



OPEN ACCESS

EDITED AND REVIEWED BY
Michael Baudis,
University of Zurich, Switzerland

*CORRESPONDENCE
Mohamed E. Abazeed,
mabazeed@northwestern.edu

SPECIALTY SECTION
This article was submitted to Cancer
Genetics and Oncogenomics,
a section of the journal
Frontiers in Genetics

RECEIVED 26 August 2022
ACCEPTED 20 September 2022
PUBLISHED 03 October 2022

CITATION
Yang H-T, Crawford DC and
Abazeed ME (2022), Editorial:
Translating clinical genomics and health
informatics into precision oncology.
Front. Genet. 13:1029212.
doi: 10.3389/fgene.2022.1029212

COPYRIGHT
© 2022 Yang, Crawford and Abazeed.
This is an open-access article
distributed under the terms of the
[Creative Commons Attribution License
\(CC BY\)](https://creativecommons.org/licenses/by/4.0/). The use, distribution or
reproduction in other forums is
permitted, provided the original
author(s) and the copyright owner(s) are
credited and that the original
publication in this journal is cited, in
accordance with accepted academic
practice. No use, distribution or
reproduction is permitted which does
not comply with these terms.

Editorial: Translating clinical genomics and health informatics into precision oncology

Hsih-Te Yang¹, Dana C. Crawford² and Mohamed E. Abazeed^{3,4*}

¹Atrium Healthcare, Charlotte, NC, United States, ²Case Western Reserve University, Cleveland, OH, United States, ³Feinberg School of Medicine, Northwestern University, Chicago, IL, United States, ⁴Robert H. Lurie Comprehensive Cancer Center, Northwestern University, Chicago, IL, United States

KEYWORDS

precision oncology, personalized medicine, clinomics, panomics, clinical prediction models

Editorial on the Research Topic

[Translating clinical genomics and health informatics into precision oncology](#)

Some cancers are driven by targetable genomic alterations that dysregulate key pathways influencing cell growth and survival. These tumors lend themselves to significant, although inevitably transient, clinical responses upon targeted drug treatment. Accordingly, omic-based theranostics (*i.e.* diagnostics that guide therapeutic interventions) have transformed the management of a substantial number of cancers (King et al., 1985; Druker et al., 2001; Slamon et al., 2001; Demetri et al., 2002; Lynch et al., 2004). However, after a highly productive discovery phase resulting in the identification of multiple clinically actionable genomic targets, new information capabilities have been limited by challenges in target druggability, treatment toxicity, and intratumoral heterogeneity. Therefore, the ability to harness tumor omic information to its full clinical potential has not yet been realized.

The impact of these challenges has led to a pivot toward a more comprehensive understanding of the complexity of tumor information and the role of distinct biological processes in driving tumor phenotypes and, hence, patient outcomes. To achieve the requisite level of (epi)genome-phenome knowledge, the interrogation of large-scale datasets for integral biological pathways that regulate tumor behavior has come into greater focus. Technological developments in high-content biological and data platforms have led to the generation of large-scale multi-omics datasets in radiomics, genomics, epigenomics, transcriptomics, proteomics, metabolomics, and phenomics (Song et al., 2020). Collections of publicly available datasets like The Cancer Genome Atlas (TCGA) and The Cancer Imaging Atlas (TCIA) provide critical resources to a research environment that is frequently hindered by access to clinically relevant big data (TCGA Research Network; Cancer Imaging Program). These and similar datasets can have broad utility in advancing a more comprehensive and integrated understanding of individual cancers.

