Evolution of Premalignant Disease

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Where does cancer come from? Although the cell-of-origin is difficult to pinpoint, cancer clones harbor information about their clonal ancestries. In an effort to find cells before they evolve into a life-threatening cancer, physicians currently diagnose premalignant diseases at frequencies that substantially exceed those of clinical cancers. Cancer risk prediction relies on our ability to distinguish between which premalignant features will lead to cancer mortality and which are characteristic of inconsequential disease. Here, we review the evolution of cancer from premalignant disease, and discuss the concept that even phenotypically normal cell progenies inherently gain more malignant potential with age. We describe the hurdles of prognosticating cancer risk in premalignant disease by making reference to the underlying continuous and multivariate natures of genotypes and phenotypes and the particular challenge inherent in defining a cell lineage as “cancerized.”

As the second-leading cause of death worldwide after cardiovascular disease, cancer claimed >14 million lives in 2012 (Torre et al. 2015). This number is expected to increase over the next few decades as a result of the aging human population and increasing prevalence of cancer risk factors worldwide. Although this statistic is alarming from a public health perspective, we must initially pose the question, “what exactly constitutes a cancer case?” With advancement of basic biological understanding, the definition of cancer itself has continued to evolve over the past 2000 years. In fact, even current established hallmarks of malignancy (Hanahan and Weinberg 2011) continue to be revised and reveal the complexity involved in distinguishing between cancerous and noncancerous tissue. In a literal sense, global regions may differ in clinical definitions of the morphological threshold between a preinvasive carcinoma in situ and an invasive tumor (terms defined in detail below), and even within an individual hospital two pathologists may disagree whether a diagnosis of dysplasia versus malignancy should be applied to a particular biopsy specimen. The ability to identify a precancerous state has a major influence on our understanding of cancer prevalence and hence our aims for cancer detection, prevention, and therapy.

Underlying these ambiguities is the fact that cancer progression is a stochastic process that occurs through somatic evolution (Nordling 1953; Nowell 1976; Vogelstein and Kinzler 2004; Merlo et al. 2006; Yates and Campbell 2012). How is the initial cancer cell initiated? From early development until death, normal dividing cells within the body act as asexual, quasi-organisms subject to evolutionary pressure from microenvironmental constraints. Because of imperfect DNA replication and
carcinogen exposure, somatic genomic abnormalities (SGAs; such as point mutations and copy number alterations, and epigenetic changes) accumulate and some may confer a fitness advantage, such as increased reproductive rate. Advantageous alterations, often known as “drivers,” will be clonally selected. Within a selected clone, subsequent driver mutations may be acquired, leading to subclonal expansions and branched lineages from the genotype of the most recent common ancestor (MRCA) of the clone.

This microevolutionary process of Darwinian natural selection on the timescale of the human lifetime can produce genetically diverse clonal populations, tumors with complex clonal architectures and heterogeneous tissue microenvironments (Michor et al. 2003; Merlo et al. 2006; Graham and McDonald 2010; Baker et al. 2013; Greaves 2015). Eventual neoplasms also show marked heterogeneity in their cellular morphology and clonal architecture (Greaves and Maley 2012). Consequently, although there are definitive clinical phenotypes identified through the natural history of a disease (e.g., the adenoma–carcinoma paradigm in colorectal cancer [CRC] progression), there, in fact, exists a “continuum” of cell types, both genetically and phenotypically speaking, throughout all stages of carcinogenesis.

A main focus of this work will be the relationship between genotype and phenotype, including the impact of defining disease progression based on histology (e.g., the Barrett’s metaplasia–dysplasia–cancer sequence) versus genetic composition (e.g., the sequential or otherwise acquisition of genetic changes). The “closeness to cancer” is typically categorized by histology, as most epidemiological studies aim to assess cancer risk in a group of people with a particular disease stratified by clinical stage of neoplastic progression, such as comparing cancer development risk in Barrett’s esophagus (BE) patients without dysplasia versus patients with high-grade dysplasia (HGD). There are many defined premalignant, or precancerous, diseases that confer a higher risk of cancer progression versus that of a nonaffected individual (Table 1) (Fitzgerald 2010).

In this review, we will question exactly what constitutes premalignant disease. From the view of multistage theory (Moolgavkar 1978), does in fact every malignancy originate from a premalignant cell progeny that could theoretically be detected by a perfectly sensitive screen? Biologically, is a first malignant cell ever born de novo, that is, it inherits no SGAs known to increase cancer risk nor has a phenotypically premalignant ancestor cell? And most important, can we reduce cancer mortality by identifying certain premalignant changes early?

