Cancer Informatics in 2017: A New Beginning and a Bright Future

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Summary
Objective: To summarize significant research contributions on cancer informatics published in 2017.
Methods: An extensive search using PubMed/Medline, Google Scholar, and manual review was conducted to identify the scientific contributions published in 2017 that address topics in cancer informatics. The selection process comprised three steps: (i) 15 candidate best papers were first selected by the two section editors, (ii) external reviewers from internationally renowned research teams reviewed each candidate best paper, and (iii) the final selection of three best papers was conducted by the editorial board of the Yearbook.
Results: The three selected best papers present studies addressing many facets of cancer informatics, with immediate applicability in the research and clinical domains.
Conclusion: Cancer informatics is a broad and vigorous subfield of biomedical informatics. Strides in knowledge management, crowdsourcing, and visualization are especially notable in 2017.

Keywords
Neoplasms, informatics, health information technology, genomics, data visualization

Introduction
The information being generated in the cancer clinic and the basic science lab is beginning to elucidate fundamental understandings of cancer beyond the traditional anatomically-driven approach. The nascent field of cancer informatics intends to take full advantage of the many data streams with several fundamental goals: 1) organizing the data in ways that are comprehensible and meaningful to clinicians, researchers, and patients; 2) using the data to advance the treatment of cancer; and 3) manipulating the data, most commonly through visualization, to yield new insights. In this inaugural year of the Cancer Informatics section, we introduce the readers to a very broad and deep field which will continue to rapidly develop. As pointed out about cancer informatics research by Mathé, et al., [1] in the survey paper of the Cancer Informatics section of this 2018 IMIA Yearbook, “the cancer informatics revolution has been the beneficiary of a data explosion”. Big data being generated through genomics, metabolomics, and proteomics is the clearest avenue towards precision oncology.

In 2018, the selection of papers in cancer informatics intends to illuminate the current progress of research with a focus on efforts to translate research towards immediate clinical applicability.

Paper Selection Method
Two electronic databases were searched, PubMed/MEDLINE and Google Scholar. Searches were performed in January 2018 to identify peer-reviewed journal articles published in 2017, in the English language, related to cancer informatics research. In addition to the search through electronic databases, manual searches of key themes were performed in well-known informatics journals (e.g., Journal of the American Medical Informatics Association, Applied Clinical Informatics, Bioinformatics, Journal of Biomedical Informatics, etc.). Additionally, the contents of the journals JCO Clinical Cancer Informatics, JCO Precision Oncology, and Cancer Informatics were searched, as well as the contents of the 2017 special issue of Cancer Research: Focus on Computer Resources. For relevant articles that were PubMed indexed, we also searched for additional relevant articles using PubMed’s “Similar articles” service. Finally, we also hand-searched the proceedings of MedInfo 2017, the 2017 AMIA Annual Symposium, and the 2017 AMIA Joint Summits.

One of the two section editors performed the searches. A PubMed search for the MeSH terms “Neoplasm” and “Informatics” yielded a total of 3,158 references. The Google Scholar search for “cancer informatics” returned 29,800 results. Given these vast results, we focused on identifying articles with translational or clinical applications, as opposed to more fundamental bioinformatics methodologies. Then, the two section editors undertook independently the initial screening of titles and abstracts to identify papers relevant to the field of interest. Both section editors classified the papers into three categories: definitely include, possibly include, or exclude. They then reviewed in detail the possibly include full-text articles to finally reach a mutual list of 15 candidate best papers. Papers were considered according to their originality, innovativeness, scientific and/or practical impact, and scientific quality.
In accordance with the IMIA Yearbook selection process [2], the 15 candidate best papers were evaluated by the two section editors and by additional external reviewers (at least four reviewers per paper). Three papers were finally selected as best papers (Table 1). A content summary of the selected best papers can be found in the appendix of this synopsis.

**Conclusions and Outlook**

The three selected best papers are representative of three distinct subdomains of cancer informatics: knowledge management, visualization, and crowdsourcing.

Chakravarty, et al., [3] described a large public web resource, OncoKB (http://oncokb.org/) with a goal of providing evidence-based information about primarily somatic genomic variants for clinicians and researchers. The content is curated and stored in an internal data model which is exposed via a public application programming interface (API) on demand. As of this writing, OncoKB contains information on 477 genes, 3,855 variants, 60 tumor types, and 86 drugs. Twenty-five genes are linked to FDA-approved or standard of care treatment evidence, and 39 are linked to more limited clinical or biological (non-human) evidence. The practice of oncology is increasingly informed by biologic factors beyond the traditional biomarkers, in particular those obtained through clinical genomic sequencing; OncoKB is a cornerstone in the emergent ecosystem of cancer genomics knowledge management.

Newton, et al., [4] presented TumorMap (https://tumormap.ucsc.edu/), an interactive portal for the exploration of molecular similarities across cancer samples. TumorMap uses Google’s Map technology to arrange samples in a hexagonal grid based on their similarity, after applying user-selected dimensionality reduction techniques. The resulting maps can be colored by various attributes such as clinical, molecular, phenotype, and outcome data and metadata. This novel and practical methodology allows not only the assessment of similarity and dissimilarity but also supports experimental research purposes. Outside of the clinical setting, high-dimensional -omics data are being routinely generated on large numbers of cancer samples and exploring these data in an intuitive and meaningful way is an unmet need that the authors have addressed.