The contributions within this Research Topic are in line with ongoing efforts to better characterize the associations between omic determinants and tumor phenotypes. 1) Li et al. explored a putative role for circular RNA (circRNA) in predicting the immune landscape of lung adenocarcinoma. Their data suggests that a circRNA-related risk score model is associated with the level of immune cell infiltration in the tumor that could impact the efficacy of adjuvant treatments (chemotherapy, immunotherapy, or targeted agents). 2) Chen et al. studied the impact of germline copy number variants (gCNV) on the prognosis of patients with non-small cell lung cancer (NSCLC). They developed a prognostic nomogram model and validated their findings in an external cohort. Whether the two gCNVs they have identified, CNVR395.1 and CNVR2239.1, regulate or are accidentally associated with (*i.e.*, represent a confounder) NSCLC phenotypes require experimental validation. 3) Ye et al. examined the prognostic utility of a signature of ferroptosis, a form of programmed oxidation-related cell death associated with iron accumulation, a decrease in antioxidant cellular capacity, and the accumulation of lipid peroxidation or reactive oxygen species (ROS). In addition to clinical and functional studies validating aspects of this signature, a nomogram that incorporated clinical features provided reasonable concordance with overall survival. 4) Lastly, Wang et al. studied the association of long non-coding RNAs (lncRNA) with prognostication in NSCLC. They reveal lncRNA that are associated several clinical and pathological variates including histology, tumor size, stage, and, relatedly, overall survival. Altogether, these studies represent varied attempts to link omic variates, biological pathways, and patient outcomes. Many more attempts that integrate the panoply of omic variates from bulk and single cells coupled with well annotated clinical outcomes are needed.

The convergence of technology and large-scale dataset development has created an unparalleled opportunity to couple omic information with clinical outcomes for many patients with cancer. Rigorous evaluations will require many additional validations and innovations that are guided by additional advances in tumor omic profiling, data interpretation and

integration, and innovative clinical trial designs. The rigor with which these studies are validated and prospectively tested will determine whether and how comprehensive tumor information is incorporated into the routine care of patients. Despite the stated potential, there remain several pitfalls for the informed incorporation of omic data for clinical testing and use. A main hurdle is the risk that large-scale association data emerges without evidence-based clinical approaches for data interpretation and integration. It will be pivotal to proactively construct scientific frameworks with adequate guardrails that can prioritize tumor omic information for clinical use.

Author contributions

MA wrote and H-TY and DC edited the manuscript.

Conflict of interest

MA received speaking, advising, and/or providing educational programs for the following companies or entities: the American Society of Clinical Pathology (ASCP) and Mirati Therapeutics, Inc. MA is a named inventor in a patent for the use of Deep Profiler and iGray to personalize radiotherapy dose.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

References

- Cancer Imaging Program (n. d). The cancer imaging archive. Available at <https://www.cancerimagingarchive.net/>.
- Demetri, G. D., von Mehren, M., Blanke, C. D., Van den Abbeele, A. D., Eisenberg, B., Roberts, P. J., et al. (2002). Efficacy and safety of imatinib mesylate in advanced gastrointestinal stromal tumors. *N. Engl. J. Med.* 347, 472–480. doi:10.1056/NEJMoa020461
- Druker, B. J., Talpaz, M., Resta, D. J., Peng, B., Buchdunger, E., Ford, J. M., et al. (2001). Efficacy and safety of a specific inhibitor of the BCR-ABL tyrosine kinase in chronic myeloid leukemia. *N. Engl. J. Med.* 344, 1031–1037. doi:10.1056/NEJM200104053441101
- King, C. R., Kraus, M. H., and Aaronson, S. A. (1985). Amplification of a novel v-erbB-related gene in a human mammary carcinoma. *Science* 229, 974–976. doi:10.1126/science.2992089
- Lynch, T. J., Bell, D. W., Sordella, R., Gurubhagavatula, S., Okimoto, R. A., Brannigan, B. W., et al. (2004). Activating mutations in the epidermal growth factor receptor underlying responsiveness of non-small-cell lung cancer to gefitinib. *N. Engl. J. Med.* 350, 2129–2139. doi:10.1056/NEJMoa040938
- Slamon, D. J., Leyland-Jones, B., Shak, S., Fuchs, H., Paton, V., BajAmonde, A., et al. (2001). Use of chemotherapy plus a monoclonal antibody against HER2 for metastatic breast cancer that overexpresses HER2. *N. Engl. J. Med.* 344, 783–792. doi:10.1056/NEJM200103153441101
- Song, M., Greenbaum, J., Luttrell, J., Zhou, W., Wu, C., Shen, H., et al. (2020). A review of integrative imputation for multi-omics datasets. *Front. Genet.* 11, 570255. doi:10.3389/fgene.2020.570255
- TCGA Research Network (n. d). The cancer genome Atlas program. Available at <https://tcga-data.nci.nih.gov/tcga/tcgaHome2.jsp>.