Review

The biology of personalized cancer medicine: Facing individual complexities underlying hallmark capabilities

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ABSTRACT

It is a time of great promise and expectation for the applications of knowledge about mechanisms of cancer toward more effective and enduring therapies for human disease. Conceptualizations such as the hallmarks of cancer are providing an organizing principle with which to distill and rationalize the abject complexities of cancer phenotypes and genotypes across the spectrum of the human disease. A countervailing reality, however, involves the variable and often transitory responses to most mechanism-based targeted therapies, returning full circle to the complexity, arguing that the unique biology and genetics of a patient’s tumor will in the future necessarily need to be incorporated into the decisions about optimal treatment strategies, the frontier of personalized cancer medicine. This perspective highlights considerations, metrics, and methods that may prove instrumental in charting the landscape of evaluating individual tumors so to better inform diagnosis, prognosis, and therapy. Integral to the consideration is remarkable heterogeneity and variability, evidently embedded in cancer cells, but likely also in the cell types composing the supportive and interactive stroma of the tumor microenvironment (e.g., leukocytes and fibroblasts), whose diversity in form, regulation, function, and abundance may prove to rival that of the cancer cells themselves. By comprehensively interrogating both parenchyma and stroma of patients’ cancers with a suite of parametric tools, the promise of mechanism-based therapy may truly be realized.

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1. Introduction

The complexity and variability of cancer as a disease has long been recognized. It is manifested in the dramatically different time courses of disease inception, progression and pathogenic impact that are evident across the gamut of tissues susceptible to aberrant proliferative growths. Thus, it is well established that solid tumors (and diffuse proliferations of leukemias) differ dramatically from one organ to another, as well as from one transformed cell type to another that are susceptible to cancer within a particular tissue or organ. This variability is unambiguous at multiple levels: genetically, chromosomally, histologically, physiologically, pathologically, and in terms of prognosis. Different
oncogenic mutations drive particular cellular proliferations leading to given cancer types, and different tumor suppressors are characteristically lost in function, often by distinctive mechanisms. Tumors in different organs can therefore look much different in terms of gross- and histo-pathology, and have dramatically different effects on the patient. As knowledge about cancers at all of these levels increased in the 1980s and 1990s, building an encyclopedia of dizzying complexity and variability, the conceptual landscape has been increasingly obscured by this overwhelming wealth of detail and difference. The hallmarks of cancer elaborated by Hanahan and Weinberg (2000) presented an organize principle from which to consider this complexity, as differing ways to a similar end: the proposition was that most all cancers acquire, by one means or another, a similar set of capabilities necessary to manifest malignant disease. In the recently updated conceptualization (Hanahan and Weinberg, 2011), there are six core and two emerging hallmark capabilities: sustaining proliferative signaling, evading growth suppressors, resisting cell death, enabling replicative immortality, inducing angiogenesis, and activating invasion and metastasis, along with deregulating cellular energetics (and metabolism), and avoiding immune destruction. Virtually all tumors evidence these capabilities, albeit to varying degrees. Thus, one imagines, as an individual tumor develops, that it faces a series of barriers and roadblocks that must be surmounted or circumvented, some of which are tissue specific, thus requiring tissue-specific solutions if cancer cells are to grow profusely and progress to malignancy and other pathologic states. But even in facing the same challenge, different tumors may solve the problem in different ways, developing distinctive mutations and adaptations that allow it to acquire these 6–8 capabilities (Hanahan and Weinberg, 2011). Therefore, the notion arises that the ostensibly confusing individual variability underlays a profound commonality, of acquiring the same hallmark capabilities. The variable means of acquiring such functional capabilities are embedded in two enabling characteristics that each has intrinsic variability — genome instability with (random but genetically selected) mutation, and tumor-promoting inflammation; both demonstrably facilitate acquisition of hallmark capabilities while at the same time being inherently diverse in their manifestation (Hanahan and Weinberg, 2011). It is apparent that the cancer hallmarks conceptualization has proved to be a useful heuristic tool for the cancer research community, with its overarching view of the landscape, albeit one that is viewed from a high altitude with consequently low resolution mechanistically for a given tumor type and subtype, and for individual tumors therein.

In the decade since the introduction of the hallmarks concept (Hanahan and Weinberg, 2000), the encyclopedia of cancer has continued to expand in scope and detail, adding more facts and more depth to the molecular genetic and histopathologic descriptions of cancers in organs throughout the body. Adding to that complexity is the increasing awareness that there is remarkable variability even within a given tumor type, above and beyond the differences in that tumor type from other types. Thus we are now in an era of molecular subtypes, where particular tumor types can be subdivided into molecular subtypes with significantly different features, ranging from signature oncogenic mutations to genome-wide transcriptional profiles, to histo-pathological subtleties (for example in inflammatory cell profiles). Concurrently, highly heralded targeted therapies have been introduced into clinical practice, with the anticipation that such mechanism-based drugs would significantly improve prognosis (Sawyers, 2004; Weinstein, 2002). Yet, although cancer treatments have shown incremental efficacy over the last two decades, the overall survival rate of patients with advanced and/or metastatic cancer (including cancers of the prostate or breast, lung, and colon-rectum, which are responsible for approximately half of cancer incidence in the United States) remains extremely poor. The sobering reality is that the efficacy of most targeted therapies is transitory for most patients, with almost inevitable relapse to progressive disease (de Bono and Ashworth, 2010). Moreover, a common pattern is emerging: targeted therapies are only producing appreciable therapeutic responses in a fraction of patients of a particular type, and even within a specific subtype. Some tumors, while envisioned to respond, do not, and instead continue to grow and progress to states of heightened malignancy. Yet others, often a rare fraction, show enduring responses (de Bono and Ashworth, 2010).

Thus, despite the overarching clarity imparted by the hallmarks conceptualization, we are back full circle to facing the stark reality of patient/tumor heterogeneity. The net result is that we will need to understand the specific complexity and variability of tumors in the context of the individual, so as to determine, based on the characteristics of their disease, how to best treat it so as to produce beneficial responses and enduring efficacy. This realization has opened the frontier of personalized cancer medicine, where individual difference will not be ignored but rather recognized and characterized, and then leveraged for the benefit of that patient, providing better advice on prognosis and therapeutic options, in particular identifying drugs and treatment regimens (and combinations) that are likely to produce beneficial responses for that person in particular. A number of recent reviews and commentaries have discussed aspects of this new frontier (de Bono and Ashworth, 2010; Haber et al., 2011; Luo et al., 2009; Martini et al., 2011), as do articles in the current special issue of Molecular Oncology. Herein we reflect further on the cancer biology of personalized cancer medicine, posing questions and suggesting approaches anchored nevertheless in the notion of the hallmarks of cancer as an integrative tool that may help address the challenges of personalized cancer medicine.

