Profilig Critical Cancer Gene Mutations in Clinical Tumor Samples

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Abstract

Background: Detection of critical cancer gene mutations in clinical tumor specimens may predict patient outcomes and inform treatment options; however, high-throughput mutation profiling remains underdeveloped as a diagnostic approach. We report the implementation of a genotyping and validation algorithm that enables robust tumor mutation profiling in the clinical setting.

Methodology: We developed and implemented an optimized mutation profiling platform (“OncoMap”) to interrogate ~400 mutations in 33 known oncogenes and tumor suppressors, many of which are known to predict response or resistance to targeted therapies. The performance of OncoMap was analyzed using DNA derived from both frozen and FFPE clinical material in a diverse set of cancer types. A subsequent in-depth analysis was conducted on histologically and clinically annotated pediatric gliomas. The sensitivity and specificity of OncoMap were 93.8% and 100% in fresh frozen tissue; and 89.3% and 99.4% in FFPE-derived DNA. We detected known mutations at the expected frequencies in common cancers, as well as novel mutations in adult and pediatric cancers that are likely to predict heightened response or resistance to existing or developmental cancer therapies. OncoMap profiles also support a new molecular stratification of pediatric low-grade gliomas based on BRAF mutations that may have immediate clinical impact.

Conclusions: Our results demonstrate the clinical feasibility of high-throughput mutation profiling to query a large panel of “actionable” cancer gene mutations. In the future, this type of approach may be incorporated into both cancer epidemiologic studies and clinical decision making to specify the use of many targeted anticancer agents.

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Introduction

Many tumors contain hallmark mutations within oncogenes or tumor suppressor (TS) genes that may confer a heightened susceptibility to targeted anticancer therapies. Well-established examples include KIT mutations in gastrointestinal stromal tumors (GISTs) that predict response to imatinib or nilotinib, and non-small cell lung cancers with EGFR mutations that are sensitive to erlotinib[1,2,3]. The presence of other mutations predicts a lack of response to targeted therapy. For example, lung and colorectal cancers that harbor mutations in the KRAS oncogene are unresponsive to treatment with anti-EGFR agents[4], and inactivating
Patients and Tumor Tissue Collection

Anonymized tumor specimens were obtained from the Cooperative Human Tissue Network (CHTN), Surgical Oncology University of Siena, Italy, and Dana-Farber Cancer Institute; human glioma samples were obtained from the clinical archives of the departments of Pathology at Children’s Hospital Boston and Brigham and Women’s Hospital. (The required tumor content was >70%; necrosis <10%.) Institutional review board (IRB) exemption was obtained for all samples from the Dana-Farber/Partners Cancer Care Office for the Protection of Research Subjects. DNA extraction was performed as described in Methods S1.

OncoMap Assay Design and Genotyping

Selection of cancer gene mutations for assay design and mass spectrometric genotyping were performed as previously described[6] with modifications indicated in Methods S1. Assay, primer and probe sequences are indicated in Table S1.

Sample Barcoding and Sequencing

Primers flanking KRAS codon 12 were used to PCR-amplify genomic DNA from 91 fresh frozen and 93 FFPE samples (primer sequences indicated in Table S2). PCR amplicons were analyzed by Sanger sequencing. Alternatively, KRAS codon 12 was PCR-amplified with barcoded primers as described in Methods S1. PCR amplicons were pooled and analyzed (Illumina Genome Analyzer II; single lane). Sequence data was analyzed as described in Methods S1.

Results

Characteristics of Clinical Tumor Samples

A total of 903 clinical tumor specimens derived from 12 different tissue sites were assessed for this study (Table 1). The majority of samples were adenocarcinomas (n = 625) from tumors of the breast, lung, prostate, and GI tract. A collection of GISTs (n = 34) and gliomas (n = 155) were also included. Of these, 643 specimens were obtained from fresh frozen tissue and 260 derived from FFPE blocks. Estimated tumor content exceeded 70% in all samples as measured by pathological review (see Methods).

Performance of the Tumor Mutation Profiling Algorithm

To facilitate cancer gene mutation profiling in clinical tumor specimens, we developed OncoMap, a panel of genotyping assays that assessed 396 unique mutations in 33 cancer genes. The complete mutation profiling algorithm (Figure 1A) involved mass spectrometric genotyping followed by both automated calling and manual review to generate a list of candidate mutations. These candidates were subjected to secondary genotyping validation (see Methods S1). Assumming that all primary profiling and assay validation reagents are in place, 7–10 days are required to complete the entire OncoMap sequence.

