

Federated Analysis for Cancer Variant Interpretation Melissa Cline, Ph.D.



Genetics strongly influences the risk of breast cancer



www.cancer.gov/brca-fact-sheet



Common Heritable Cancer Syndromes

Syndrome	Gene	Incidence	Cancers
Hereditary breast and ovarian cancer syndrome	BRCA1 BRCA2	1/300-800 Ashkenazi: 1/40	Breast, ovary, melanoma, prostate, pancreatic
Hereditary ovarian cancer syndrome	RAD51C RAD51D BRIP1	Unknown	Ovary
Lynch syndrome	MLH1 MSH2 MSH6 PMS2 EPCAM	1/660-2000	Uterine, colon, ovary, pancreatic, gastric, small intestine, central nervous system, renal, sebaceous
Cowden syndrome	PTEN	1/200,000	Breast, uterine, thyroid, colon, renal, sebaceous
Li-Fraumeni syndrome (LFS)	P53	Unknown	Sarcomas, breast, adrenal, brain, lung, endometrial
Peutz-Jeghers	STK11	1/25,000- 300,00	Gastrointestinal, breast, ovarian, sex cord stromal, uterine, cervical (adenoma malignum)

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How many cancers are heritable?

Cancer Genetic Risk Assessment





Heritable cancer risk can be manageable

- Screening to detect any cancer in its early stages, when it is most treatable.
- Medications that lower cancer risk.
- **Risk-Reducing Surgery** to remove high-risk tissue before cancer can develop.

To take advantage of these strategies, you need to get tested, and the test needs to recognize your genetic cancer risk.

Genetic testing for hereditary cancer mutations can save lives.



ACMG/AMP Germline variant classification

- Pathogenic variant
- Likely pathogenic variant
- Variant of uncertain significance (VUS)
- Likely benign variant
- Benign variant

ACMG/AMP Germline variant classification



ACMG/AMP Germline variant classification



Benign variant

Prevalence of Variants of Uncertain Signifcance (VUS)



The Variant Interpretation Bottleneck



Data as of April 2021

Variant interpretation involves assessing many forms of evidence together

Types of Variant Evidence

- Clinical Family Studies
- Clinical Patient Observations
- Functional Assays
- Population Frequencies
- Computational Prediction





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Most variant interpretation requires some clinical evidence

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Most cancer-associated variants are highly rare

- No single institution is likely to have enough patient data for robust variant interpretation
- Patient-level data is impossible to share for privacy reasons
- How can one share knowledge about genetic variants while safeguarding patient privacy?





BRCA Challenge

- Global consortium launched by the GA4GH.
- Vision: assemble team of experts to pioneer data sharing for BRCA and cancer, as an exemplar for other genes and disorders.



Stephen Chanock

Sir John Burn

Rachel Liao





SANTA CRUZ Genomics

Solution: BRCA Exchange



Benedict Paten, UCSC



Gunnar Rätsch, ETH Zurich



APIs, data format standards, Security, ELSI mechanisms



BRCA Exchange _____ Variant Details BBCA1 NM 007294 3rc 4985T-C Transcript Visualization ынны 1 2 3 5 6 7 8 9 10 11 12 13 -14 - 15 16 17 18 19 20 21 22 23 24 15 16 Substitution (1 base) Deletion (0 bases) Insertion (0 bases) Option 🗆 🔲 Clinically Important Functional Domains (ENIGMA Consortium) Donor Sites Huntsman Cancer Institute Functional Domains Clinical Significance (ENIGMA) Clinical Significance (ClinVar) Clinical Significance Benig PAGMA Comment on Clinical Significance • * * * • Submitter Women's Health and Genetics/Laboratory Corporation of Am., Binigh (15 Ame 2021 for class as per Plon et al. 2008 (PMID: 18951446). Class 1 based on posterior probabil **** + Submitter Invites Likely Benigh (23 May 2021) Assertion Method ENGMA BRCA1/2 Classification Orbiris (2015) Likely Benign (27 February 2021) ++++ + Submitter GeneDx Date last evaluated 10 August 2015 -----Benigh (3) February 2021) Collection Method Curation + + + + Submitter Ambry Genetics Clinical Significance PMD: 21990134 **** • Submitter BIC (BRCA1) Allele Origin Germine Condition(s) (Mode of Breast-Ovarian Cancer, Familial, Susceptibility To, 1 (OMM)) Clinical Significance (BIC) Allele Frequency Reference Sets In Silico Prior Prediction (prior to considering other evidence) a in Silco Print Probability of Pathonenicity 0.04 gnomAD V2.1 Excrees, Non-Gancer (Graphical) 0.00 0.25 0.50 0.75 1.00 gnomAD Exomes Applicable Prior Probability is from Splicing-level Estimation (Donor Splice Site Impact) VEP Consequences Cl splice donor variant Gredita: Computational algorithm and claplay derived from the HCI Breast Cancer Genes Prior Probabilities website.