To address the first questions, we know that (usually) cancer arises monoclonally from a cell lineage that acquired multiple mutations in cancer-associated genes such as tumor suppressor genes (TSGs) and oncogenes (Michor et al. 2004, Vogelstein et al. 2013) and/or numerous epigenetic alterations (Feinberg et al. 2016). As an example, Knudson (1971) famously showed that retinoblastoma in children was caused by two rate-limiting events, the biallelic inactivation of TSG RB1. If a child inherits a single inactivated allele, this event translates to a single “hit” or initiating event. Next, the inactivation of this TSG initiates clonal expansion of mutated cells because it permits unsuppressed cell proliferation, the sine qua non of carcinogenesis (Moolgavkar and Knudson 1981). However, covert premalignant lesions already exist in relatively high percentages in infants—1% of all newborns were found to have cells with acute lymphoblastic leukemia mutation and histopathologically identifiable precursor lesions, 100 times the corresponding clinical cancer rates (Mori et al. 2002). Notably, such early genetic events have been found in monozygotic twins who share the same premalignant clones in utero (Greaves et al. 2003). Thus, there is evidence that even childhood cancers that require few mutations for a malignant phenotype are initiated from a premalignant clone, and even normal fetal development produces silent, genetically diverse premalignant lesions. Further, regardless of stochastic clone fate, cell lineage-tracing experiments suggest that clonal diversity generated by neutral drift of actively self-renewing stem cells may be a universal pattern in all stem cell compartments necessary to
achieve homeostasis (Klein and Simons 2011; Blanpain and Simons 2013). This is found strikingly in aging populations, with 10% of adults over age 65 having multiple somatic mutations in their blood cells, frequently in three genes that have been previously implicated in hematologic cancers (Genovese et al. 2014).

Here, we discuss the controversies surrounding the semantics of premalignant disease and review some main aspects of premalignant evolution such as accumulation of driver mutations, influence of tissue architecture, tissue aging, and roles of the microenvironment. Lastly, we examine some important implications in disease prognostication and patient management.

**WHAT IS PREMALIGNANCY?**

The term “premalignant” describes a condition that may (or is likely to) become cancer (National Cancer Institute 2016). In practice, the meaning of this term may differ between genet-

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For the purposes of this review, we will refer to “premalignant” conditions as all clinically diagnosed morphological lesions known to be a precursor to a certain malignancy and thus they increase an individual’s risk of developing cancer. We note that it may be the case that not all morphological changes that are associated with an overall increase in cancer risk are themselves premalignant; some benign abnormal lesions will never progress to a neoplasm with time. Alternatively, progression to a neoplasm is also not a prerequisite for a lesion to be termed premalignant—in fact, most premalignant lesions remain benign throughout the lifetime of a host individual, as evidenced by their current higher diagnosed frequencies than that of their associated neoplasms. The term “precancer” used in this article encompasses “premalignant” tissues including metaplasia like BE and also the (presumptively named) “preinvasive” neoplastic lesions including dysplasia and carcinoma in situ. Preinvasive lesions are neoplasms that have neither developed the ability to penetrate deeper layers of epithelium nor acquired the propensity to metastasize and grow in other parts of the body (again, we note later that it may be that some so-called preinvasive lesions will never invade). We provide examples with references of such precancerous conditions for the most common epithelial cancers in Table 1. The extensiveness of this list highlights the high prevalence of cancer precursors being diagnosed in current medical practice and prompts the hypothesis that in fact every epithelial cancer arises from a potentially detectable precancerous condition.

It is interesting to note that there has been considerable controversy within the medical community about whether some lesions such as carcinoma in situ should be classified as cancer or precancer, especially in regard to particular cancer sites (Esserman et al. 2014; Greaves 2014). Notably, the incidence of ductal carcinoma in situ (DCIS) of the breast has dramatically increased as a result of increased mammography screening in the past three decades, currently constituting 20% to 25% of all screen-detected breast cancers in the United States in women ages 40 to 64 (Virnig et al. 2010). Thus, mammography has often been heralded as a true success story for the goal of screening: to detect precancerous or early cancer lesions to intervene with treatment before invasive cancer (potentially) develops. However, cancer screening can bring harms along with benefits. A recent observational study of >100,000 women in the United States diagnosed with DCIS found that cancer-specific mortality was only 3.3% (95% CI, 3.0%–3.6%) at 20 years (Narod et al. 2015). In fact, <1% of the patients in this 20-year study by Narod and colleagues died from breast cancer. We note that all patients in this study received some kind of intervention (mostly, surgery and less often radiation therapy) and so the true “unperturbed” natural history of DCIS left in situ remains unknown (and indeed this is a common issue across tissues). Nevertheless, in response to this study, numerous dichotomizing editorials have been written, some proposing the reconsideration of unnecessary, aggressive therapy for DCIS and even the reassessment of whether the goal of breast cancer screening should be to detect these clustered amorphous calcifications (Esserman and Yau 2015). Others are reluctant to support allegations of overdiagnosis (precancers detected at screening that would not have otherwise become clinically apparent or cause death) and/or overtreatment that may lead to decreased screening efforts for such a prominent women’s health issue (Recht et al. 2016). This is supported by a UK meta-analysis of breast cancer screening trials that concluded that screening recommendations are more beneficial than harmful; the findings predict that for every single breast cancer death averted, about three overdiagnosed cases would be treated (Independent UK Panel on Breast Cancer Screening 2012). These studies shed light on the oftentimes political and psychological aspects of using a term that literally translates to “cancer in place” versus “premalignant,” and perhaps motivate the adoption of an even less loaded term such as “abnormal” that does not necessarily imply a relationship to cancer at all. Consequently, a recent U.S. National Cancer Institute (NCI) working group on cancer screening suggested that even the term “cancer” should be reserved for describing “lesions
with a reasonable likelihood of lethal progression if left untreated” (Esserman et al. 2013), although the issue remains contentious.