Table 1. Human tumor samples (fresh frozen and FFPE) investigated in this study.

| Sample Type | Total Number of Samples Tested | Number of fresh frozen vs (FFPE) samples | Total Number of Samples with Mutation | Percentage of samples with mutation |
|-------------|--------------------------------|---------------------------------------|-------------------------------------|-----------------------------------|
| Brain       | 155                            | 0 (155)                               | 44                                  | 28.40%                            |
| Breast      | 53                             | 20 (33)                               | 21                                  | 39.60%                            |
| Colon       | 159                            | 113 (46)                              | 102                                 | 64.20%                            |
| Endometrium | 23                             | 23 (0)                                | 18                                  | 78.30%                            |
| Esophagus   | 117                            | 117 (0)                               | 29                                  | 24.80%                            |
| Gastric     | 233                            | 233 (0)                               | 90                                  | 38.20%                            |
| Kidney      | 26                             | 26 (0)                                | 1                                   | 3.80%                             |
| Lung        | 26                             | 0 (26)                                | 15                                  | 57.70%                            |
| Ovary       | 9                              | 9 (0)                                 | 1                                   | 11.10%                            |
| Prostate    | 95                             | 95 (0)                                | 10                                  | 10.50%                            |
| Thyroid     | 7                              | 7 (0)                                 | 4                                   | 57.10%                            |
| Total Numbers | 903                        | 643 (260)                             | 335                                 | 37.10%                            |

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OncoMap performance was assessed after determining the “ground-truth” mutational status at \textit{KRAS} codon 12 using a DNA barcoding and massively parallel sequencing-by-synthesis strategy (see Methods S1) applied to 91 frozen and 93 FFPE-derived tumor DNAs. When these deep sequencing results were compared to genotyping assays interrogating the same \textit{KRAS} codon, we found the OncoMap sensitivity and specificity to be 93.8% and 100%, respectively, in DNA from fresh/frozen tissue; and 89.3% and 99.4% in FFPE-derived DNA (Figure 1B). In contrast, the sensitivity of conventional Sanger sequencing was 83.3% for fresh frozen tissue and 76.0% for FFPE-derived DNA, confirming the heightened performance of genotyping-based mutation profiling [7].

OncoMap performance was evaluated more extensively by testing 215 fresh frozen tumor samples spanning multiple tumor types in which the mutational status for nucleotides interrogated by 52 OncoMap assays had been established previously. By this analysis, mutation profiling achieved a similarly high specificity of 99.8%. Assay sensitivity was optimal in tumor tissues where the tumor content exceeded 50% (data not shown). Although individual sensitivity values could not be determined for all mutations assayed, calculations using unidirectional \textit{KRAS} assays (reflective of the majority of OncoMap assays) yielded nearly identical sensitivity and specificity values (Figure S1). These results suggested that the genotyping-based mutation profiling platform was suitable for many clinical applications.

“Actionable” Cancer Gene Mutations across Diverse Cancer Types

Of the 903 clinical tumor specimens profiled, 37% (n = 335) contained one or more mutations. In total, 417 mutations were identified, with 63 samples exhibiting co-occurring mutations. Whereas 286 mutations were found by assays interrogating known or candidate oncogenes, 131 mutations were observed in TS genes. Thus, the OncoMap platform provided more extensive mutation information than earlier mutation profiling efforts which focused exclusively on oncogene mutations. As expected, the distribution of mutations reflected patterns previously observed in human tumors, although the frequency of TS mutations was lower (reflective of the reduced coverage of such mutations by OncoMap). Examples include \textit{PIK3CA} (26%) and \textit{TP53} (13%) mutations in breast cancer; \textit{APC} (11%), \textit{BRAF} (10%), \textit{KRAS} (38%), \textit{PIK3CA} (11%) and \textit{TP53} (9%) mutations in colorectal cancer, and \textit{EGFR} (12%), \textit{KRAS} (23%) and \textit{STK11} (8%) mutations in lung cancer. The expected mutation distributions were observed in both fresh frozen and FFPE specimens, underscoring the potential utility of OncoMap in the clinical setting.