2. Variable responses to targeted therapies reveal and motivate personalized cancer medicine

By targeting molecular pathways that are both specific and essential to a given cancer type or subtype, targeted therapies have been envisioned to provide more effective and less toxic treatment options than conventional therapies such as chemotherapy and radiotherapy. The concept of targeted therapies (‘magic bullets’) raised high expectations, which were initially fueled by the clinical efficacy of imatinib in chronic myeloid leukemia (CML) patients (Druker et al., 1996, 2006;
during therapy molecular characteristics (and evidently, the necessity) to match these drugs with the
et al., 2011; Sawyers, 2009). It is an emerging theme that the
evidently circumvent tumor dependency on HER2 signal-
that the majority of these tumors are not strictly dependent
positive tumors respond to trastuzumab therapy, indicating
overexpression, either by gene amplification or markedly in-
prescribed for patients whose breast cancer cells show HER2
disease recurrence and a worse prognosis (Esteva et al.,
expression in breast cancer is associated with increased
cancer cells
and multi-layered control of proliferative signaling pathways
in cancer cells—such that tumors only rarely depend on
a single regulatory pathway for growth and survival. Such re-
dundancies may underlay the lack of response in some pa-
patients whose tumors express oncogenic targets and
therefore were treated with targeted drugs in anticipation of
therapeutic benefit, and yet there was none.
The existence of intrinsic resistance despite signature on-
cogenic mutations is exemplified by the clinical experience
with HER2-positive breast cancer. HER2 (ERBB2) is an epider-
mal growth factor receptor (EGFR) family protein whose over-
expression in breast cancer is associated with increased
disease recurrence and a worse prognosis (Esteva et al.,
Trastuzumab, an anti-HER2 monoclonal antibody, is
prescribed for patients whose breast cancer cells show HER2
overexpression, either by gene amplification or markedly in-
creased protein level. However, only 30% of such HER2-
positive tumors respond to trastuzumab therapy, indicating
that the majority of these tumors are not strictly dependent
on HER2-mediated signaling for their growth. Many non-
responsive HER2-positive breast tumors harbor mutations
that evidently circumvent tumor dependency on HER2 signal-
and mediate refractoriness to trastuzumab. These include
parallel activation of insulin growth factor 1 (IGF1) receptor
signaling, overexpression of EGF family ligands, or mutations
that activate the PI3K pathway; in fact, PI3K pathway activa-
ion in HER2-positive breast cancer provides a biomarker
to identify patients who are unlikely to respond to trastuzumab
therapy (Esteva et al., 2010).

Similarly, only 10% of metastatic colorectal cancer (mCRC)
patients respond to cetuximab, a monoclonal antibody that
blocks the EGFR (ERBB1), which is broadly expressed in
mCRC (Walther et al., 2009). Both retrospective and prospec-
tive clinical studies have shown that mCRC patients with
KRAS mutations—or mutations in NRAS, Braf, PIK3CA and
HER2—show negligible response to cetuximab, irrespective
of elevated expression of EGFR (De Roock et al., 2010). It should
be noted that both Ras and PI3K lie downstream in the HER2
and EGFR signaling pathways, which may explain why
the lack of constitutively-activated Ras and PI3K pathways is
necessary (albeit not sufficient) for trastuzumab and cetuxi-
mb to be effective.

2.2. Emergence of secondary resistance to growth-
inhibitory drugs variably limits the efficacy of targeted
therapies in different patients

In addition to drug refractoriness ab initio, the emergence of
secondary (or acquired) resistance may limit the duration of
clinical benefit from targeted therapies. Secondary drug resis-
tance occurs in a fraction of imatinib-treated CML patients,
BRAF(V600E) mutation regularly relapse after a window of therapeutic response provided by the
mutant-BRAF inhibitor, vemurafenib (Flaherty et al., 2011); recent data suggest that resistance to vemurafenib may be
mediated—at least in part—by the appearance of aberrantly
spliced BRAF(V600E) isoforms (Poulikakos et al., 2011). Much
like intrinsic resistance, acquired resistance mutations can
also affect “parallel” signaling pathways. For example, 20%
of NSCLC specimens that had developed resistance to gefitinib
or erlotinib show amplification of the MET oncogene
(Engelman et al., 2007). Furthermore, recent studies have an-
notated PIK3CA mutations in NSCLC patients that developed
resistance to these TKIs (Sequist et al., 2011; Turke et al.,
2010). Thus, both drug-resistant target mutations and indi-
rect, target-circumventing resistance mutations can under-
mine the long-term efficacy of the targeted drugs.

In summary, both intrinsic redundancy in proliferative sig-
naling pathways and therapy-induced selection of drug-
resistant mutations evidently limit the efficacy of targeted
therapies. Defining the complexity of activated proliferative
signals—both at diagnosis and during therapy—in individual
tumors carrying a common and druggable oncogenic target is
therefore required to predict and harness the potential of such
targeted therapies. Furthermore, monitoring the early emer-
gence of acquired mutations during the response phase may
help to identify the most suitable drug for subsequent treat-
ment(s) aimed to elicit long-lasting cancer control.

3. Addressing tumor heterogeneity and complexity

Despite the simplicity imparted by the hallmarks conceptual-
ization and the availability of targeted drugs that can interfere
with several of the hallmark capabilities of cancer, the
remarkable diversity, discussed above, in therapeutic responses
among patients with the same cancer presents a need for
deeper knowledge about the biology and genetics of an
individual’s tumor. Although distinctive genetic or epigenetic signatures reflecting particular natural histories of tumorigenesis and progression undoubtedly bear upon the broad spectrum of tumor phenotypes (and variable responses to therapies) observed amongst individuals with the same tumor type or subtype, it is evident that such phenotypic variation is also impacted by several other factors. Thus, in addition to inter-tumoral (or patient specific) variation, significant intra-tumoral variation is evident in many tumor types, in forms spanning histological to genetic heterogeneity. Such variability may be reflective of regional foci of clonal evolution (Yachida et al., 2010), as well as tumor self-seeding from disseminated cancer cells (Comen et al., 2011). Moreover, the heterogeneity of putative tumor-initiating cells (or cancer stem cells) may contribute to subtype variation in certain solid tumor types, as suggested by studies employing mouse models of glioblastoma, intestinal, and prostate cancer (Visvader, 2011). Additionally, specific traits of the accessory cells that make up the tumor-supportive stroma may influence tumor phenotypes and response to therapy (see below). Collectively, such intra-tumoral variation—often reflecting the existence of multiple subclonal tumor populations—might represent an important yet unrecognized determinant for the emergence of secondary drug resistance.

Addressing the complexity and variability of individual tumors and the identification of clinically useful biomarkers will therefore be fundamental to determine how to best treat cancer patients and produce responses of enduring efficacy. Furthermore, the emergence of secondary resistance after initially effective tumor responses calls for repeated tumor profiling analysis during the course of therapy. These realizations motivate the development and implementation of tractable methodology for efficiently auditing the specific complexities of cancer patients, both via tumor biopsies and various forms of blood-borne tumor-derived exfoliates (Figure 1). We highlight below potentially applicable technologies for unveiling patient-specific characteristics so as to guide, optimize, and manage personalized therapeutic regimens.

3.1. Cancer genomes

Cancer genome sequencing efforts have revealed the remarkable genetic complexity of individual cancers, while raising expectations for rapid advances in cancer classification and diagnosis, prognosis and therapeutic design thanks to the identification of “gene signatures” that recur in certain cancer types.