Precancerous phenotypes, including those provided in Table 1, are diagnosed based on the morphological features of a tissue. However, these can be further subdivided into tissue-specific phenotypic groups at smaller scales, the range of which resides in a hypothetical, infinite space known as “phenotype space.” Additionally, the advancement of genomic analysis technology has enabled extensive analyses of the “genotype space” of tissues, defined by the complex patterns of DNA mutations and chromosomal alterations observed in premalignant cells. The ultimate goal of modern genetics is to understand how genotype relates to phenotype (Botstein and Risch 2003; Benfey and Mitchell-Olds 2008; Rockman 2008). In mathematical terms, the genotype–phenotype (G-P) map is surjective (for each phenotype there exists at least one genotype that maps to that point in phenotype space) but it is not one-to-one; the notion of a “genetic blueprint” of precancerous and associated cancerous tissues is inadequate, because research continues to show that the relationship between the two spaces is complex (Pigliucci 2010; Nuzhdin et al. 2012). For instance, many genetic alterations may be evolutionarily neutral, and thus have no consequence for the phenotype at all, or certain phenotypic traits may only be “expressed” in certain microenvironmental contexts.

In Figure 1, we provide an example of an evolutionary trajectory of an individual’s genotype mapping to a canonical paradigm of tissue progression (such as seen in BE) through normal, metaplastic, dysplastic, and finally cancer regions in phenotypic space. The outline of the process is: random mutation occurs on the genotype level, pleiotropic effects of such a variant on the phenotype are determined by the G-P map, and finally the phenotype interacts with the environment, and indeed may be modulated by the microenvironment (Houle et al. 2010; Chandler et al. 2013). Natural selection will tend to cause the clonal expansion of the most fit phenotypes in the current microenvironmental context. Analyzing the properties of evolutionary pressures influencing regulatory networks and associated phenotypes helps us to learn the causal relationships captured by the G-P map (Chanock et al. 2007; Gagneur et al. 2013). The overlapping areas in phenotypic space qualitatively illustrate the issue of ambiguity in categorizing phenotypes during neoplastic progression. Thus, although a diagnosis in the clinic is chosen from a handful of pathological stages, the underlying multivariate natures of genotypes and phenotypes means that such classifications are in fact continuous and fluid. This continuity perhaps explains some of the evident challenges in reliably identifying particular premalignant states, such as a low-grade dysplasia (LGD) in BE (Kerkhof et al. 2007; Curvers et al. 2010).

**ROLE OF DRIVER ALTERATIONS IN PREMALIGNANT EVOLUTION**

Although pathological assessment of the presence or absence of premalignant phenotypes remains the gold standard for cancer risk, somatic mutations that naturally occur throughout a human lifetime may also reveal information about the progress toward cancer long before a phenotype such as dysplasia manifests in a tissue (see Fig. 1). During neoplastic evolution, we typically differentiate between “driver” mutations, defined to be those that confer growth or survival advantages to cells that will be positively selected during the evolution of a cell lineage (presumably only when the mutant cells find themselves in the correct microenvironment), and “neutral,” or “hitchhiker,” mutations that passively accumulate in cell progenies (Calabrese et al. 2004; Stratton et al. 2009; Greaves 2015). Generally, lesions with driver mutations are associated with clonal expansion and are found more frequently in premalignant and malignant lesions than is expected from the normal background mutation rate (Maley et al. 2004, Lawrence et al. 2014). However, the exact role(s) of driver mutations in premalignant evolution remains elusive, even with the vast amounts of genetic and epigenetic information on cell lineages being generated through increasingly high-throughput, cost-effective ge-
Figure 1. Trajectory of premalignant evolution via the genotype–phenotype (G-P) map. The G-P map traces the evolutionary path of cells in the two infinite spaces of genotype and phenotype. (A) Time points are specified to illustrate the progression through phenotype space (pink regions denote regions of similar phenotype) from normal to precancerous to cancer, driven by changes in genotype. Movement through the spaces is caused by: (1) the projection of the genotype through epigenetic and microenvironmental “filters” accounted for in the G-P map that produces a phenotype from a genotype; (2) the action of natural selection in the phenotype space, influencing-variation generated in genotype space that moves the population to points of higher (contextual) fitness; (3) genotype of fit parents is preserved deterministically; (4) somatic mutation and/or other genetic event(s) moves average point in genotype space. We highlight the overlap in defined phenotypic regions—at one time point during progression to cancer, this cell population could be classified as either metaplastic or dysplastic histologically. (B) A second example of a stochastic trajectory that jumps as a result of a punctuated event in genotype space, such as whole-genome doubling, without discernable phenotypic change. (C) Small changes in genotype space may also lead to large changes in phenotype space possibly caused by rare variants having large phenotypic effects, epistatic effects between accumulated mutations, and/or other epigenetic or microenvironmental effects. For example, a TP53 inactivating mutation may occur in morphologically normal tissue, but not be selected until subsequent events (e.g., genome doubling) occurs. N, normal tissue; M, metaplasia; H, hyperplasia; D, dysplasia; C, cancer.
nome-wide assays (Metzker 2010). This is partly because of the fact that, beyond the identification of selected mutations in premalignant tissues, such as APC biallelic inactivation in colorectal adenomas (Shibata et al. 1997), TP53 inactivation in BE (Barrett et al. 1999), and BRCA1 and BRCA2 mutations in premalignant breast and ovarian tissues (Welsch and King 2001; Antoniou et al. 2003), the influence that these mutations have on carcinogenesis is both context and stage-specific. As a recent example, Weaver and colleagues (2014) found that even common mutations in SMAD4 and TP53 in BE were not strictly pathological grade-specific, thus decreasing their prognostic power as an unequivocal driver for neoplastic progression. Moreover, rather than relying on a single key driver mutation as necessary for premalignant initiation, there is evidence that premalignant lesions accumulate a “full house” of driver mutations before a first malignant cell is born and then subsequent subclonal expansions within a tumor occur neutrally (Sottoriva et al. 2015; Williams et al. 2016).

