

Closing Wrap Up



Mathematical Frameworks for Integrative Analysis of Emerging Biological Data Types

June 15 - 19, 2020

Zoom from Banff International Research Station, Canada



[Aedin Culhane](#) (Dana-Farber Cancer Institute, Harvard TH Chan School of Public Health)
[Elana Fertig](#) (John Hopkins University)
[Kim-Anh Lê Cao](#) (University of Melbourne)



Banff International Research Station
for Mathematical Innovation and Discovery

 [#BIRSBioIntegration](#)

Goals of this workshop

Multi-omics integration of single cell data

- is an active and emerging field
- May provide insight that cannot be obtained from single datasets
- **lacks established performance benchmarks,**
- gold standard datasets, assessment standards.

Bring together **interdisciplinary computational scientists**

- to examine cutting edge techniques for integrative analysis of diverse multi-omics.
- Provide & assess **open source resources** for multi-platform analysis
- Formulate **goals and future directions** to advance multi-omics analysis

Products: Guidelines, build collaboration, code & datasets, a **white paper**

Transparency
Collaboration
Open science
Fairness
Inclusion

#BIRSBioIntegration Community



3 challenging data challenges



16 contributed talks focusing on analysis

5 keynotes

9 Brainstorming sessions



Data and GitHub code shared



339 Commits to manubot



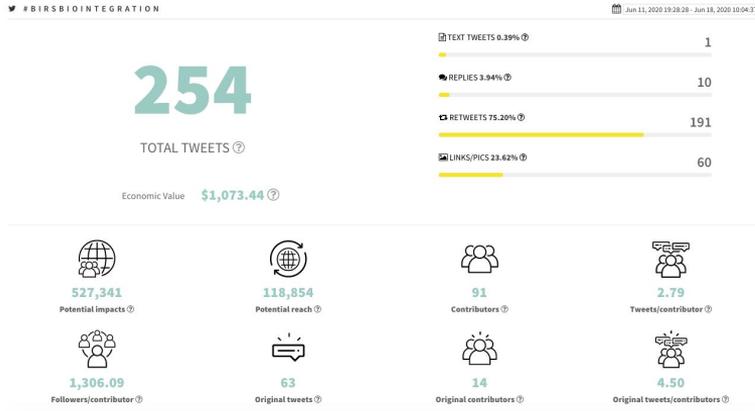
156 Members, 16 Active Channels on Slack



Outreach Beyond Banff



#BIRSBioIntegration



Live Stream <http://www.birs.ca/live>

These visitors

Visitors	743
Unique visitors	464
Actions	3,751
Average actions	5.0
Total time	4d 18h
Average time per visit	9m 13s
Bounce rate	8.2%

Top traffic sources ▼

Direct	393
Social media	149
Searches	102
Links	86
Email	13

<https://twitter.com/hashtag/BIRSBiointegration>

Emerging Research: Five keynote speakers



Prof. GC Yuan
Dana-Farber Cancer Institute,
Harvard TH Chan School of
Public Health