The Cancer Genome Atlas, a catalogue of the DNA mutations found in various human tumors, is revealing that hundreds of coding genes may be mutated in each cancer type/subtype. This figure is suspected to be an underestimation of the mutational landscape of individual tumors, as the analytic techniques that have been used most frequently, such as PCR-based sequencing of DNA extracted from total tumor lysates, may not detect mutations occurring in subclonal tumor cell populations (Loeb, 2011; Sellers, 2011). This is illustrated, for example, by exomic sequencing of separate, histologically distinct regions micro-dissected from individual human pancreatic ductal carcinomas (PDACs): each showed a specific mutational repertoire indicative of multiple subclonal populations within the same primary tumor (Yachida et al., 2010). Consistent with these findings, recent whole-genome deep (or high-throughput) sequencing analysis of individual tumors has revealed that a particular tumor specimen may contain up to 50,000 different non-synonymous mutations affecting several hundred genes, with mutation rates of 0.5–20 per megabase (Loeb, 2011; Sellers, 2011). As opposed to exomic sequencing, unbiased whole-genome deep sequencing should in principle be able to detect complex genetic rearrangements, insertions and deletions, which may enact pivotal gain- and loss-of-function driver events not manifested by point mutations, as shown recently in prostate cancer (Berger et al., 2011). The remarkable number of mutations found in cancer genomes supports the notion that cancers express a “mutator phenotype” that is an enabling characteristic for acquisition of hallmark capabilities, caused for example by loss-of-function mutations in genes that maintain genomic integrity (Hanahan and Weinberg, 2011; Loeb, 2011).

Whole-genome deep sequencing, ideally of multiple regions spanning a tumor, holds promise to markedly improve the annotation of the mutational landscape of individual tumors. Although the aforementioned subclonal mutations in focal tumor regions are unlikely to provide growth advantage to the tumor at large (and thus unlikely to be clinically useful targets of first-line therapy), they might convert into alternative driver mutations upon functional blockade of the original driver mutation, and thus contribute to relapse concomitant with drug resistance. It remains to be ascertained, for example by re-sequencing tumors before therapy and following therapeutic resistance, whether deep sequencing can identify such latent mutations and potentially predict the likely forms of adaptive resistance, thereby guiding second-line treatment strategies.

3.2. Cancer epigenomes

The past decade has witnessed an explosion of knowledge about the epigenetic regulatory changes that occur in cancer cells, leading to the concept that cancer is also a disease of the “epigenome” (Baylin and Jones, 2011). Specific changes in histone acetylation and methylation patterns are being observed in various cancer types. DNA methylation and other covalent, somatically heritable chromatin modifications are also being found. These epigenetic changes are evidently relevant to gene expression and may have an important role in driving tumorigenesis. Excessive DNA methylation of normally unmethylated gene promoter regions is observed in many cancer types, leading to suppression of gene function; for example, the tumor suppressor gene CDKN2A is frequently inactivated by DNA hypermethylation in various tumor types (Baylin and Jones, 2011). DNA hypermethylation and consequent gene promoter inactivation may also decrease the expression of microRNAs (miRNAs, or miRs) that function to finely tune proliferative signaling in normal cells (see below).

Specific DNA hypermethylation signatures have emerged as potential biomarkers to assess cancer risk, aid early detection, and predict therapeutic responses, at least for some cancer types (Baylin and Jones, 2011). Furthermore, there is
increasing evidence that certain epigenetic marks may associate with drug resistance. A recent study showed proof-of-concept that cancer cells may employ a dynamic survival strategy to protect the population from eradication by potentially lethal drugs (Sharma et al., 2010). According to this model, individual cancer cells transiently assume a reversible, drug-tolerant state that requires the activity of histone demethylases; remarkably, the drug-tolerant cancer cells can be selectively ablated by treatment with histone deacetylase (HDAC) inhibitors. Thus, "epigenetic therapy" using inhibitors of HDACs or DNA methyltransferases could be used in combination with targeted drugs to improve their efficacy and/or limit the emergence of secondary resistance. Of note, two DNA methyltransferase inhibitors, azacitidine and decitabine, have been approved recently for myelodysplastic syndrome (MDS) and are currently being tested in acute myeloid leukemia (AML) patients (Baylin and Jones, 2011).

Importantly, the discussion thus far has focused on the cancer cells, and largely ignored the other constituent cell types composing the tumor microenvironment. It can be envisioned that epigenetic changes will prove important also in defining regulatory states in the tumor stroma, for example in the cancer associated fibroblasts (CAFs) that are evidently reprogrammed (or educated) to adopt stable aberrant phenotypes.

3.3. Cancer transcriptomes

Genome-wide transcriptional analysis of the expressed mRNAs, initially using microarray technologies and more...
recently RNA deep sequencing, has been widely used to audit cancer cell genomes in whole tumor lysates from both human tumors and mouse models of human cancers. Increasingly sophisticated bioinformatics tools have been applied to such databases, revealing gene signatures of coordinately up- and down-regulated gene transcripts that correlate with and can thereby identify particular tumor subtypes and stages of progression, and in some cases predict prognosis (Golub et al., 1999; Quackenbush, 2006; Reis-Filho and Pusztai, 2011; Sotiriou and Piccart, 2007, 2009; van’t Veer and Bernards, 2008; Walther et al., 2009). Moreover, cross-filtering of cognate mouse and human tumor transcriptome databases is proving useful in reducing complexity and potentially identifying functionally significant alterations from amongst the “white noise” of non-specific consequential variation (Ding et al., 2011; Zender and Lowe, 2008).

An added dimension to this approach has come from assessing alterations in the expression profile of non-coding RNAs, micro-RNAs in particular. miRNAs are small RNAs that finely tune the expression of multiple target genes; altered miRNA expression has been observed in a variety of human pathological conditions, including cancer (Bartel, 2009). The function of miRNAs may be deregulated in cancer in several manners, including miRNA gene mutation, deletion or promoter hypermethylation, but also as a consequence of mutations in their target sequences (Esquela-Kerscher and Slack, 2006; Nana-Sinkam andCroce, 2011; Palmero et al., 2011). Although attenuated expression of the miRNA transcriptome is frequently observed (Kumar et al., 2007; Lu et al., 2005), various miRNAs are overexpressed in specific tumors compared to their healthy tissue of origin (Ma et al., 2010a,b; Volinia et al., 2006), and thus are termed “oncomiRs”. Furthermore, the predicted targets for several cancer-associated miRNAs include protein-coding tumor suppressors and oncogenes, supporting a role for miRNAs in cancer pathogenesis (Esquela-Kerscher and Slack, 2006; Nana-Sinkam and Croce, 2011; Palmero et al., 2011). The rapid implementation of miRNA profiling techniques (such as miRNA SAGE, microarray and deep-sequencing) has helped in defining “miRNA gene signatures” for many cancer types, and several miRNAs now represent robust disease-specific predictive and prognostic biomarkers (Kasinski and Slack, 2011). miRNA signatures can also be identified in the peripheral circulation of cancer patients (Mitchell et al., 2008), and several tumor-associated miRNAs have been isolated from circulating microvesicles and exosomes (Rani et al., 2011). Of note, miRNAs are relatively stable compared to mRNAs and can be profiled from paraffin-embedded tissues and body fluids such as blood and urine (Palmero et al., 2011; Wittmann and Jack, 2010).

In addition to cancer cells, miRNAs regulate the biology and functions of diverse tumor-associated stromal cells, including CAFs, tumor endothelial cells, and inflammatory/immune cells (Anand and Cheresh, 2011; Aprileikova and Green, 2012; O’Connell et al., 2010; Schetter et al., 2010). There is also evidence that miRNAs can be horizontally transferred between tumor and stromal cells via microvesicles or exosomes (Simpson et al., 2009; Valadi et al., 2007; Yang et al., 2011). It can be envisioned that both miRNA and protein-coding RNA profiling of fractionated constituent cells of the tumor stroma will prove instructive about tumor transcriptomes ranging from prognosis to potential intrinsic resistance mechanisms.