Genetic diversity itself may act as a proxy for cancer development risk, regardless whether or not the collection of mutations assayed is purely advantageous and cancer-promoting, because the degree of genetic heterogeneity correlates with the probability that the population will contain (or perhaps generate) a “well-adapted” clone capable of initiating tumor growth. Clonal diversity is highly prognostic in the premalignant condition BE (Maley et al. 2006), and, surprisingly, in the example of BE may remain at relatively constant levels for a number of years (Martinez et al. 2016). The potential utility of clonal diversity as a prognostic measure across disease types was underlined by a recent pancancer study (Andor et al. 2016).

Is the term “driver” useful when thinking about carcinogenesis in premalignant disease? Many so-called driver alterations are also found in normal and premalignant tissues, occasionally at even higher frequencies than found in corresponding malignancies (Kato et al. 2016). As a recent example, Martincorena and colleagues found putative driver mutations in 18% to 32% of normal, sun-exposed skin cells at a density of ~140 driver mutations per square centimeter (Martincorena et al. 2015). Patients with inflammatory bowel disease (IBD) are also found to have putative driver mutations in nondysplastic tissue, with mutations in TP53 TSG being detected commonly in patients who subsequently develop cancer (Leedham et al. 2009; Galandiuk et al. 2012). How can such large numbers of cancer-causing mutations maintain benign phenotypes? One possibility could be that early driver mutations may not greatly increase a clone’s malignant potential without additional genetic events. Bass and colleagues found TP53 mutation to be an early event in esophageal adenocarcinoma (EAC) tumorigenesis, as it was found in concomitant nondysplastic BE tissue (Stachler et al. 2015). The investigators suggest, however, that the early acquisition of this founder TSG mutation alone in nondysplastic BE will most likely never progress to EAC unless cells also undergo whole-genome doubling as a catastrophic chromosomal event.

A second, perhaps more important feature is that the fitness advantage bestowed by driver mutations is inherently context-specific, such that a particular mutation will only be beneficial to the cell within the correct microenvironment. TP53 mutation in the intestine is a good example of context-dependent fitness: in an inducible mouse model, TP53-mutant clones only had a fitness advantage in a chronic inflammatory setting, whereas in normal epithelium they were subject to neutral drift (Vermeulen et al. 2013). Consequently, simply “counting” driver mutations is likely to be an inadequate way to understand premalignant disease; attention must also be paid to the evolution of a tumor-promoting or at least tumor-permissive microenvironment in tandem with the evolution of tumor cells themselves.

Expanding on this perspective, many traditional models of carcinogenesis, wherein selected driver mutations gradually accumulate and increase cancer risk, may actually conflate multiple distinct mechanisms for malignant transformation. There is often undue significance for subsequent cancer risk attributed to mutation of the pretumor cells themselves. Two examples of these anomalies to traditional carcinogenesis...
models are illustrated in Figure 1B, wherein a stochastic jump in genotype space to an abnormality such as whole-genome doubling results in evolutionary stasis in metaplastic phenotype, and Figure 1C, wherein a relatively unchanged genotype progresses from normal to cancer phenotype, perhaps because of extrinsic factors (e.g., changes to the microenvironmental or immune landscape) and/or epigenetic alterations.

Herein lies our main question—what constitutes a premalignant cell? Most cells harboring so-called driver mutations that are common in premalignant disease will never evolve to manifest an invasive malignant phenotype. Is premalignancy then a disease unto itself, regardless of its connection to cancer? Some premalignant diseases require surgery to alleviate symptoms that is irrespective of possible future cancer, such as colectomy for patients suffering from severe colitis symptoms. On the other hand, patients with colorectal adenomas that persist in prolonged evolutionary stasis are also considered at risk for CRC but evidence suggests that patients could often live out their lives with these benign premalignant lesions that never progress (Hofstad et al. 1996).

As evidenced in the previous section, the entire human body, from birth until death, fosters mutations that could lead to cancer if other factors conspire jointly to bring about a cancer phenotype in a tissue. Therefore, our current definition of “premalignant” as “the presence of abnormal tissue that is currently considered a cancer precursor and is associated with an increased cancer risk” is terribly unspecific because arguably all the renewing cells in our body undergo somatic evolution that lead to cancer—in a sense the entire body is therefore premalignant! We need a diagnosis of “premalignancy” to identify cells that are “more at risk” (or particularly at risk) of producing a cancer cell as we age, therefore we require a narrower definition of premalignancy based on an assemblage of factors (when mutations occurred, in what permissive microenvironment, and in tandem with what epigenetic alterations and phenotype) to enable effective prognostication (and clinical utility).