Mon



**Prof. Bernd
Bodenmiller**
University of Zurich

Tues



Prof. Oliver Stegle
German Cancer Research
Center & EMBL

Wed



**Prof. Susan
Holmes**
Stanford University

Thurs



Prof. Vincent Carey
Harvard Medical School,
Brigham & Women's Hospital

Fri

Contributed talks from hackathon participants

sc seq-FISH

Alexis Coullomb

Hang Xu

Dario Righelli

Amrit Singh

Joshua Sodicoff

sc Targ Proteomics

Yingxin Lin

Chen Meng

Pratheepa Jeganathan

Kris Sankaran

Lauren Hsu

Duncan Forster

scNMT-seq

Al J Abadi

Joshua Welch

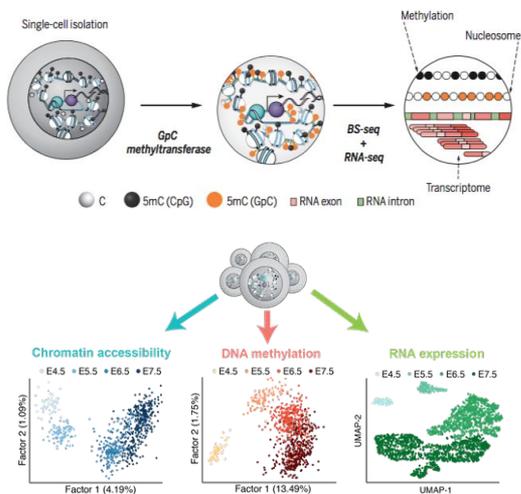
Arshi Arora

Wouter Meuleman

3 Hackathon Challenges

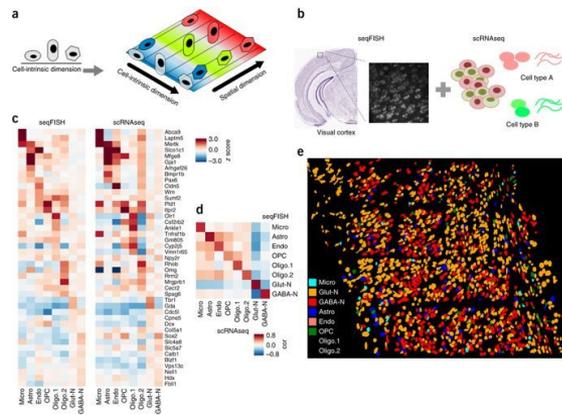
Gastrulation (scNMT)

826 cells matching across all data sets (transcriptome, DNA accessibility and DNA methylation) after quality control and filtering.



Adult mouse visual cortex seqFISH, scRNAseq

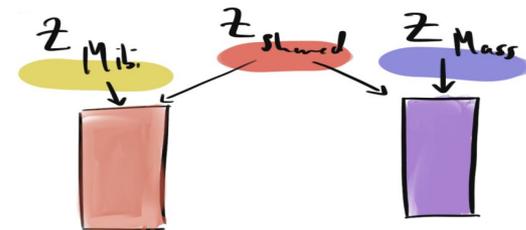
- seqFISH - 1,597 single cells x 125 genes mapped (Zhu *et al* 2018)
- scRNA-seq. ~1,600 cells (Tasic *et al* 2016)



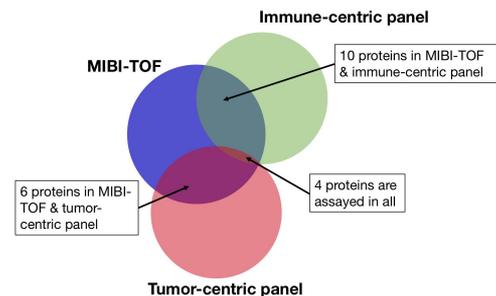
Breast Cancer sc Proteomics

Non-overlapping patients

MIBI 40 TN, Mass Tag 7 TN



... with 20 overlapping proteins

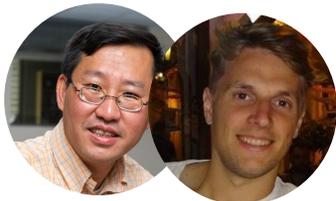


Hackathon Challenge Brainstorms

Spatial Fish	Targeted Proteomics	RNA - DNA	Summary
<p>Expt design, Platform Specific bias, Inclusion of spatial information</p>	<p>Normalisation, Partial feature overlap Non-overlapping cells Integrating by phenotype Inherent spatial nature of biological data,</p>	<p>Binary data Transfer learning or imputation using other atlases, Non-linear integration</p>	<p>Summary of common challenges: Non-overlapping features and/or cells, from data-driven towards mechanistic driven,</p>
<p>Objective Assessment,</p>	<p>Scale/metrics from single cell to cell communities</p>	<p>DNA features summary,</p>	<p>Generic towards context specific methods</p>
	<p>Annotation Atlases and maps for benchmarking</p>	<p>Annotation of histone db</p>	<p>Incorporate prior knowledge</p>



9 Brainstorming sessions



seqfish_theme

**Guo-Cheng Yuan &
Ruben Dries**

Dana-Farber Cancer Institute,
Harvard TH Chan School of Public
Health & Boston University



sc_targ_proteomics_theme

**Aedin Culhane &
Olga Vitek**

Dana-Farber Cancer Institute,
Harvard TH Chan School of Public
Health & Northeastern University



scNMT-seq_theme

**Ricard Arguelaget &
Oliver Stegle**

German Cancer Research Center &
EMBL



summary_analyses_theme

**Kim-Anh Lê Cao &
Casey Green**

University of Melbourne & Uni
Pennsylvania



benchmark_theme

**Mike Love &
Matt Ritchie**

University of North Carolina-Chapel Hill
& Walter and Eliza Hall Institute



Susan Holmes
Stanford University

interpretation_theme



Vincent Carey
Harvard Medical School and
Brigham & Women's Hospital

software_theme



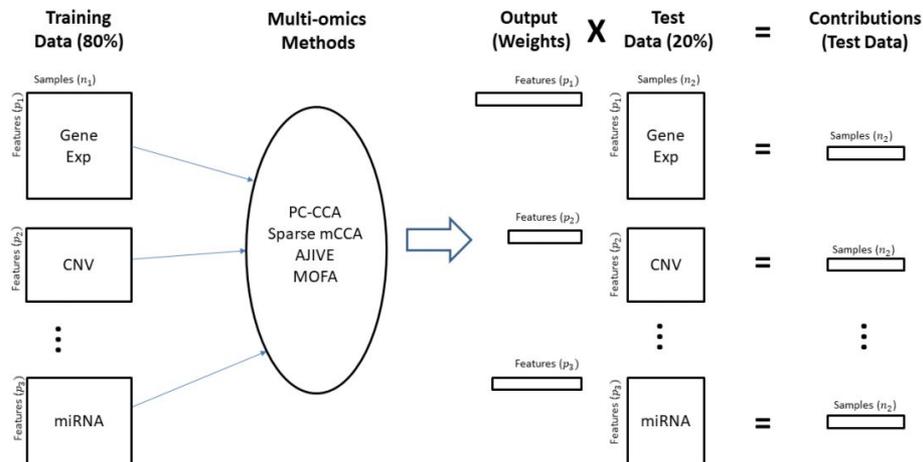
Elana Fertig
Johns Hopkins University

future_theme

Benchmarking	Interpretation	Software	Future
<p>Establish performance benchmarks and assessment standards</p>	<p>Issue of benchmarking datasets immunology gated discrete</p>	<p>Representation mutli-view data Spatial Modality Colocation eQTL</p>	<p>High cell/large tissue (HCA, Allen, HTAN)</p>
<p>Assessment metrics Datasets benchmarks</p> <p>Deliver open source resources for multi-platform analysis (data wrangling)</p> <p><u>Awesome-multi-omics</u></p>	<p>Vocabulary for inside data science versus towards biologists</p> <p>Glossary for paper (appendix)</p> <p>Figures and visualization for communication versus discovery.</p>	<p>Annotation 4D, blueprint -Cell State-Cell State. Dropouts</p> <p>Scalability - containers</p> <p>Connecting to consortiums</p> <p>Color blind standard (import for UMAP)</p>	<p>Need pertubations/ dynamic datasets</p> <p>Data sharing</p> <p>Molecular coverage Deeper sampling</p> <p>Which data for which question</p> <p>Training on model</p>

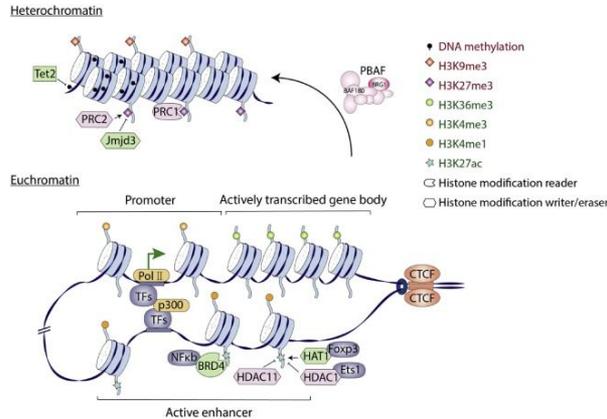
Community Coordination & Communication

- Representations
- Scale
- Metrics
- **Unified language**
- Annotation, ontology resources
- Leverages skills in other disciplines (spatial)
- Training - across disciplines
-
- Benchmarking dataset - ground truth
- **What would be most interesting?**



DNA “accessible” for gene expression?

