Developing and Evaluation of Computational Phenotypes of Metastatic Breast Cancer Using All of Us Data

Chenyu Li, MS¹, Israel O. Dilan¹, Shyam Visweswaran, MD PhD¹, Michael J. Becich, MD PhD¹, Xia Jiang, PhD¹, Richard D. Boyce, PhD¹ ¹Department of Biomedical Informatics, University of Pittsburgh School of Medicine

Introduction

Breast cancer is the most frequent cancer among women worldwide,¹the metastatic breast cancer is the main cause of mortality for breast cancer patients. Computational phenotyping, an informatics approach, that extracts phenotypes from real-world data such as electronic health records (EHRs), has the potential to advance medicine's understanding of who is most at risk for metastatic breast cancer. All of US (AoU) is a research program funded by the National Institutes of Health that aims to recruit one million highly diverse patients across the United States.² The program makes data from participants accessible to both participants and approved researchers through the AoU "workbench." In this study, we attempt to implement computational phenotypes from multiple sources to describe the prevalence of metastatic breast cancer in the AoU patient cohort. We used the AoU research workbench to implement the metastatic breast cancer phenotypes and describe cases within this rapidly growing a real clinical research database.

Methods

We applied two different computational phenotypes in December 2021: *Phenotype 1* was published on the PheKB knowledgebase of computational phenotypes by eMERGE group³. *Phenotype 2* was from a classification and regression tree (CART) metastatic breast cancer phenotyping algorithm developed from a journal paper⁴. For *Phenotype 1*, we searched SNOMED-CT codes mapped from ICD-10 and ICD-9 codes listed on PheKB breast cancer phenotype document using AoU workbench cohort construction tool followed the phenotyping workflow. For *Phenotype 2*, we followed the CART algorithm logic reimplemented the phenotype using AoU workbench Python environment. Descriptive analyses were performed to analyze the demographics differences between two phenotypes. **Result**

Of 201,920 participants in the AoU EHR database, Phenotype 1 identified 6,957 breast cancer patients while Phenotype 2 identified 3,679 patients. A small subset of these cohorts had metastatic breast cancer, 20 (0.28%) for Phenotype 1 and 190 (5.4%) for Phenotype 2. Only 8 participants were identified by

Breast cancer	Phenotype 1	Phenotype 2
Non metastatic	6,937	3,489
Metastatic	20	190
Total	6,957	3,679

(5.4%) for Phenotype 2. Only 8 participants were identified by both phenotypes. Phenotype 1, which used AoU workbench breast cancer AoU participants identified by each

cohort construction tool SNOMED search, identified residents from only 2 states while Phenotype 2, which queried ICD and CPT codes on workbench Python environment, identified residents from 7 states.

Discussion

This study is the first to apply metastatic breast cancer phenotypes on AoU data. There were considerable differences in the number of metastatic patients identified by the two computational phenotype algorithms. While the AoU workbench enabled the implementation of the two phenotypes, we found some limitations. It was unexpected to see metastatic breast cancer participants only came from two states in Phenotype 1. AoU data contribution centers harmonized the EHRs data to the OMOP CDM, the data mapping process may vary from center to center. Therefore, researchers should be aware that concept coding will influence the participants capture. We suggest that comprehensive definition and detailed phenotyping algorithms of computational phenotypes should be reported in AoU data research. In the future, we plan to assess how the AoU breast cancer population differs from other established databases and how different phenotypes will influence metastatic breast cancer prediction models.

Conclusions

Our application of metastatic cancer phenotypes to the AoU data using the workbench helped describe the participants' subgroup characteristics and is a valuable case study of the AoU workbench as a clinical research informatics platform.

Acknowledgements

This study was partially supported by the U.S. Department of Defense Award No. W81XWH1910495(to XJ) and NIH OT2 OD026554. Other than supplying funds, the funding agencies played no role in the research.

References

1. Ferlay J et al. Global Cancer Observatory: Cancer Today. Lyon: International Agency for Research on Cancer; 2020

2. Denny JC, Rutter JL, Goldstein DB, et al. The "All of Us" Research Program. N Engl J Med 2019;381(7):668-

3. Shang Nea. Breast Cancer PheKB. Secondary Breast Cancer PheKB 2018. https://phekb.org/phenotype/1052.

4. Nordstrom BL, Whyte JL, Stolar M, Mercaldi C, Kallich JD. Identification of metastatic cancer in claims data. Pharmacoepidemiology and Drug Safety 2012;**21**:21-28 doi: 10.1002/pds.3247[published Online First: Epub Date]