QUANTIFYING THE PACE AND PATTERN OF EVOLUTION

The debate continues whether premalignant clonal populations most often evolve gradually through a sequence of genetic alterations and waves of subsequent expansions that reach fixation in a tissue, or rather if they remain relatively unchanged for long periods of time then suffer rare, yet “catastrophic,” genetic events that cause large leaps toward malignancy. There is increasing evidence supporting the latter pathway, known as punctuated equilibrium (Cross et al. 2016). This process allows numerous intermixed subclones to coevolve covertly before rare clonal expansions occur owing to punctuated events such as phenotype change driven by epistatic interaction following the acquisition of a “full house” of genetic alterations or high-impact individual mutations, or possibly by large-scale rearrangements of the genome such as chromothripsis or whole-genome doubling (Stephens et al. 2011; Greaves 2015; Reid and Paulson 2015; Sottoriva et al. 2015; Williams et al. 2016). For practical purposes, how can we detect the likelihood of such “catastrophic” events in vivo during premalignant evolution? How much information about future cancer progression is predetermined early enough in preneoplastic progression to be useful clinically? Most current phylogenetic analyses that infer clonal patterns and relationships using phylogenetic theory rely on measurements at a single time point, and thus typically cannot resolve missing pieces of information such as the temporal order of early events that have already become clonal, nor detect ancestral clones that went extinct (the evolutionary “dead ends”). However, serial and spatially extensive sampling in longitudinal analyses (Li et al. 2014; Martinez et al. 2016; Siegmund and Shibata 2016) will continue to provide better insight into premalignant evolutionary trajectories with the goal of predicting the occurrence, pace, and prognosis of a future malignancy.

These temporal aspects of evolutionary events during carcinogenesis are also linked with the biological aging of human tissues (Finkel et al. 2007). By analyzing subclones within a
neoplasm, one can build a “molecular clock” that considers genetic changes to occur at a constant rate; each new epi(genetic) mutation constitutes another “tick” of the clock, thus tracking biological age (Alexandrov et al. 2015). Molecular clocks have been constructed using methylation data to infer rates of clonal expansion and determine clonality in aging colonic adenomas (Yatabe et al. 2001; Graham et al. 2011; Humphries et al. 2013) and by tracing lineages of mitochondrial DNA mutated clones in prostate intraepithelial neoplasia (PIN) (Gaisa et al. 2011). Also, studies of both CpG island hypermethylation (Issa et al. 2001) and telomere shortening (Risques et al. 2008) in patients with ulcerative colitis (UC) suggest that this premalignant condition exhibits accelerated or premature tissue aging during CRC progression. For asymptomatic premalignant conditions such as BE, constructing a molecular clock from DNA methylation data allows the quantification of epigenetic drift and thus the statistical inference of hidden clinical time points such as BE onset age (Curtius et al. 2016). Because chronological age is recognized as one of the strongest predictors of cancer risk, renewed attention has been given to exploring the roles of biological tissue age and cellular senescence in premalignant and neoplastic progression (Campisi 2013).

The mosaic patterns of genetically and epigenetically diverse clonal populations in premalignant tissues evolve in space as well as time (Tycko 2003; Leedham et al. 2008; Guo et al. 2015). How do these clones and subclones physically expand in a tissue? Tissue architecture constrains evolution by limiting the ability of mutant clones to expand, and so influences (lengthens) the waiting time to cancer (Martens et al. 2011). As provided in Table 1, premalignant epithelial clones expand within squamous tissue, such as epidermal, cervical, squamous oral, and esophageal mucosa by basal replacement of neighboring stem cells (Clayton et al. 2007; Klein et al. 2010; Doupé et al. 2012; Alcolea et al. 2014) or within glandular tissue via a process of crypt bifurcation termed crypt fission, such as that which occurs in intestinal metaplasia (IM) of the stomach and Barrett’s esophagus (Greaves et al. 2006; McDonald et al. 2008; Nicholson et al. 2012). In these two different epithelia, clonal expansion involves different levels of selection: in squamous epithelium selection is acting at the level of individual cells, whereas in glands selection acts on the meta-population of the glands themselves. In squamous lesions, it is reasonable to believe that cells that are best able to adhere to the basement membrane will have increased survival and this phenotype will be positively selected. For glands, it is uncertain whether the ability to force crypt fission or the ability to survive environmental insult (such as acid/peptic digestion in the stomach) is the trait under selection (McDonald et al. 2015a). As in many ecological systems, there is likely an evolutionary trade-off occurring during phenotype selection—for example, increased proliferative potential at the expense of decreased offspring survival.

Architectural constraints during expansions affect both the size and types of premalignant clones that are positively selected for in a neoplasm. Clones compete for space in a complex process termed “clonal interference” (Garcia et al. 1999; Baker et al. 2013; Streichan et al. 2014; Greaves 2015). Because of minimal information on mutational event timing and spatial pervasiveness in past studies based on single time points (Maley et al. 2004), it is uncertain if mutant clone “sweeps to fixation” occur as a defining feature of premalignant disease. A recent “Big Bang” model of cancer formation illustrates the importance of timing of mutations during evolution as the determinant of a subclone’s pervasiveness in a neoplasm and the relatively rare opportunities for clonal sweeps to occur if at all (Sottoriva et al. 2015). Understanding the scales of units of selection, such as the time to fixation within stem cell niches within colonic and BE crypts, and the rate at which crypts divide, reveal important genotypic–phenotypic relationships in premalignant disease that constrain somatic evolution (Baker et al. 2014; McDonald et al. 2015a,b). From a clinical standpoint, measuring these physical scales of clonal architecture may help us to predict how many covert lesions are currently being missed during screening caused by
imperfect tissue sampling (Reid et al. 2000; Whiting et al. 2002; Tschanz 2005), thus impacting our current interpretations of premalignant versus malignant prevalence that is currently estimated from epidemiological studies, and impacts our prognostication ability (Curtius et al. 2015; Dhawan et al. 2016).