- DNA -> Regulation -> RNA -> Protein -> Regulation
- heterochromatin v euchromatin (silent v active) DNA defines the genome accessible for transcription
- Genome organization variability in cell types, states, (differentiation, development, stress, disease) unknown
- If regions are expected background off and other expected “accessible” (within a expt negative control?)



Using the Genome in experimental design

Which chromatin features under selection (active) and which features are evolutionary silent (historical)?

How precisely can chromatin define normal cell types

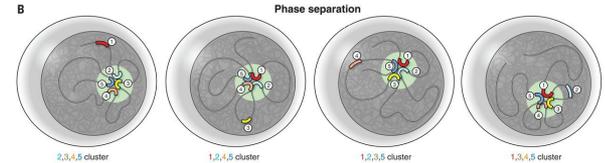
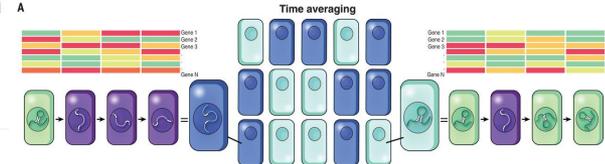
REVIEW

Molecular basis and biological function of variability in spatial genome organization

Elizabeth H. Finn*, Tom Misteli*

+ See all authors and affiliations

Science 06 Sep 2019;
Vol. 365, Issue 6457, eaaw9498
DOI: 10.1126/science.aaw9498



“Stable functional states and cell populations can be generated by two mechanisms: time- or population averaging of gene activity (**Fig. 4A**) or the formation of functionally equivalent but morphologically diverse cellular structures (**Fig. 4B**).”

The accessible genome “open” for gene expression

Bulk RNAseq normalization approaches assumed 50% genes silent in sample

>50% RNAseq in single cells are silent?

Impact on DE gene expression analysis of scRNAseq if the

Heterchromatin \ni G $p(E) = 0$
Euchromatin \ni G $p(E) > 0$

(imputation, dropout..)

Predicting # functional mRNA molecules

Delineate heterochromatin and transcriptional silencing.

Histone marks, Methylation of promoter/enhancers

Transcription bursts (3 state model)

Nascent mRNA, half life (cap/tail)

miRNA

How do we distinguish cause vs effect of interactions?

*Activity dependent on functional network of gene

Protein complexes

Activation enzyme (precursor -> active form cleavage)

Post -translational modification

Co-localization

[Requires Multi-omics](#) * activity can be measured with proteins or inferred by expression of downstream targets

bulk - single cell

BULK

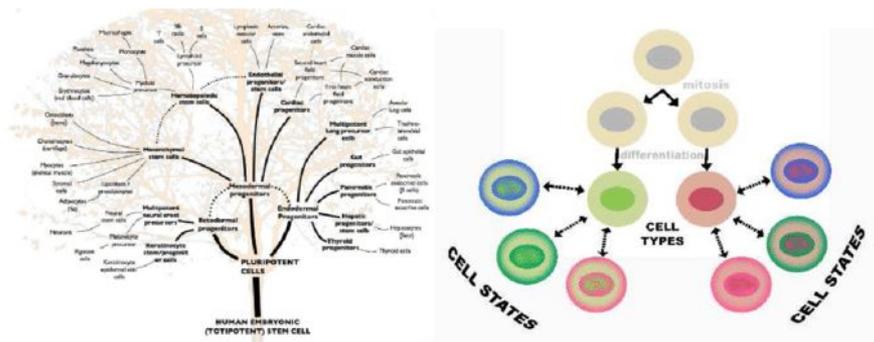
Qualitative
assessments of cell
identity



sc

Quantitative,
high-resolution cell
atlases

Cell lineage -> Cell Type \neq Cell State



Cell State - dependent on local autocrine, paracrine, community signalling. More dynamic/variant.

Cell Type - relatively stable except for chromatin reorganization (stress/CNV/ dev)

=> Would predict bulk RNAseq captures



Cell

Volume 173, Issue 2, 5 April 2018, Pages 291-304.e6



Article

Cell-of-Origin Patterns Dominate the Molecular Classification of 10,000 Tumors from 33 Types of Cancer

Katherine A. Hoadley^{1, 21, R, B}, