledge and integrate the ubiquitous effects of AS–IDP–PTM. Just as genes per se have long been rejected as the exclusive or privileged level of determination of phenotype and evolutionary change, new understanding of the complexities of gene expression and the conditional identities of its protein products call into question a deterministic GRN-based reductionism in developmental and evolutionary biology.

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