MICROENVIRONMENT WITHIN PREMALIGNANT DISEASE

Tissue architecture and the premalignant microenvironment have a complex interrelationship, which plays a crucial role during natural selection and carcinogenesis. Microenvironmental changes such as fibroblast activation and modulation of the immune microenvironment (de Visser et al. 2006; Grivennikov et al. 2010; Elinav et al. 2013) may alter the fitness effects of (epi)genetic changes being introduced in lesions burdened with both driver and passenger mutations (Mbeunkui and Johann 2009). Microenvironmental pressures on tumor cell evolution can be both promoting (e.g., through the availability of growth factors or predation of inhibitory normal cells) and constraining (e.g., through preventing invasion by construction of an extracellular matrix). The microenvironment also regulates self-renewal in the epithelium (Davis et al. 2015). With respect to the genotype—phenotype map shown in Figure 1, microenvironmental factors act as natural selection pressures on phenotypes (arrow to point 2) that then (because of genetic mutation or phenotypic plasticity) explore the adaptive landscape such that genotypes of future generations reflect the phenotype that successfully surmounted any microenvironmental challenge that may be present (point 3). From this theoretical perspective, Gatenby and Gillies (2008) proposed a model to identify six such barriers, including hypoxia and acidosis in dysplastic epithelium, caused by the microenvironment that must be overcome throughout all stages of carcinogenesis. It is not simply a serendipitous match that a premalignant tissue is found in a particular ecosystem—clonal adaptation can only occur as a result of the current selective microenvironment.

Furthermore, premalignant microenvironments are not closed systems. Extrinsic environmental and lifestyle factors, such as exposure to cigarette carcinogens, dietary habits, and infections, can have a profound effect on the tissue ecosystem and dramatically influence carcinogenesis. For example, human papillomavirus (HPV) high-risk strains cause most cervical intraepithelial preinvasive lesions (Barron et al. 2014) and Helicobacter pylori infection aids in the initiation of IM of the stomach (Yakirevich and Resnick 2013). Nonsteroidal anti-inflammatory drug (NSAID) use, which likely modulates the inflammatory microenvironment, was shown to reduce risk of progression to EAC in patients with BE by slowing clonal evolution via reducing the acquisition rate of somatic genetic alterations (Kostadinov et al. 2013). Gastroesophageal reflux disease (GERD) increases the risk of BE progression to EAC, which is interestingly inversely related to EAC risk attributed to H. pylori infection (Anderson et al. 2008). Mathematical modeling of EAC progression suggests that age-dependent GERD symptoms can affect not only BE incidence but also proliferation rates of HGD during progression to EAC (Hazelton et al. 2015).

Recently, a study showed that the histological changes associated with GERD in esophageal tissue might be explained by a delayed inflammatory response prompted by secretion of cytokine proteins rather than immediate caustic injury caused by acid exposure (Dunbar et al. 2016). Immune cells of course play an important role in the premalignant microenvironment too. In patients with UC and Crohn’s disease, collectively called IBD, intestinal inflammation causes extensive ulceration from altered interactions between resident microbes and the bowel mucosa (Xavier and Podolsky 2007). IBD patients are at higher risk of CRC, and the severity of inflammation is an independent predictor of cancer risk (Rutter et al. 2004). As a final example, consistent changes in stromal gene expression and inflammatory pathways in precursor microenvironments have been found to co-occur across a variety of different gastrointestinal cancers (Saadi et al. 2010).
From an (epi)genetic standpoint, the preconditioning for cancer progression in a large area of tissue that resulted from a colonization of cells with a selected mutation, also known as “field cancerization,” may in fact be a common feature in epithelial carcinogenesis (Garcia et al. 1999; Baker et al. 2013). In both cancer progressors and nonprogressors, large, stable clones have been found that consist of cells with loss of heterozygosity (LOH) for chromosomes 9p and 17p in BE (Buas et al. 2014; Li et al. 2014) and those with distinct CpG island methylation signatures in colorectal adenomas (Luo et al. 2014.). As mentioned previously, a field effect caused by early common mutations in TP53 have also been found in BE segments (Stachler et al. 2015), and similarly in patients with UC (Leedham et al. 2009; Galandiuk et al. 2012) and lung cancer patients (Franklin et al. 1997; de Bruin et al. 2014; Pipinikas et al. 2014). Overall, the concept of monoclonal displacement of normal tissue in such a pervasive sense provokes important clinical questions whether antitumor treatments must aim to eliminate the entire clonal field of preneoplastic cells to prevent recurrence.