### 3.4 Cancer proteomes

Whereas DNA sequencing and microarray techniques may help identifying mutated or misregulated genes that could represent druggable targets, mass spectrometry-based cancer proteomics may enable the identification of tumor-derived proteins that can serve as biomarkers of disease and response to therapy (Hanash and Taguchi, 2010; Ludwig and Weinstein, 2005). Indeed, noninvasive proteomic analysis of body fluids (particularly serum or plasma) has identified potential biomarker signatures for several types of cancers, including asymptomatic or radiologically undetectable tumors. However, proteomic-based biomarker discovery faces major technical hurdles that will need to be surmounted, namely: (i) the low abundance of most protein biomarkers in blood and other body fluids, particularly in early stage tumors; (ii) the difficulty to establish the relevance of a putative biomarker for tumor biology in experimental models; and (iii) the difficulty to validate the relevance of candidate biomarkers in well-controlled clinical trials (Liotta and Petricoin, 2011; Sawyers, 2008).

Nevertheless, by way of illustrating the potential of this technology, it is worth considering a recent study that investigated the utility of mass-spectrometry to identify plasma protein gene signatures reflective of pathways driving the development of specific lung tumor subtypes, both in human subjects and mouse models of the disease (Taguchi et al., 2011). To this aim, the authors employed three NSCLC models that harbor human-disease causing mutations (e.g., EGFR and KRAS/p53 mutations), and compared proteomic profiles of such lung cancer models with those of mouse models of unrelated tumors or inflammation. Notably, model-specific protein signatures were identified, such as an EGFR signature that was specific to the EGFR mutant mouse model. Importantly, selected protein signatures could be validated in human lung cancer cell lines as well as sera from yet asymptomatic lung cancer patients (Taguchi et al., 2011).

Thus, together with highly sensitive and noninvasive imaging techniques, blood-based cancer biomarkers may improve early diagnosis of certain cancer subtypes, inform on therapy options, or enable early discovery of secondary resistance in treated patients. Toward that end, better-standardized procedures for collecting and analyzing biological samples will be required to increase the reach of proteomics research and establish its ultimate clinical utility. In this regard, isolation of blood circulating exosomes might provide a mean to enrich for tumor-derived proteins (Simpson et al., 2009). It remains to be seen, however, whether the sensitivity and bandwidth of proteomic profiling will be sufficient to interrogate patient-specific differences that could guide individualized treatment strategies.

The challenge now is to integrate the increasing wealth of information provided by RNA and DNA profiling, as well as proteomics, to increase our understanding of cancer heterogeneity and devise personalized therapies that can effectively exploit such information. A cross-disciplinary systems biology approach will be needed to translate the information contained in multidimensional data sets into molecular pathways.
and networks that may represent potential targets for therapy or useful predictive and prognostic biomarkers in individual cancer patients.

4. Variability and dynamics of stromal cell components

Much of the frontier of personalized cancer medicine is being charted from the perspective of the cancer cell and its individual patient-specific alterations. Yet, as already alluded above, there is good reason to predict that heterologous cell types recruited to assemble the tumor microenvironment will also prove to be significantly variable in their abundance and functional characteristics, in ostensibly similar tumors of the same organ types, and even within subtypes and stages of progression. It is well established that multiple cell types and subtypes beyond the cancer cells themselves variable in regards to differentiation, epithelial-to-mesenchymal (EMT) status, and stem cell phenotypes contribute to the pathologic manifestation of tumors, including endothelial cells and pericytes of the tumor vasculature, CAFs, and a multitude of hematopoietic (immune inflammatory) cell types (Figure 1), amongst which are both tumor-promoting and tumor-antagonizing constituents (Hanahan and Weinberg, 2011). We comment below on the major classes of stromal cells in the tumor microenvironment in regard to knowledge and implications about their biology that is relevant to the concept of personalized cancer medicine.

4.1. Tumor blood vessels

Both human and mouse tumors show considerable variation in the density, morphology and functionality of blood vessels (Langenkamp and Molema, 2009; Nagy et al., 2010). Furthermore, the endothelial cells of tumor blood vessels can display features fenestration patterns, pericyte and inflammatory-cell association, proliferation and apoptosis rates, and gene expression profiles that vary not only among different tumor types or individual tumors, but also in a local–regional manner within a given tumor. Such heterogeneity may be influenced by the site in which a tumor arises (i.e., the specific organ or tissue microenvironment), the tumor growth stage, the biophysical properties of the surrounding stroma (e.g., interstitial pressure, collagen cross-linking and extra-cellular matrix tension), and many other ill-defined spatiotemporal differences such as angiogenic gene expression by tumor and stromal cells (Carmeliet and Jain, 2011a; Chung and Ferrara, 2011; Hanahan and Weinberg, 2011; Kerbel, 2008; Leite de Oliveira et al., 2011; McDonald and Choyke, 2003; Nagy et al., 2010; Potente et al., 2011; Weis and Cheresh, 2011). As a consequence of these many variables affecting vascular phenotypes, experimental tumors growing subcutaneously in mice may differ significantly from spontaneous tumors in terms of vascular density, functionality, phenotype, and gene expression. For example, human PDACs are characteristically poorly vascularized, with a prevalence of non-perfused blood vessels. These features are not recapitulated by subcutaneously growing PDACs, which lack the abundant desmoplastic stroma and high interstitial pressure that limits blood vessel functionality in autochthonous (or orthotopic) tumors. Even orthotopic xenotransplant PDAC tumors growing in the pancreas are hypervascular in comparison to genetically engineered mouse models (GEMMs) and human PDACs (Olive et al., 2009); for other tumor types, however, orthotopic transplants may more accurately reflect the bona fide angiogenic and stromal components of the tumor microenvironment, and as such present an important experimental alternative to subcutaneous tumor xenografts (Francia et al., 2011).

Although human tumors display type-specific vascular patterns, our understanding of such variability is mostly limited to histo-pathological features (Langenkamp and Molema, 2009; Nagy et al., 2010). Different human cancers are variably sensitive to antiangiogenic drugs (e.g., inhibitors of endothelial TK receptors or antibodies that block vascular growth factors); furthermore, tumors of a generally refractory type may sporadically show dramatic responses to antiangiogenic drugs (Carmeliet and Jain, 2011b; Chung et al., 2010; Goel et al., 2011; Leite de Oliveira et al., 2011). It can be argued that tumor type or subtype-specific aspects of the biology of associated endothelial cells (e.g., gene expression programs) may play a role in determining such different responses. Studies in GEMMs of cancer have begun elucidating the molecular bases of such variation. It has emerged, for example, that certain tumor types are less dependent on VEGF (and hence resistant to VEGF blockade) than others, as they can use noncanonical (e.g., non-sprouting) modalities of vascular growth or alternative proangiogenic pathways for sprouting angiogenesis (Abdullah and Perez-Soler, 2011; Bergers and Hanahan, 2008; Leite de Oliveira et al., 2011). For instance, some tumors can grow along pre-existing blood vessels without evoking an angiogenic response (and indeed are refractory to VEGF blockade); this process is referred to as vascular co-option (Carmeliet and Jain, 2011a) and is specially observed in well-vascularized tissues such as the brain (Holash et al., 1999). Other tumors may display vasculogenic mimicry, a process by which tumor cells alter their gene expression profile toward an undifferentiated phenotype, and gain the ability to form vascular-like structures that do not depend on VEGF for their growth (Maniotis et al., 1999; Soda et al., 2011). If such vascular subtleties are not recapitulated in transplant tumor models representing a particular organ-specific cancer, then targeted antiangiogenic drugs may fail to accurately demonstrate their effects and limitations. Additionally, there is clinical evidence for intrinsic resistance (Bergers and Hanahan, 2008) to pharmacological inhibition of the predominant VEGF signaling pathway, seen for example in the progression-free and overall survival plots of renal cancer patients receiving the VEGF pathway inhibitors bevacizumab (Yang et al., 2003), sunitinib (Motzer et al., 2007) and sorafenib (Escudier et al., 2007), wherein some ostensibly similar patients do not respond at all to the antiangiogenic therapy, and rather continue progressing, apparently refractory to the anti-angiogenic therapies (Albiges et al., 2011; Garcia et al., 2010; Rini et al., 2008).