+ Protein-level Estimation 0.03

Splicing-level Estimation 0.04

CRAVAT - MuPIT 3D Protein View

chr17 n 43070929 AvG in the 3D structure of BRCA1 BBCT Domain

all to open MuPIT viewer from the CRAWT project, and view the variant in a 3D protein structs.

The promAD data sets used by BPCA Exchange are the "non-cancer" subsets of three sources. Data from TOGA and other cancer onhots are excluded to ensure that the frequencies used to assess pathogenolly represent individuals no advended by cancer. Additional data for this variant, including detailed populations, quality acores, and flags relative to other transcripts, are evaluate at growtAD. gnomAD Exomes (scaled) ATE AME ASI EAS THE NET SAL OTH Popmax Filtering AF (95% confidence): 0.00001968 (EA5) If the filter alide frequency of a variant is above the maximum credible population AF for a condition of interest, then this variant should be filtered (e not considered a candidate causative variant).

Banigs (25 March 2020)

Benign (EF December 2020

VUS D0 March 2020

anomAD V2.1 Express, Non-Cancer (Numerical) gnomAD V3.1 Genomes, Non-Cancer (Graphical) variant, including detailed populations, qualit gnomAD Genome gnomAD Genomes (scaled) AFE AME AM FIN MID MFE SAS OTH APE AND AND ASU EAS FIN MID INTE SAS OTH Promov Elleving AE 85% confidence/: 0.0001506 (EAS) If the filter able frequency of a variant is above the maximum credible population AF for a condition of interest, then this variant should be filtered (e not considered a candidate causative variant).

gromAD V3.1 Genomes, Non-Cencer (Numerical)

Show Empty Items Verient Nomenclature Gene Symbol BRCAT Reference cDNA Sequence NM_007294.3 HGVS Nucleotide c-4985T>C HGVS Protein p.Phe16625er Abbreviated AA Change F10028 BIC Designation 6104T>C ClinGen Allele Registry CA003117 Genome (GRCh38) NO_000017.1119.430709294-0 Genome (GRCh37) NC_000017.10:g-41222946A-G ENA 6 CMTD 5/70 Beacons https://beacon-network.org/#/search? chrome178pose412229466ref+Adailele=O8ra=ORCh37 GA4GH VRS Identifier 9H9NVA.628TJGMvPOFMCyLQL17_CTIPM6KJqf55 2017 0171-02, FP1605, FP1 Clinical Significance (LOVD) **Dinical Classification: benign** + Submitter(a) ENIGMA consortium (Brisbane, AU) Submitter(s) Thomas Hansen (Copenhagen, DK) Cirical Classification: **Cirical Classification:** Submitter(s) Peter Devilee (Leiden, NL) a Submitteria) BBIDGES connectium (Europe NL) Clinical Classification: Submitter(s) Yukihide Mornozawa (Yokohama, JP Submitteria) Johan den Dunnen (Rotterdam, NL) stion: benign, W. Multifactorial Likelihood Analysis Posterior probability of 0.00021 pathcoenicity

HOME VARIANTS COMMUNITY HELP MORE -

Prior probability of 0.04 pathogenicity Missense analysis 0.03 pathogenicity prior Co-occurrence likelihood 0.00355 Segregation Likelihood Ratio Summary Family History 1.44 Literature Reference Easton et al. 2007 Functional Assay Results Assays were selected by the ENGMA consortium as high quality assays that met internal standards permitting estimation Genetic variation can impact RNA transcripts and/or protein function or abundance. Assevs labeled with a P services was when the impact of genetic variation at the protein level. Assays labeled with a P measure the impact of genetic variation at the protein level. Assays labeled with a R/P measure impact at both the RNA and protein level. The data present even. Reports, or the active's interpretation of the functional impact of each variant, are presented as they appear in the publications. A report of Many Provided means that the authors performed more than one assay for each variant but provided no overall report of functional impact. Additional assays may be added in the future.

Functional assay data reflect laboratory models of disease, and should not be used as a substitute for clinical variant interpretation. Further, please note that Inrespective of the authors' terms to reflect impact on function, the results presented **do not reflect a final** Disclaimer variant classification.

Findlay et al, 2018

RNA Score 0.1218836751 RNA Class Not Depleted Report Descriptions FUNC: Functional, INT: Intermediate, LOP: Loss of function

P Fernandes at al. 2019

inctional classification results from saturation genome editing, provided by the Starts lab or (Finday et al, Nature, 2016). Please see their site for further information. 600 400 \$V 200 A -55 -50 -45 -40 -35 -30 -25 -20 -15 -10 -05 00 05 10 Function Score

Loss of Function of 0.0 and the median nonsense 3NV scored -0.12.