Seyedinasrollah, et al., [5] described the process of the Prostate Cancer DREAM Challenge, the first crowd-sourced competition in metastatic prostate cancer and possibly in any cancer, which was launched in 2015; primary results are described elsewhere [6]. DREAM challenges are crowd-sourced competitions meant to accelerate progress in various biomedical informatics problems; importantly they are built on the FAIR data principles: Findable, Accessible, Interoperable and Reusable. This challenge focused on discontinuation of chemotherapy treatment for the reason of toxicity, which is as common as discontinuation for lack of efficacy. Predicting which patients are likely to experience early treatment failure is a critical issue in the oncology domain. The challenge attracted 34 independent teams from around the world and has led to post-challenge community collaborations. While the best results were only slightly better than reference, integrated time-dependent area under the curve (iAUC) of 0.791 versus 0.743, the successful operation of the challenge, which included data from four independent phase III randomized controlled trials, suggests a paradigm for crowd-sourced tasks in the oncology domain.

The other candidate best papers are in the same line with innovative and/or effective cancer informatics approaches.

Huang, et al., [7] and Kurmit, et al., [8] described knowledge management approaches that bear some similarities with the OncoKB effort. Similar to the previously described Clinical Interpretation of Variants in Cancer (CIViC) [9] effort and MyCancerGenome [10], these knowledge bases help to build a rich ecosystem.

The American Association for Cancer Research (AACR) Project GENIE Consortium [11] described a substantial multi-institutional effort to aggregate next generation sequencing results obtained in routine clinical care with clinical annotations. The consortium recently expanded from eight to 19 institutions and the freely available data will become an increasingly valuable commodity for future research efforts.

Three of the candidate best papers [12–14] developed information extraction systems specific to the oncology domain. Bui, et al., [12] focused on the extraction of oral chemotherapy exposure and Gao, et al., [13] focused on pathology reports. In the tasks of primary site identification and histologic grade classification, hierarchical attention networks, a form of deep learning, achieve much better results than naïve Bayes, logistic regression, support vector machine, random forest, and other traditional machine learning models. Savova, et al., [14] described DeepPhe, a
Software application for mining and presenting cancer center’s social media page and analysis of comment topics on a large dataset representing signs of problems with medication adherence, care, analysis of patient forum messages for patient-reported outcomes into routine cancer care brought the patient and caregiver voices into the process to select IMIA Yearbook best papers. Methods Inf Med 2015;54(2):135–44.


Gandy, et al., 2017 described a software application for mining and presenting relevant cancer clinical trials per cancer mutation. Clinical trial matching in the era of precision oncology presents significant informatics challenges which are likely to be the focus of future editions of this chapter.

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References


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Appendix: Content Summaries of Selected Best Papers for the IMIA Yearbook 2018, Section Cancer Informatics


OncoKB: a precision oncology knowledge base
JCO Precis Oncol 2017 Jul;2017

The practice of oncology is increasingly informed by biologic factors beyond the traditional biomarkers, in particular those obtained through clinical genomic sequencing. Variants have prognostic and predictive implications and comprise a large and growing knowledge space. Chakravarty et al., have built a large public web resource, OncoKB (http://oncokb.org/) with a goal of providing evidence-based information for clinicians and researchers. The content is curated and stored in an internal data model which is exposed via a public API on demand. As of this writing, OncoKB contains information on 477 genes, 3,855 variants, 60 tumor types, and 86 drugs. Twenty-five genes are linked to FDA-approved or standard of care treatment evidence, and 39 are linked to more limited clinical or biological (non-human) evidence. OncoKB is a cornerstone in the emergent ecosystem of cancer genomics content management, and is developed and maintained by the Knowledge Systems group in the Marie Josée and Henry R. Kravis Center for Molecular Oncology at the Memorial Sloan Kettering Cancer Center (MSK), in partnership with Quest Diagnostics.


TumorMap: exploring the molecular similarities of cancer samples in an interactive portal
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Outside of the clinical setting, high-dimensional omics data are being routinely generated on large numbers of cancer samples. This includes genomic, transcriptomic, proteomic, and epigenomic profiles that bring data to terabyte and petabyte scales. Exploring these data in an intuitive and meaningful way is an unmet need; despite the rapidly falling costs of technology, analysis remains a major bottleneck. In this Focus on Computer Resources article, the authors describe TumorMap (https://tumor-map.ucsc.edu/), an interactive portal for the exploration of molecular similarities across cancer samples. TumorMap uses Google’s Map technology to arrange samples in a hexagonal grid based on their similarity, after applying user-selected dimensionality reduction techniques. The resulting maps can be colored by various attributes such as clinical, molecular, phenotype, and outcome data and metadata. This novel and practical methodology allows not only the assessment of similarity and dissimilarity but also supports experimental research purposes.


A DREAM challenge to build prediction models for short-term discontinuation of docetaxel in metastatic castration-resistant prostate cancer

Discontinuation of chemotherapy treatment for the reason of toxicity is as common as discontinuation for lack of efficacy. This may have major consequences when treatments have a very narrow therapeutic index. Predicting which patients are likely to experience early treatment failure is a critical issue in the oncology domain. One approach to this intractable problem is to engage the larger community to solve towards a common problem. The DREAM challenges are crowd-sourced competitions meant to accelerate progress in the resolution of various biomedical informatics problems; importantly they are built on the FAIR data principles: Findable, Accessible, Interoperable and Reusable. The article by Seyednasrollah et al., describes the process of the Prostate Cancer DREAM Challenge, the first crowd-sourced competition in metastatic prostate cancer and possibly in any cancer, which was launched in 2015; primary results are described elsewhere. The challenge attracted 34 independent teams from around the world, and has led to post-challenge community collaborations. While the best results are only slightly better than reference, integrated time-dependent area under the curve (iAUC) of 0.791 versus 0.743, the successful operation of the challenge, which included data from four independent phase III randomized controlled trials, suggests a paradigm for crowd-sourced tasks in the oncology domain.