Figure 1. The OncoMap process and performance in fresh frozen and FFPE-derived DNA. A. An overview of the OncoMap process from tumor to mutation profile. See text for details. B. Receiver operator characteristic curves (ROCs) show the sensitivity and specificity for various cutoff values on the sample score of the validation samples. ROCs are plotted for fresh frozen (left) and FFPE-derived (right) DNAs, using bidirectional \textit{KRAS} assays and Illumina data as a truth-set (see Methods S1).

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Mutations Predicting Response to Targeted Therapies

Table 2 indicates a set of “actionable” cancer gene mutations identified herein. The OncoMap platform robustly detected mutations that constitute established markers of response to targeted therapies. For example, EGFR mutations predictive of response to erlotinib and gefitinib were identified at the expected frequency (12%) in non-small cell lung cancer, and KIT mutations linked to sensitivity to imatinib and nilotinib were detected in 76% of GISTs. Interestingly, ERBB2 mutations were detected in four gastric adenocarcinomas, raising the possibility of testing a HER-2 inhibitor such as trastuzumab in gastric cancer patients selected based on this genetic criterion [8]. Several tumors harbored mutations that may predict response to investigational agents. For example, BRAF^{V600E} mutations were identified in colorectal (n = 16), ovarian (n = 1), thyroid (n = 4) and endometrial (n = 1) cancers, as well as pediatric gangliogliomas (n = 22; see below). Tumors harboring these mutations may respond to a selective BRAF inhibitor[6,9]. We also identified 96 samples across seven different cancer types [breast n = 14, colorectal n = 24, endometrial n = 15, esophageal n = 4, gastric n = 34, prostate n = 3, and pediatric astrocytoma n = 2] harboring mutations in either PIK3CA or PTEN. These mutations might be expected to enrich for tumors responsive to the PI3 kinase inhibitors currently in development.

Mutations Predicting Resistance to Targeted Therapies

Along with mutations that confer heightened sensitivity to targeted therapies, OncoMap robustly detected mutations associated with resistance to several agents. Established examples include KRAS mutations in non-small cell lung cancer (23%) and colorectal cancer (38%) that confer resistance to erlotinib, gefitinib[12] or cetuximab (colorectal)[4,10,11]. While 94% of KRAS mutations identified localized to codons 12 or 13, 6% occurred elsewhere in the gene (most commonly at codon 61). Since most studies of KRAS-associated resistance have focused exclusively on codons 12 and 13[3,12,13], OncoMap identified additional KRAS mutations that may influence sensitivity to anti-EGFR treatment. OncoMap also identified an HRAS mutation in a lung adenocarcinoma and an NRAS mutation in a colorectal adenocarcinoma. Mutations involving alternate RAS isoforms are rare in these cancer types; conceivably, these might also confer resistance to anti-EGFR therapy.

Mutation profiling also identified mutations that confer “secondary” resistance to targeted therapies (e.g., resistance alleles arising during the course of targeted therapy). In GIST tumors, where 31 KIT mutations were identified in 25 samples; both primary (imatinib-responsive) and secondary (imatinib-resistant) KIT mutations were observed, in keeping with prior mutation profiling studies [7]. In particular, several KIT mutations involving exon 9 (KIT Y503R or F504insAY; 5 cases) were detected in untreated GIST tumors. These mutations are associated with an increased drug requirement to elicit a clinical response [14,15]. A PDGFRα mutation (D842V) predictive of resistance to imatinib [15] was also identified in one GIST sample.

The AKT1^{E17K} mutation has previously been reported in breast, colorectal and ovarian cancer. The OncoMap platform identified rare AKT1 mutations in these tumor types, as well as single AKT1^{E17K} instances in endometrial, esophageal squamous, gastric, and prostate cancers. This mutation may predict resistance to PI3 kinase inhibition (and conceivably receptor tyrosine kinase inhibition) in some contexts [16].

Cancers with Co-Occurring “Actionable” Mutations

The presence of co-occurring mutations involving critical cancer genes may modify the clinical response to single-agent targeted therapy. Here, 20 adenocarcinomas (10 colorectal, two endometrial and 8 gastric) exhibited co-occurring PIK3CA and KRAS mutations. While coincident mutations in these genes have previously been reported in cancers of the large intestine [17], PIK3CA and KRAS mutations have typically exhibited a mutually exclusive pattern of occurrence in endometrial cancer [10,19]. An endometrial adenocarcinoma with co-occurring FGFR2 and PTEN mutations was also identified. Two tumors harbored coincident BRAF and PTEN mutations (one endometrial, one colorectal), and an additional colorectal sample contained both a PIK3CA and a BRAF mutation. These tumors might be expected to exhibit resistance to receptor tyrosine kinase inhibition.