Uncertain No Functional Impact

Report: FUNC

Report 1.0

Author Findlay et al, 2018 Publication PMD-30209390 Report FUNC Functional Envictment 0.2548575376 Score

Variant identification

Clinical Assertions

Evidence of Variant Pathogenicity

Sharing clinical data for variant interpretation

- Variant interpretation relies on sensitive patient-level data
- But, the information actually needed are variant-level summaries, *which are much less sensitive!*
 - Case-control ratios
 - Allele Frequencies
 - Variant co-occurrences
- Vision: instead of sharing the patient-level data directly, share a container to compute the variant-level summary data

Privacy-preserving data sharing through federated analysis

Traditional Data Sharing



Federated Analysis



Variant Co-occurrence Estimation

- **Rationale:** BRCA1/2 are "essential" genes. The cell needs at least one working copy.
- Two pathogenic BRCA variants *in trans* is usually embryonic lethal, but can lead to a rare disease with a top life expectancy of 40
- If a VUS is observed in an unaffected individual of age 40 or greater and the VUS is either:
 - *in trans* with a known pathogenic variant (green),
 - or in a homozygous genotype (*grey*)

Then this supports a benign interpretation



Federated analysis of BioBank Japan Data

- RIKEN holds a large cohort of cancer patients and controls from BioBank Japan, which they cannot share directly.
- We shared a Docker container with them to analyze their patient cohort for variant co-occurrences and allele frequencies
- The container generated variant-level data, which we are now using together with the ENIGMA Consortium to interpret BRCA variants!









Next Federated Analysis: Fanconi Anemia (FA)

- Rare childhood-onset disease that arises from two pathogenic BRCA (and other) variants *in trans.*
- Diverse phenotype that includes early onset cancer
- Ongoing research on which germline variants are implicated in FA.
- Vision: Mine the Kids First data on Cavatica for new examples of FA.









Challenge #1: not all clouds are created equal

- Had previously been running the co-occurrence Docker container on Terra on NHLBI BioData Catalyst
- Cavatica is based on Seven Bridges
- Porting the container from Terra (WDL) to Seven Bridges (CWL) required work
- Bigger issue: making containers portable across cloud platforms



Focus on the Science with a few Elements of Style

January 14, 2022 Anne Deslattes Mays, PhD National Scholar @adeslat anne.deslattesmays@nih.gov



Researcher challenges to data and compute







Focus on the science...



- Using platforms as a service lets the researcher focus on the science
 - Platform assists in managing compute costs (automatic logging out)
 - Platforms facilitate collaboration
 - Platforms abstract out the technical machine details, allowing work at a higher level
- Good science provides repeatable steps for others to follow
 - Well documented methods allows the next scientist to repeat what you have done
 - Document those steps using markdown
 - Give attribution to those who wrote those processes or workflows you have expanded upon or modified
- Using these practices of containerization abstracts away the platform-specific layer making your work more portable and accessible for use on other platforms



Elements of Style in Reproducible Workflow Creation

- All workflows share the same basic steps differences are in the technical details
 - Input
 - Process
 - Output
- Containerize at the process level
 - Dockerfile to capture the details
 - \circ $\,$ Conda to control the environment $\,$
 - GitHub to build the Dockerfile and Create the Image
 - Automate and keep up-to-date with GitHub actions
 - Document the objective
- Stitch together with the standard workflow languages dictated by the platform
- Use GitHub to store and share the workflow and markdown to document your steps





Elements of Style in Reproducible Workflow Creation and Analysis: An INCLUDE & Kids First Training Event



Coming soon!







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Eunice Kennedy Shriver National Institute of Child Health and Human Development



Challenge #2: Container certification

- Goal of the BRCA Challenge: gather data from diverse global repositories
 On-going work in integration with the GA4GH Starter Kit
- Challenge: how can you demonstrate to external collaborators that your container is trustworthy?

Ongoing work related to container certification

- NHLBI BioData Catalyst and Dockstore
 - Emerging best practices including: no root dependencies, no data egress
- GA4GH Cloud Workstream Test Bed
 - Quantifies basic system demands such a CPU usage
- BioCompute Objects
 - Community-driven initiative to build a framework for characterizing and sharing computational workflows
 - Has FDA support



Summary

- Container methods and federated analysis show promise for sharing information from patient data for variant interpretation
 - Publication to appear in Cell Genomics
- Ongoing challenges include:
 - Cloud portability
 - Workflow certification
- There are promising developments in both areas

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NIH Big Data to Knowledge (BD2K)



NIH NATIONAL CANCER INSTITUTE

Informatics Technology for Cancer Research (ITCR)