IMPLICATIONS FOR PROGNOSTICATION AND PATIENT MANAGEMENT

Characterizations and diagnoses of premalignant lesions have profound effects on cancer incidence and patient care. In particular, there has been an intensive effort to discover prognostic “biomarkers,” defined as measured attributes of a lesion that are associated with increased cancer risk, to aid in patient management (particularly to prevent overdiagnosis of cancer risk). Conventional pathologic grading/staging that describes tissue morphology remains the most common biomarker, and epidemiological studies provide estimates of annual subsequent cancer risk found among a patient population diagnosed with a certain premalignant state (see Table 1). These estimates can be low for early, mostly indolent premalignancies like nondysplastic BE (0.25% annual progression) and typically increase for later stages, such as HGD within BE (~6% annual progression) (Spechler 2013). However, for many premalignant phenotypes, there is large variability in published estimates of associated cancer risk, such as the range of 0.15% to 80% for annual risk of squamous carcinoma of the skin in patients with actinic keratosis (AK), caused by factors like interobserver variability, sampling bias, and patient cohort differences (Greaves 2014). In particular, interobserver variability is an issue for nearly every premalignant diagnosis and may differ at different stages of premalignancy (Fischer et al. 2004; Odze 2006; Leja et al. 2013; van den Einden et al. 2013; Osmond et al. 2014). For example, in a study by U.S. pathologists on BE, interobserver agreement was substantial for HGD and cancer (kappa agreement score, $\kappa = 0.65$), but was only fair ($\kappa = 0.32$) and then slight ($\kappa = 0.15$) for LGD and indefinite for dysplasia, respectively (Montgomery 2005). These results highlight the continuous nature of phenotype diagnoses, from normal appearing cells to slightly abnormal to more distinguishable neoplasms, and the difficulty in precisely identifying the “overlapping” features of phenotype space (Fig. 1) even by expert pathologists.

Because of the heterogeneous estimates of typically low rates involved in precancerous evolution, it is difficult to formulate effective surveillance schedules, that is, screen intervals for premalignant patients to have a checkup examination, which are simultaneously cost-effective, which avoid overdiagnosis/overtreatment, and that catch consequential lesions early. Most current guidelines suffer from ad hoc predictions based on mean progression rates seen in specific patient study cohorts that do not account for interindividual heterogeneity. Nonetheless, the primary goal of cancer screening has been to use this information to our advantage: if we look for and find cancer-risk signatures early in patients, we can reduce cancer mortality by performing an intervention, thus reducing the probability of malignancy arising in the future.

For precancers with long latent periods before cancer initiates, such as adenomas in the colon and cervical intraepithelial neoplasia (CIN), screening and subsequent interventions are very effective and have substantially de-
creased colon and cervical cancer incidences (Barken et al. 2012; Zauber et al. 2012). In current public health policy, a more or less “one-size-fits-all” approach using population average risk estimates has been used in natural history Markov models to assess optimal screening ages and intervals to help inform screening guidelines (Winawer et al. 2006; de Koning et al. 2014). Additionally, other mathematical models for optimizing screening schedules take a more evolutionary approach by incorporating stochastic premalignant and malignant clonal expansions explicitly (Jeon et al. 2008; Hanin and Pavlova 2013; Curtius et al. 2015; Kroep et al. 2017). Although case-control studies aim to identify “who” benefits most from treatment during a set surveillance screen schedule, the goal of this modeling is to predict “when” and at what intervals should premalignant patients return to the clinic during follow-up. This relies on determining when (detectable) precursor lesions develop as we age and how long they dwell in a certain detectable disease state. Mathematical modeling of the stochastic evolutionary process can help to derive and to optimize the timing of clinical screens so that the probability is maximal that an individual is screened within a certain “window of opportunity” for intervention when early cancer development may be observed. By using the available estimates collected from epidemiological studies with long-term patient follow-up, in silico approaches aid in screening design and can perform useful cost-effectiveness analyses to compare proposed screening and intervention strategies.

In other cases, fast-growing premalignancies will be missed by screening (“underdiagnosis”) whereas patients found with slow-growing, or indolent, precursors undergo screening that suffers from length time bias (“overdiagnosis”) and thus will likely continue with costly surveillance programs until death by another cause (Esserman et al. 2013; Reid and Paulson 2015). In the case of breast cancer, a microsimulation model found that annual cost of mammographies in the United States to detect DCIS lesions and early cancers could be reduced by $8 billion, twice the annual budget of the NCI in 2014, if screening was performed biannually (as the U.S. Preventative Services Task Force now suggests) rather than annually (O’Donoghue et al. 2014). The balancing act between benefits and harms of screening can be difficult for a charged issue such as breast cancer screening, and is only complicated with the fact that issues of cost are less likely to be discussed between a physician and patient (Fox et al. 2012). Specifically, nearly all patients diagnosed with DCIS during a screen will undergo surgery followed by radiotherapy. Although a subset of 20% to 50% of those patients will progress to invasive breast cancer if left untreated (Cowell et al. 2013), the remaining patients will foster solely preinvasive phenotypes for their lifetimes. Thus, current medical recommendations seek to achieve a balance of cancer risk averted with compromised quality of life (e.g., owing to treatment costs and associated morbidities). Beyond screening for frank premalignant lesion existence, the long-term aim of cancer prevention is to “risk stratify” patients, that is, to accurately identify each patient as having low or high probability of progression to cancer—and at present the identification of premalignant phenotypes is the sole measure of risk. As per previous sections, we note that the definition of premalignant disease itself feeds into the debate of the goal of risk stratification—if all human cells are evolving to be more cancerized by the accumulation of somatic genetic alterations as we age, we are all part of the “at risk” population and thus further care must be taken in defining the premalignant prerequisites of having a consequential disease before administering certain treatments.