Molecular Classification of Pediatric Brain Tumors by Cancer Gene Mutation Profiling

The ability to perform robust mutation profiling of clinical and archival tumor tissue may promote systemic molecular characterization of “orphan” cancers, including some pediatric tumors where sufficient tissue is often lacking. To explore this hypothesis, OncoMap was used to query a series of pediatric low-grade gliomas (LGGs) whose mutation spectrum is incompletely defined. This analysis included genomic DNA from 127 pediatric and 28 adult gliomas (for comparison) spanning five histologic subtypes of pilocytic astrocytoma, ganglioglioma, and diffuse astrocytoma.

Candidate Prognostic or Therapeutic Targets in Pediatric Brain Tumors

Several potentially “actionable” cancer genes were found mutated in pediatric LGGs. Examples include PDGFRα (n = 1) and PIK3CA (n = 2), which may predict response to existing drugs (e.g., imatinib or nilotinib) or agents in development (e.g., PI3 kinase inhibitors). Two pediatric LGGs harbored mutations in MYC, a well established oncogene homologue of MYC, which is amplified and indicative of poor prognosis in pediatric neuroblastoma [20,21]. Thus, tumor mutation profiling may refine patient stratification and/or disease prognosis in some pediatric brain cancers.

BRAF^{V600E} Mutations Are Common in Pediatric Gangliogliomas

Duplication of the BRAF locus has been reported as the most frequent aberration in pediatric LGAs (66% [18]), whereas activating point mutations in BRAF occur less commonly (4–6%) [22, 23]. We identified activating BRAF^{V600E} mutations in 11% (10/88) of pediatric LGAs—a higher percentage than previously reported [22,23]. Interestingly, the BRAF^{V600E} mutation was most prevalent within the ganglioglioma subtype of pediatric LGAs (classical and non-classical; 8/14 tumors, p = 0.00005), as shown in Fig. 2. BRAF^{V600E} mutations were not identified in any of the adult tumors examined, although TP53 mutations commonly occurred in this setting (10/28 cases). Conversely, only 2 pediatric gliomas harbored a TP53 mutation; in both cases, this was coincident with another mutation (EGFR and FLNB, respectively). These results suggest that OncoMap might facilitate classification of low-grade astrocytomas and other rare tumor types based on genetic criteria.

Discussion

Clinical oncology is in the midst of transition from a treatment paradigm dictated primarily by the anatomic site of tumor origin to one in which genetic and/or molecular characteristics play a decisive role in guiding choice of therapy. Moreover, the
proliferation of targeted agents in development and clinical practice necessitates concomitant implementation of companion diagnostic approaches that enrich for subpopulations most likely to respond to a drug. New diagnostic approaches are therefore needed to profile any tumor for pivotal genetic mutations in multiple cancer genes simultaneously, in contrast to most existing tests that focus on single genes (or proteins).

In this study, we adapted genotyping-based mutation profiling for the characterization of both frozen and FFPE-derived tumor specimens spanning 12 cancer types. We robustly detected cancer gene mutations that direct clinical use and predict resistance to existing agents such as tyrosine kinase inhibitors (e.g., EGFR and KRAS mutations). We also identified multiple mutations that may

| Tissue Type | Gene | Amino acid | N  | %  |
|-------------|------|------------|----|----|
| Brain (astrocytoma, ganglioglioma, glioblastoma) | BRAF | V600E | 22 | 14.2 |
| | EGFR | T263P | 1 | 0.6 |
| | PDGFRA | D842V | 1 | 0.6 |
| | PIK3CA | R88Q | 1 | 0.6 |
| | | H1047R | 1 | 0.6 |
| Breast (lobular and ductal carcinoma) | AKT1 | E17K | 1 | 1.9 |
| | PIK3CA | R88Q | 1 | 1.9 |
| | | N345K | 1 | 1.9 |
| | | E542K | 1 | 1.9 |
| | | E545K | 2 | 3.8 |
| | | H1047R | 9 | 17.0 |
| Colon (adenocarcinoma) | AKT1 | E17K | 1 | 0.6 |
| | BRAF | D594G | 2 | 1.3 |
| | | L597Q | 1 | 0.6 |
| | | V600E | 13 | 8.2 |
| | KRAS | A146T | 1 | 0.6 |
| | | G12A/C/D/S/V | 45 | 28.3 |
| | | G13C/D | 13 | 8.2 |
| | | Q61H/L | 2 | 1.3 |
| | NRAS | Q61K | 1 | 0.6 |
| | PIK3CA | R88Q | 1 | 0.6 |
| | | E542K | 2 | 1.3 |
| | | E545K | 1 | 0.6 |
| | | H1047R | 9 | 17.0 |
| Endometrium (adenocarcinoma) | AKT1 | E17K | 1 | 4.3 |
| | BRAF | G466A | 1 | 4.3 |
| | FGFR2 | S252W | 1 | 4.3 |
| | | C362R | 1 | 4.3 |
| | KRAS | G12A/V | 3 | 13.0 |
| | | G13D | 1 | 4.3 |
| | PIK3CA | R88Q | 1 | 4.3 |
| | | C420R | 1 | 4.3 |
| | | H701P | 1 | 4.3 |
| | | H1047R | 4 | 17.4 |
| | PTEN | R1303G/Q* | 2 | 1.3 |
| | | R233* | 2 | 1.3 |
| | | K267fs*9 | 2 | 1.3 |
| Gastrointestinal tract (GIST) | KIT | Y503_F504insAY | 7 | 21.2 |
| | | W557G | 2 | 6.1 |
| | | W557/K595del | 2 | 6.1 |
| | | W557/V559*C/F | 4 | 12.1 |
| | | V559D | 1 | 3.0 |
| | | V560D | 2 | 6.1 |
| | | V560del | 2 | 6.1 |
| | | L576P | 1 | 3.0 |
| | | K642E | 2 | 6.1 |
| | | V654A | 3 | 9.1 |
| | | D816H | 1 | 3.0 |
| | | N822K | 2 | 6.1 |
| | | Y823D | 1 | 3.0 |
| | PDGFRA | D842V | 1 | 3.0 |
| | Kidney (renal cell carcinoma) | VHL | L89H | 1 | 3.8 |
| | Lung (adenocarcinoma) | EGFR | L858R | 3 | 11.5 |
| | | HRAS | Q61L | 1 | 3.8 |
| | | KRAS | G12C/D | 4 | 15.4 |
| | | Q61H | 2 | 7.7 |
| | Ovary (adenocarcinoma) | BRAF | V600E | 1 | 11.1 |
| | Prostate (adenocarcinoma) | AKT1 | E17K | 1 | 1.1 |
| | | FGFR3 | F384L | 1 | 1.1 |
| | | PIK3CA | H1047R | 1 | 1.1 |
| | | G1049R | 1 | 1.1 |
| | | PTEN | R173H | 1 | 1.1 |
| | Thyroid (papillary carcinoma) | BRAF | V600E | 4 | 57.1 |

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Table 2. Cont.
guide the use of emerging agents. Finally, in a specific investigation of pediatric low-grade astrocytomas, we demonstrated the special value this platform may have for rare “orphan” cancers by identifying mutations that may inform molecular classification as well as new therapeutic avenues for children with these malignancies.

Diagnostic interventions that successfully introduce tumor mutation profiling to clinical practice must circumvent several technical and logistical hurdles. Chief among these is the attainment of robust performance in samples derived from FFPE and/or archival tumor material. We found that the OncoMap platform achieved nearly 100% specificity in both fresh/frozen and FFPE-derived tumor DNA, indicating that false positive mutation calls are likely to be relatively rare with this approach. It should be noted, however, that achieving this level of specificity required the implementation of an analytical sequence in which raw genotyping data is subjected to automated base calling followed by manual review of candidate mutations and validation of all candidates using alternative genotyping chemistries. Thus, clinical implementation of OncoMap must incorporate both genomic data generation and bioinformatic analytical expertise into a molecular pathology or clinical diagnostic setting.