GEMMs of cancer, and potentially orthotopic patient-derived xenotransplants (PDXs; see below), may serve as instructive experimental tools for auditing the molecular biology of tumor-associated blood vessels in the context of...
and gene expression signatures in vascular cells. Anti-VEGF therapy, and call for auditing genetic variation while lacking mechanistic details, may help explaining the 2009; Schneider et al., 2008). These provocative observations, chemo- and anti-VEGF therapy (bevacizumab) (Jain et al., were shown to predict tumor responses and overall survival of human tumors. Obtained, either micro-dissection of fixed tissues or fluorescence-activated cell sorting of fresh tumor biopsies, could be analyzed using the aforementioned techniques, both retrospectively and prospectively from tumors showing variable degrees of response to antiangiogenic therapy, in order to identify molecular signatures that associate with better responses.

Technological advances in laser capture microscopy and in expression profiling by genome wide RNA-sequencing from paraffin-embedded formalin fixed tissues are beginning to render this approach feasible. Furthermore, it is expected that the advancing capabilities of genetic tools for mapping polymorphisms in the human genome (and that of mice) will allow the genes comprising the various regulatory networks controlling angiogenesis to be audited for polymorphic variations, which might affect angiogenic phenotypes by altering the strength and integration of signals impacting the state of the system. Indeed it is emerging that polymorphic variation in relevant antiangiogenic targets may provide useful predictive biomarkers of response to therapy. For example, there is proving to be considerable genetic variability within the VEGF gene, evidenced by multiple single nucleotide polymorphisms (SNPs). Notably, certain VEGF SNPs were shown to predict tumor responses and overall survival of metastatic breast cancer patients receiving combination of chemo- and anti-VEGF therapy (bevacizumab) (Jain et al., 2009; Schneider et al., 2008). These provocative observations, while lacking mechanistic details, may help explaining the puzzling therapeutic heterogeneity of patient's responses to anti-VEGF therapy, and call for auditing genetic variation and gene expression signatures in vascular cells.

### 4.2. Fibroblasts

Cancer-associated fibroblasts represent another important stromal cell population in most tumor types (Kalluri and Zeisberg, 2006; Tlsty and Coussens, 2006; Pietras and Ostman, 2010). While often referred to generically as CAFs, there are evidently a number of distinctive fibroblastic cell types encompassed by this designation, including recruited myofibroblasts expressing smooth muscle actin, and activated (“reactive”) resident tissue fibroblasts (Pietras and Ostman, 2010). The functionality of CAFs has been demonstrated for example by co-injection of CAFs together with tumor cells, which enhances tumor growth by promoting ECM synthesis and stiffening, inducing angiogenesis, and recruiting growth-promoting inflammatory cells such as macrophages (Erez et al., 2010; Kalluri and Zeisberg, 2006; Orimo et al., 2005; Tlsty and Coussens, 2006). CAFs are particularly abundant in certain human cancers, such as PDAC, and in various carcinomas at advanced stages of progression (e.g., breast and colorectal cancer). In such tumors, CAFs produce copious fibrotic tissue — referred to as “desmoplastic stroma” — particularly at the boundaries between the invasive cancer and the host tissue. The presence of desmoplastic stroma is associated with enhanced tumor cell invasiveness and is a major determinant of tumor cell malignancy (Egeblad et al., 2010; Levental et al., 2009). The high interstitial pressure measured in the desmoplastic stroma limits tumor blood vessel perfusion and the delivery of chemo and other forms of therapy. Furthermore, CAFs may also mediate resistance to anti-VEGF therapy, likely via their production of stromal-cell derived factor-1 (SDF1, or CXCL12) — a potent chemotactic signal for proangiogenic myeloid cells (Orimo et al., 2005) — or by directly stimulating angiogenesis via their secretion of platelet-derived growth factor-C (PDGF-C) (Crawford et al., 2009) and basic fibroblast growth factor (FGF2) (Pietras et al., 2008).

As noted above, CAFs are likely to comprise distinct sub-populations, whose relative abundance may vary in different tumor types (Kalluri and Zeisberg, 2006; Pietras and Ostman, 2010). However, little is known about potential patient-specific CAF heterogeneity in human tumors of the same type/subtype. Does their composition and “education/induction/recruitment” by the cancer cells and the tumor microenvironment vary in functionally significant ways across cohorts of individuals with the same cancer type and subtype? And, do their resultant gene expression signatures and functions vary, or might constitutional genetic polymorphisms affect their involvement? Elucidating the prevalence of CAF variability within tumor types and subtypes could reveal important determinants, new individualized therapeutic targets, and predictive biomarkers for tumor responses to anticancer therapies.

### 4.3. Leukocytes

Leukocytes make up a large proportion of the tumor-associated stromal cells in most human and mouse tumor types (de Visser et al., 2006). Subsets of these cells, such as CD8+ cytotoxic T-cells and natural killer (NK) cells, may play roles in restricting tumor development and progressive growth (de Visser et al., 2006; Dunn et al., 2006). On the other hand, a growing body of research in mouse cancer models has now implicated multiple leukocyte species as causal players in cancer initiation and progression. For example, macrophages and granulocytes are known to generate an “inflammatory” tumor microenvironment that promotes angiogenesis and ECM remodeling, enhances tumor cell motility and invasion, and suppresses antitumor adaptive immune responses (Mantovani et al., 2008; Motz and Coukos, 2011; Murdoch et al., 2008; Qian and Pollard, 2010; Squadratto and De Palma, 2011). For these reasons, tumor-promoting inflammation is now regarded as an enabling characteristic of cancer (Hanahan and Weinberg, 2011).

It is increasingly evident that variable degrees of infiltration by tumor antagonizing inflammatory cells, principally CD8+ cytotoxic T-cells, can be prognostic for outcome amongst patients with ostensibly similar tumors in terms of type and grade, with greater infiltration predicting better outcome (Fridman et al., 2011; Galon et al., 2006; Koebel et al.,...
On the other hand, the relative abundance of immunosuppressive T-regulatory cells (Tregs) (Bates et al., 2006; Pages et al., 2010) or CD4+ effector T-cells (Denardo et al., 2011; Kohrt et al., 2005; Zhou et al., 2009) can be prognostic of comparably poor outcome, as can the abundance of various myeloid cell types, including macrophages, neutrophils and immature myeloid cells (Qian and Pollard, 2010; Steidl et al., 2011; Zhou et al., 2009), a subset of which are referred to as myeloid-derived suppressor cells on account of their ability to inhibit T-cell functions (Sica and Bronte, 2007). Furthermore, studies in mouse models of cancer have revealed that profuse myeloid cell infiltration correlates with resistance to antiangiogenic therapies targeting the VEGF-signaling axis (Bergers and Hanahan, 2008; Ferrara, 2010; Squadrito and De Palma, 2011), or accelerated tumor re-growth following local tumor irradiation (Ahn and Brown, 2008; Kioi et al., 2010; Kozin et al., 2010). Additionally, there is considerable phenotypic and functional heterogeneity among macrophage and neutrophil subtypes, most simply reflected in the amalgam of conventionally or alternatively activated macrophages found in tumors, as well as in inflamed or healing tissues (Biswas and Mantovani, 2010; Coffelt et al., 2010; Gordon and Martinez, 2010; Nucera et al., 2011; Piccard et al., 2011). While less advanced in regard to appreciating individual variability, one can envision that the abundance and characteristics of these and other types of inflammatory cells will indeed prove to be deterministic and hence important parameters for personalized diagnosis. Moreover, there is increasing evidence that the bioeffect functions and hence pro- versus anti-tumor activities of both lymphoid and myeloid cell types can vary significantly in a organ- and tumor type dependent manner, as suggested by studies employing sophisticated mouse models of cancer engineered to develop in distinctive immunological backgrounds (Andreu et al., 2010; Clampicotti et al., 2011; Daniel et al., 2005; Denardo et al., 2009). Such organ- and tumor type peculiarities of infiltrating leukocytes may underlay the puzzling controversy surrounding the general prognostic value of macrophage and neutrophil infiltration in human tumors (Azambuja et al., 2011; Qualls and Murray, 2011; Steidl et al., 2011). Moreover, apparent contradictions in prognostic value between studies in the same tumor type may reflect differences in the inflammatory dynamics in tumors of distinctive subtype, grade, stage, and size, as well as patient-to-patient variability. These complexities call for better methods with which to audit the complexity of tumor-associated inflammatory cells in mouse and human. With the advent and continuing refinement of automated flow cytometry and tissue immunohistochemistry processing devices enabling 8–10 antigens to be audited in parallel, one can envision that tumor biopsies will in the future be screened to assess their particular constitution of inflammatory cell types and subtypes in a comprehensive and standardized manner. Such knowledge may possibly contribute to informed decisions about treatment strategies. The premise of this approach is exemplified by a recent study that analyzed multiple leukocyte populations in a large cohort of breast cancer patients treated by surgery alone, which identified “leukocyte signatures” — i.e. different compositions of infiltrating leukocytes in individual tumors — that can predict the patient’s likelihood to respond favorably to adjuvant chemotherapy (Denardo et al., 2011). In particular, the combination of high macrophage and low cytotoxic T-cell abundance — but not of either cell population alone — identifies a subset of patients at high risk for developing chemoresistant, metastatic disease. These intriguing findings call for the systematic analysis of leukocyte infiltrates in tumors of different types and subtypes. Defining the molecular profiles of “high-risk leukocytes” (either by immunophenotyping or gene expression profiling) may also enable the validation of novel tumor biomarkers. Furthermore, specific leukocyte signatures might be used to stratify patients carrying defined genetic lesions, in order to explore how such signatures correlate with the individual patient’s response to the targeted therapy and the emergence of secondary resistance. If such studies demonstrate the prognostic value of specific leukocyte signatures for individual tumor responses to targeted therapies, then further characterization of leukocyte complexity and function in mouse tumor models may contribute toward the development of more effective, fully personalized cancer therapies, that for example target the inflammatory component of the tumor by reprogramming leukocyte infiltrates from a tumor-promoting to an anti-tumor function (Hagemann et al., 2008; Squadrito and De Palma, 2011; Stout et al., 2009).

Finally, it should be mentioned that certain targeted drugs may indirectly activate inflammatory cells to modulate therapeutic responses in both mouse tumor models and cancer patients (Beatty et al., 2011). For example, the anti-HER2 antibody trastuzumab is thought to mediate tumor regression by both interrupting oncogenic signaling in tumor cells and inducing immunoglobulin (IgG) receptor (FcR)-mediated cytotoxicity (Park et al., 2010). Indeed, FcR receptors expressed on the surface of certain inflammatory cells can recognize IgG-coated tumor cells, and the interaction between FcRs and IgGs can in turn activate the inflammatory cells to release cytotoxic molecules that kill the tumor cells. The therapeutic effect of trastuzumab is in fact decreased in the absence of FcR signaling in mice (Clynes et al., 2000), and specific FcR polymorphisms are associated with clinical outcome in breast cancer patients (Musolino et al., 2008). These interesting observations also imply that SNPs in genes expressed by the stromal components of the tumor may influence tumor responses to targeted therapies.

**4.4. Population dynamics of stromal cell constituents**

The above depictions of the phenotypic diversity of stromal subtypes have largely ignored another facet to the challenge of understanding the heterogeneity of tumor biology: population dynamics. How do individual tumors of a particular subtype vary in the proportional abundance and organization of the various stromal cell types and subtypes? How do population dynamics of an individual’s tumor change during malignant progression and metastasis, and in response to therapy? And, what is the implication of regional variation in stromal cell composition within a solid tumor — another dimension to the mutational subclones discussed above — in terms of prognosis and predicted response or resistance to different therapies? Addressing these questions is an important agenda for the field of personalized cancer medicine.
5. Harnessing mouse models of cancer to investigate individualized therapies recognizing and exploiting tumor heterogeneity in cancer patients

Transplant tumor models – in particular human tumor cell lines that are propagated in vitro and grown in immunodeficient mice as tumor xenografts – have the potential to model important aspects of the individual variability of human cancer, typically the heterogeneity intrinsic to the cancer cells of an individual’s tumor (Sharpless and Depinho, 2006). There are, however, notable limitations to such models in terms of tumor progression and microenvironment. In fact, transplant tumor models do not in general recapitulate several key features of spontaneous, organ-specific tumorigenesis, such as: premalignant to malignant progression; angiogenic switching; acquisition of defined histo-pathological marks; and organ-specific and progression-stage dependent interplay with the immune system and other components of the tumor microenvironment. Because they do not fully recapitulate tumor microenvironmental (and histo-pathological) complexity, transplant tumor models have often failed to predict drug responses (Sharpless and Depinho, 2006). We envision increasing refinement of transplant tumor models, mostly through (i) the use of orthotopic sites of inoculation to generate tumors growing in the parental organs; (ii) the implantation of primary patient-derived material – either tumor fragments or briefly cultured cancer cells – to replicate and propagate individual variability; and (iii) co-inoculation of cancer cells (or primary epithelial cells engineered ex vivo to express defined oncogenes) together with human stromal cell types, e.g., patient-derived fibroblasts. As discussed below, such refinements of transplant tumor models – along with the implementation of both germline and somatic GEMMs – may help in developing tumor models that better recapitulate the features of individual cancers and their responses to targeted therapies.

5.1. Patient-derived xenotransplants (PDXs)

Recent studies have demonstrated that small tumor biopsies inoculated in immunodeficient mice can retain the morphological and genetic heterogeneity of the parental tumor, including individual diversity, even when propagated ectopically, e.g., subcutaneously (Bertotti et al., 2011). This successful recapitulation may be due to the “supportive” role of the parental tumor stroma (primarily CAFs but also vascular and hematopoietic cells), which is co-transplanted with the cancer cells and is possibly propagated with the latter even through a few serial passages. Indeed, small human lung cancer biopsies inoculated subcutaneously in severely immunodeficient NOD-SCID/IL2Rγ null (NSG) mice were found to retain intra-tumoral populations of human T-cells, which could also expand and colonize peripheral lymphoid organs for prolonged periods (Simpson-Abelson et al., 2008). These findings suggest that some stromal cell constituents of non-disrupted human cancer biopsies can engraft long-term in situ and even systemically in PDX models. It remains to be seen whether terminally differentiated human cells such as macrophages or vascular cells behave similarly in such xenografted tumors.

While the subcutaneous site is convenient and often the choice for logistical reasons, most tumors growing subcutaneously lack the full manifestation of the fibroblastic stroma of their parental tumors, and the tumor vasculature is often different, both in quantity and quality of tumor blood vessels, as noted above. Furthermore, subcutaneous PDX models may fail to reflect the patterns of tumor growth and spreading that are observed in cancer patients; and, while many tumor biopsies fail to engraft, others may produce xenografts with large areas of necrotic tissue due to inadequate re-vascularization of the tumor mass. Implantation of such tumor biopsies (or perhaps fractionated cancer cells and admixed CAFs following brief culture without serial passaging) at orthotopic sites may circumvent some of the limitations of ectopic injection, particularly for those tumor types that typically contain abundant fibrotic stroma (e.g., pancreatic and breast cancer) or hematopoietic cell infiltrates (e.g., ovarian tumors). For example, breast cancer biopsies xenografted orthotopically together with primary human mesenchymal cells in NOD-SCID mice retain the diversity and essential features of the original human breast cancer subtypes, including organ-specific (e.g., lung or bone) metastasis (Derose et al., 2011). Furthermore, human ovarian tumors can be successfully established as orthotopic xenografts in NSG mice, recapitulating growth and progression patterns observed in ovarian cancer patients, while retaining critical human stromal cell components such as CAFs and T-cells (Bankert et al., 2011). It is anticipated, however, that human stromal cells will be progressively lost or diluted by mouse-derived cells during tumor growth and serial transplants in either ectopic or orthotopic sites, raising challenges for biobank preservation to allow replication of results and additional analyses.

5.2. Primary cells genetically modified with oncogenic signature mutations

Defined human tumor types may also be recapitulated in mouse models by genetically engineering primary human epithelial cells (Fan et al., 1997; Heyer et al., 2010; Khavari, 2006). This approach was illustrated, e.g., by using mammary (Duss et al., 2007) or epidermal (Fan et al., 1997; Khavari, 2006) epithelial cells obtained from reduction mammaplasties or skin biopsies, respectively. The isolated cells were propagated in vitro and then transduced with integrative viral vectors (e.g., γ-retro or lenti-viruses) that carried the desired combination of signature mutations that are characteristic of the human tumor being modeled. Orthotopic injection of these genetically modified primary epithelial cells, particularly when made in combination with human primary stromal cells such as fibroblasts, can generate human tumors that quite faithfully recapitulate the histo-pathological features of genotype-specific human disease (Fan et al., 1997; Heyer et al., 2010; Khavari, 2006). These transplant models have an important advantage over germline GEMMs of cancer, in that they allow for testing of multiple mutations in multiplex combinations, in a time and cost effective manner (Wu and Robinson, 2009).
5.3 Non-germline (somatic) GEMMs of cancer

Recent studies have demonstrated the utility of non-germline (or somatic) GEMMs of cancer. This approach is based on the injection of genetically modified embryonic stem (ES) cells into pre-implantation mouse embryos (Heyer et al., 2010; Zhou et al., 2010). The ES cells are engineered to contain multiple chromosomal integrations of inducible, cell type-specific human signature mutations; the mouse chimeras that are derived from injection of the ES cells into mouse blastocysts can therefore carry multiple cancer-enabling alleles that are expressed only in the cells derived from the genetically modified ES cells, which will constitute a variably small fraction of the total cells in any given tissue. These spontaneous tumor models, as opposed to germline GEMMs, more faithfully model the stochastic nature of tumorigenesis, as the tumor arises sporadically due to the chimeric nature of the cancer-enabled cells. The non-germline nature of such models obviously prohibits propagation of the mice, but allows for generating libraries of ES cells that carry oncogenes relevant to human disease and that can be conveniently stored (Heyer et al., 2010). Tumors that arise in such chimeric mice can be serially transplanted to develop secondary cohorts of mice for experimental therapeutic trials. However, the potential immunogenicity of human oncogenic products typically requires the secondary transplants to be propagated in immunodeficient mice, in order to avoid interference or abject rejection by the adaptive immune system.

5.4 Limitations of transplant tumor models: immunological deficiencies and cross-species incompatibilities

An unavoidable limitation of the aforementioned transplant tumor models lies in the immunodeficiency of the transplant host, leaving the inflammatory response seriously biased, invariably lacking T- and B-lymphocytes, which amongst their repertoire include not only tumor antagonizing but also tumor promoting subtypes. This will remain an important and perhaps unsolvable qualification to xenotransplant models, which argues for parallel efforts toward continuing refinement of GEMMs of cancer that can reflect aspects of individual tumor heterogeneity. Another limitation of xenotransplant models is that human tumor cells may fail to appropriately respond to mouse-derived signals (e.g., growth factors and hormones) because of cross-species biological incompatibility. This was shown, for example, in a study demonstrating that human breast cancer cells inoculated in mice, while expressing prolactin receptors, are insensitive to mouse prolactin and are selected in vivo for growth independence of circulating prolactin (Utama et al., 2006). These findings suggest that the biology (and consequent drug-response profiles) of human cancers propagated in mice may be influenced by the mouse (endocrine) environment.

A strategy to circumvent some of the limitations of xenotransplantation while preserving the flexibility of transplant tumor model may involve the injection of genetically modified mouse cells, including lineage-restricted stem/progenitor cells, orthotopically into syngeneic, immunocompetent mice (Bachoo et al., 2002; Heyer et al., 2010). This "mouse-in-mouse" transplant model has the advantage over human xenotransplant models to expose tumor development to the same immunological barriers that are present in GEMMs of cancer, albeit without the genetic heterogeneity that distinguishes otherwise apparently similar human tumors.

5.5 Germline GEMMs of cancer

Conventional (or germline) GEMMs of cancer undeniably provide a powerful means to study tumor biology and response to therapy (Hanahan et al., 2007; van Miltenburg and Jonkers, 2012). Such tumor models can in many cases recapitulate the histo-pathological and genetic features of their human disease counterparts, originating from "normal cells" that reside in tissue- or organ-specific microenvironments and are subjected, at least in part, to the biological barriers to tumorigenesis that are erected by the particular normal tissue microenvironment. GEMMs of cancer can be generated to harbor the specific molecular lesions that are causally associated with human cancer. However, tissue-specific expression of oncoproteins may induce epithelial transformation throughout the targeted tissue, often represented by diffuse hyperplasias that may cause significant morbidity and even mortality in mice (Heyer et al., 2010; Zhou et al., 2010). The implementation of inducible systems in which signature mutations are activated only in a small proportion of the cells of a defined tissue and/or in a temporally defined manner (e.g., by tamoxifen-inducible CRE recombinases that are delivered topically or activated by suboptimal doses of tamoxifen) may better recapitulate the stochastic nature of tumorigenesis (Jonkers and Berns, 2002; Meylan et al., 2009). Yet, even the most sophisticated GEMMs of cancer — such as mice that carry multiple genetic lesions reflecting a defined cancer gene signature — may lack the individual variation that is commonly observed within each cancer type/subtype. This aspect can be potentially addressed by employing outbred strains of mice, or mice that are prone to genetic instability due to deficiencies in the genes that maintain genomic integrity.

Another limitation of GEMMs of cancer is that such models often show a low incidence and/or narrow spectrum of visceral metastases. Because metastatic disease represents the ultimate target of anticancer therapies, preclinical models that can recapitulate organ-specific metastasis should be developed and employed to more realistically assay the therapeutic potential of candidate drugs. Orthotopic transplants (including PDX models) may enhance the incidence of distant metastatic spread compared with subcutaneous transplants (Derose et al., 2011), especially if the primary tumor is surgically removed to allow the development of macroscopic metastases from the disseminated tumor cells (Francia et al., 2011). Highly metastatic human tumor variants can also be generated through multiple rounds of in vivo growth of tumor cell lines in immunodeficient mice, each involving orthotopic tumor cell transplantation/growth/resection, and isolation of metastatic cells from the visceral organs (Francia et al., 2011; Kerbel et al., 1984). This approach has been used to select highly metastatic tumor cell lines that may be representative of various human cancer types/subtypes (Francia et al., 2011); in some cases, genetic modification of the tumor cells was
used to introduce relevant driver oncogenic mutations in the
tumor cells (du Manoir et al., 2006). However, this approach
is not easily applicable to the derivation of patient-specific tu-
mor cell lines that can inform on treatment options for the in-
dividual, as the in vivo selection procedure is laborious and
lengthy (Francia et al., 2011).

In sum, we view the future of personalized cancer research
to incorporate both germline and somatic GEMMs in concert
with refined (preferably orthotopic) PDX models that will
each be selectively informative about the biology and genetics
of variation amongst individual tumors of a given type.

6. Concluding remarks

The importance of individual diversity in the biology of cancer
amongst patients with ostensibly similar tumors (and in par-
ticular with common driver mutations) is compellingly
revealed in the variable responses and even intrinsic resis-
tance to driver- and mechanism-targeted therapies, arguing
that we need to develop a deeper understanding of their dif-
fences, so as to fine-tune therapies to be efficacious in
each individual by recognizing particular drug sensitivities
as well as the evident insensitivities.

Thus, for example, the identification of EGFR and ALK mu-
tations in NSCLC and their association with sensitivity to EGFR
and ALK TKIs is revolutionizing the treatment of this cancer
(Gaughan and Costa, 2011). In fact, treatment strategy can
now be decided on a truly individual basis, and will produce
beneficial responses in the majority of the patients carrying
the respective mutation. Although most patients eventually
develop secondary resistance and relapse, re-biopsy can es-
tablish the molecular basis of resistance in some cases, allow-
ing physicians to offer further treatment options, e.g.,
irreversible EGFR TKIs (Kwak et al., 2005), which are a suitable
option for half of the relapsing patients. The NSCLC experi-
ence thus offers a compelling case for refining clinical trial de-
signs so as to identify patient sub-populations that are most
likely to benefit from the therapy. Of course, even in this tu-
mor type there is much to be learned, about how to best treat
those patients lacking such driver mutations; perhaps deeper
mining of the cancer cell and of the tumor stroma using the
involving armamentarium of analytic technologies combined
with advancing knowledge will reveal functionally important
molecular and cellular (stromal) targets amenable to thera-
petic modulation to the benefit of particular patients.

The grand challenge now is to generalize and expand this
categorical awakening across the spectrum of human can-
cers, exploiting the complexity and variability within a partic-
ular type/subtype to therapeutic benefit, rather than failing
because of it. The heterogeneity of cancer cells (and of cancer stem cells) as well as of stromal components — tumor vascu-
arization, tumor-promoting and tumor-antagonizing immu-
ne cell infiltrates, CAF subtypes, etc... — should be interro-
gated through molecular screening, immunohisto-
chemistry, better noninvasive imaging, and also careful ex-
amination of re-biopsy specimens obtained during
treatment as well as at treatment failure (Figure 1). Quantita-
tive immunohistochemistry or immunofluorescence assays
are being developed that can accurately validate on tissue
microarrays (TMAs) the expression of DNA/RNA signatures
identified by whole-tumor profiling; these techniques have
the advantage of being highly reproducible and easily applica-
able for standardized tumor typing/grading, inter-patient com-
parisons, and the dissection of intra-tumoral heterogeneity.
Recent applications of such TMA-based technologies have val-
Idated prognostic gene signatures for several tumor types, in-
cluding prostate and lung cancer (Anagnostou et al., 2011;
Dimou et al., 2011; Ding et al., 2011; Zender and Lowe, 2008).
Furthermore, the development of sophisticated computer-
base methodologies for TMA analysis (Beck et al., 2011)
may facilitate better standardization of the results and
also quantitatively assess clinically significant morphometric
parameters (e.g., epithelial–stromal ratio, multiple nuclear
pleomorphisms, etc...) that would escape conventional patho-
logic examination.

It has been proposed that a synchronized “GEMM-to-hu-
man” strategy should be implemented to accelerate preclini-
cal and clinical data collection and promote rationally
designed clinical trials that address individual variation and
secondary resistance (Nardella et al., 2011). According to this
strategy, each phase I/II clinical trial should be conducted in
parallel with appropriate (and possibly diversified) GEMMs
of cancer, so that relevant information obtained from patients
and mice is integrated to facilitate patient stratification and
identification of biomarkers of disease and response to ther-
apy; among the correlative metrics that can be envisioned
are analysis of genome mutational status and constitutional
polymorphic variation, transcriptome and proteome profiles,
effects of the therapeutic regimen(s), noninvasive imaging of
responses, and assessment of relapse consequent to adaptive
resistance. This strategy will likely be complemented by ana-
lyzing the spectrum of tumor phenotypes in cohorts PDX mice
reflecting the diversity of particular human cancers (Bankert
et al., 2011; Derose et al., 2011). In the future, PDX models
may evolve to involve severely immunodeficient mice (e.g.,
NSG mice) engineered to express defined human cytokines
and histocompatibility genes along with engrafted human he-
matopoietic stem cells, so as to recapitulate and sustain upon
serial passage more authentic tumor microenvironments in-
volving both tumor-antagonizing and tumor-promoting im-
mune inflammatory cells of human origin. Data obtained
from GEMMs and PDX models of cancer and from patient bi-
opies and blood borne materials (proteins, microRNA, exo-
somes, circulating tumor-derived cells) hold promise to
inform hypothesis-testing and personalized “GEMM/PDX-to-
human” trials that may address individual variation in tumor
responses to the drug(s) and/or mechanisms of acquired re-
sistance. Such coordinated therapeutic trials in mouse
models and humans, designed to recognize, reflect, and lever-
age patient-to-patient variability, are envisioned to help pio-
neer the frontier of personalized cancer medicine toward
more broadly efficacious therapies for human cancers.

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