Beyond conventional pathologic staging, increasingly significant risk stratification is made possible based on additional morphological detail, such as premalignant lesion shape in colitis patients (Choi et al. 2015) and size (Anaparthy et al. 2013), and further still with the addition of genetic, epigenetic, and copy number markers correlated with malignant progression. A review by Kinzler and colleagues provides examples of recent studies that show the ability of genomic profiling to provide insight into premalignancy and application to cancer prevention (Kinzler
et al. 2016). The idea that traditional markers such as dysplasia alone are sufficient to predict patient-specific risk is gradually becoming antiquated (Reid et al. 2010). Even a single genetic biomarker’s predictive power suffers from inter-observer variability, as seen in a retrospective study of invasive breast cancers for precursor human epidermal growth factor receptor 2 (HER2) amplification (Mamoon et al. 2014), and overdiagnosis, as was seen in the example that ubiquitous TP53 mutations are found in healthy skin yet few people will ever develop melanoma or basal cell carcinoma (Martincorena et al. 2015). Thus, satisfactory risk stratification may require a panel of biomarkers for each cancer type, and also further for each microenvironmental context and host-dependent setting, in order to increase sensitivity and specificity in determining which premalignancies are most probable to initiate malignancy. However, we again encounter the problem of correctly identifying the genotype–phenotype map (Fig. 1). Once patient-specific phenotypic, genotypic, microenvironmental, and clinical details are gathered at a screening time-point, we must predict the most probable future trajectory toward malignant phenotypes. How do we quantify if a patient’s phenotype and concurrent genotype determined at a screen is “approaching” a malignant phenotype?

As discussed in previous sections, tissue composition at all stages of carcinogenesis is extremely heterogeneous so one major limitation in current screening practice is sampling bias, and thus important premalignant clones may be missed. In effort to reconcile this problem, the analysis of tissue from multiple biopsies taken in a systematic way for use in both histology (Reid et al. 2000) and genetic analyses (Siegmund and Shibata 2016) helps in properly diagnosing pathologic grade and distinguishing between public (clonal) and private (subclonal) genetic mutations used in phylogenetic analyses. Furthermore, as discussed above, biomarkers based on quantifying the evolutionary process itself, such as measures of genetic diversity (Malet et al. 2006), may have pan-tissue prognostic value and be robust to spatial sampling (Dhawan et al. 2016).

In future designs of case-control studies, it seems likely that the most useful biomarker panel to assess patient-specific risk will be one that not only identifies worrisome phenotypes but also quantifies timescales and features of apparent premalignant evolution. Although we all accrue some cancer risk factors in our lifetime, three out of four of us will die of causes other than cancer. Thus, perhaps, the goal of risk stratification should be to identify patients with premalignant cells that are most quickly becoming closer to producing cancerous cells. This will require the analysis of an assemblage of patient-specific factors known to substantially increase cancer risk (e.g., genetic signatures including diversity, abnormal phenotypes, enabling microenvironments, biological rates of aging from molecular clocks) from, ideally, multiple serial time-points with adequate sampling to capture temporal and pervasive features. Additionally, the creation of observational registries regarding those patients who have indolent precancerous lesions (the “controls”) would benefit other patients with low-malignant potential to know more about the evolutionary dynamics of their disease (time-period for risk and type of cancer that may develop) to make informed decisions about treatment, including active surveillance strategies (Esserman et al. 2014). Readily available information produced by these studies for both doctor and patient can help to better identify high-malignant potential lesions (“cases”) while still combatting overdiagnosis by decreasing the number of surveillance screens and/or treatments undertaken by low-risk individuals (“controls”).

**CONCLUDING REMARKS**

How can we characterize premalignancy more precisely to understand the evolutionary process of carcinogenesis? As has been successfully completed for cancerous tumor types, an enticing initiative is to create a “Precancer Genome Atlas” (PCGA) by incorporating effort from multiple institutions worldwide to characterize the molecular alterations in premalignant lesions but importantly also the corresponding changes in the microenvironment and critically...
also the cellular phenotypes associated with progression to cancer (Campbell et al. 2016). As informative as the PCGA may be for pre-cancer characterization, we have herein discussed the evidence that there is an abundance of gray areas in genotype and phenotype spaces; tissues diagnosed as cancerous, precancerous, and normal can share both genetic abnormalities and phenotypic traits. Because of the fact that cross-sectional studies of genetic alterations assayed at a single time point are insufficient to reliably infer important clonal evolution events over time, it will take concerted effort to understand the context-specific nature of cancer progression and the intricacies of the genotype–phenotype map by means of crucial longitudinal studies in the future. Thus, key issues in the evolution of cancer from premalignant disease include (1) determining temporal pattern of mutation accumulation in ostensibly normal tissue and premalignant disease, (2) teasing apart the role of microenvironment as an independent or concomitant driver of disease evolution, and (3) characterizing the host-specific attributes in general, including germline genetic history, immune system activity, microbiome, and other microenvironmental factors. Clearly, solutions to these would have tremendous value clinically as a basis for a prognostic test to evaluate cancer risk.

Do we all in fact develop precancerous clones as we age? Our current understanding of evolution suggests this is probable, as evidenced by a large autopsy study that found that as many as 15% of men with a “normal” prostate-specific antigen level (the placebo group) had prostate cancer (Thompson et al. 2004). If premalignant clones are indeed ubiquitous, perhaps the sheer mention of having “precancer” often causes undue fear followed by overly aggressive treatment of patients in current medical practice. Ultimately, multidisciplinary effort across pathology, imaging, surgical, computational biology, genetics, and medical communities has the ability to broaden and deepen our understanding of premalignant evolution and to ultimately reduce cancer mortality by more effectively managing patients identified as having premalignant disease.

From an evolutionary standpoint, our current definition of what constitutes a premalignant cell (phenotype, [epi]genotype, and/or micro-environmental factors that incur cancer risk) must be revamped to be of greater clinical value because all cells in the body are increasingly becoming “closer to cancer” as we age.

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