The sensitivity of OncoMap is influenced by inherent technological parameters, individual mutation assay performance characteristics, and the quality and purity of tumor tissue. The 89–94% assay sensitivity observed in this study is sufficient for many translational and clinical applications; however, there are of course circumstances where even higher assay sensitivities will be desirable. Enrichment of tumor cells using core needle dissection or laser-capture microdissection prior to mutation profiling may offer one avenue to enhance sensitivity, particularly in tumors where the stromal or inflammatory content is high. At the same time, the sensitivity of OncoMap vastly exceeds that of Sanger sequencing, which remains the gold standard for many genetic diagnostic approaches. Furthermore, the breadth of cancer genes and specific mutations interrogated by OncoMap—supported by the aforementioned rigorous sensitivity and specificity determinations—substantially exceeds that of existing commercial mass spectrometric-based genomic profiling approaches[24].

Genomically guided therapies may play an especially prominent role in rare tumors where large randomized trials are often impractical. To test the ability of the OncoMap to identify uncommon and/or sample-limited tumors that may benefit from specific classes of therapeutic agents, we profiled a panel of pediatric low-grade gliomas for common cancer gene mutations. Knowledge of the genetic abnormalities present in pediatric LGAs is limited, though recent studies have identified BRAF translocations, chromosomal duplications, and occasional base mutations in low-grade astrocytomas[18,25], as well as diverse mechanisms for activating the ERK/MAPK pathway in pilocytic astrocytomas[23]. These findings suggest that the small molecule inhibitors of BRAF already in adult trials may also represent promising therapeutics for subsets of these tumors. Our results indicate that the frequency of BRAF point mutations in pediatric LGAs as a whole may be higher than previously reported, and specifically that gangliogliomas possess BRAFV600E mutations at very high frequency. This observation may also aid in diagnostic identification of these tumors. Gangliogliomas, which tend to be indolent tumors, may therefore share some properties with cutaneous nevi, whose melanocyte precursors also derive from the nervous system (neural crest), exhibit indolent growth, and where the BRAFV600E mutation is also highly prevalent. We also identified several mutations not previously reported in pediatric astrocytoma, some of which (EGFR, PIK3CA, PDGFRα) represent potentially actionable targets. In concordance with previous reports[22,26,27] we observe that mutations in genes frequently observed in adult anaplastic astrocytomas and/or glioblastomas, such as TP53 or PTEN, are only rarely encountered in pediatric pilocytic and low-grade diffuse astrocytomas.

Although our findings support the clinical feasibility of high-throughput tumor mutation profiling, we recognize that the mass spectrometric genotyping approach has certain limitations that may preclude its implementation as a definitive cancer diagnostics platform. These include the finite number of specific point mutations that can be assayed (designated a priori within a subset of cancer genes), difficulties in designing genotyping assays that identify small insertions or deletions (“in-dels”) larger than ~50bp in size, an inability to detect most TS gene mutations (which may occur anywhere within the gene, not just “hotspot” regions) or additional genomic alterations such as high-level gene amplifications or deletions that may also affect key cancer genes, and the somewhat labor-intensive nature of manual review and orthogonal assay validation. Over the long term, the adaptation of new genomics technologies such as second generation sequencing may offer a unifying approach to comprehensive tumor mutation profiling. However, the OncoMap platform may offer one immediate avenue by which systematic mutation profiling might be initiated to guide clinical trial design as well as use of existing targeted agents across many cancer types.

In summary, this study represents the first large-scale application of the OncoMap platform for tumor mutation profiling in the clinical and archival setting. These results therefore enliven a framework wherein systemic tumor profiling might emerge as a widely feasible means to guide patient stratification for rational cancer therapeutics. Despite the inherent complexity of cancer, incorporating the growing knowledge of the molecular basis of cancer into both large-scale molecular epidemiologic studies and, ultimately, clinical decision making should ultimately speed the advent of more effective anticancer therapies.

Supporting Information

Methods S1 Profiling critical cancer gene mutations in clinical tumor samples.
Illumina data as a truth-set (see Methods S1). DNAs, against unidirectional OncoMap KRAS assays, using plotted for fresh frozen (left panel) and FFPE-derived (right panel) derived DNA. Receiver operator characteristic curves (ROCs) are

Figure S1 Performance of OncoMap in fresh frozen and FFPE-derived DNA. Receiver operator characteristic curves (ROCs) are plotted for fresh frozen (left panel) and FFPE-derived (right panel) DNAs, against unidirectional OncoMap KRAS assays, using Illumina data as a truth-set (see Methods S1). Found at: doi:10.1371/journal.pone.0007887.s003 (0.03 MB PPT)

References

1. Lynch TJ, Bell DW, Sordella R, Gurubhagavatula S, Okimoto RA, et al. (2004) Activating mutations in the epidermal growth factor receptor underling responsiveness of non-small-cell lung cancer to gefitinib. N Engl J Med 350: 2129–2139.
2. Paez JG, Janne PA, Lee JC, Tracy S, Greulich H, et al. (2004) EGFR mutations in lung cancer: correlation with clinical response to gefitinib therapy. Science 304: 1497–1500.
3. Pao W, Miller V, Zakowski M, Doherty J, Papi L, et al. (2004) EGFR receptor gene mutations are common in lung cancers from “never smokers” and are associated with sensitivity of tumors to gefitinib and erlotinib. Proc Natl Acad Sci U S A 101: 13306–13311.
4. Khamabat-Ford S, Garrett CR, Meropol NJ, Basik M, Harbison CT, et al. (2007) Expression of epiregulin and amphiregulin and Kras mutation status predict disease control in metastatic colorectal cancer patients treated with cetuximab. J Clin Oncol 25: 3220–3227.
5. Looupakis F, Pollina L, Stasi I, Ruzzo A, Scartozi M, et al. (2009) PTEN expression and KRAS mutations on primary tumors and metastasis in the prediction of benefit from cetuximab plus irinotecan for patients with metastatic colorectal cancer. J Clin Oncol 27: 2622–2629.
6. Venter JC, Adams MD, Myers EW, Li PW, Mural RJ, et al. (2001) The sequence of the human genome. Science 291: 1304–1351.
7. Thomas RK, Baker AC, DeBiasi RM, Winkler W, Laframboise T, et al. (2007) High-throughput oncogene mutation profiling in human cancer. Nat Genet 39: 347–351.
8. Van Cutsem E, Kang Y, Chiang H, Shen L, Sawaki A, et al. (2009) Efficacy results from the ToGA trial: A phase III study of trastuzumab added to standard chemotherapy (CT) in first-line human epidermal growth factor receptor 2 (HER2)-positive advanced gastric cancer (GC). J Clin Oncol 27: LBA1509.
9. Flaherty K, Puzanov I, Sosman J, Kim K, Ribas A, et al. (2009) Phase I study of PLX4032: Proof of concept for V600E BRAF mutation as a therapeutic target in BRAF V600E mutation-positive metastatic melanoma. J Clin Oncol 27: LBA4509.
10. Schwab M, Altalaloon K, Klopmanauer KH, Varmus HE, Bishop JM, et al. (1983) Amplified DNA with limited homology to myc cellular oncogene is shared by human neuroblastoma cell lines and a neuroblastoma tumour. Nature 305: 245–248.
11. Brodeur GM, Seeger RC, Schwab M, Varmus HE, Bishop JM (1984) Amplification of N-myc in untreated human neuroblastomas correlates with advanced disease stage. Science 224: 1121–1124.
12. Pinter S, Janzarik WG, Remke M, Ernst A, Werlt W, et al. (2008) BRAF gene duplication constitutes a mechanism of MAPK pathway activation in low-grade astrocytomas. J Clin Invest 118: 1739–1749.
13. Forshew T, Tateossian RG, Lawson AR, Ma J, Neale G, et al. (2009) Activation of the ERK/MAPK pathway: a signature genetic defect in posterior fossa pilocytic astrocytomas. J Pathol 218: 172–181.
14. Price N, Sullivan R, Harmelin V, Ehrlich M (2009) Mapping of BRAF V600E mutations in tumor samples using the Sequenom OncoCarta™ Panel. Nature Methods 6.
15. Sievert AJ, Jackson EM, Gai X, Hakonarson H, Judkins AR, et al. (2009) Duplication of 7q34 in pediatric low-grade astrocytomas detected by high-density single-nucleotide polymorphism-based genotype arrays results in a novel BRAF fusion gene. Brain Pathol 19: 449–450.
16. Cheng Y, Pang JC, Ng HK, Ding M, Zhang SF, et al. (2008) Pilocytic astrocytoma do not show most of the genetic changes commonly seen in diffuse astrocytomas. Histopathology 57: 437–444.
17. The Cancer Genome Atlas Research Network (2008) Comprehensive genomic characterization defines human glioblastoma genes and core pathways. Nature 455: 1061–1068.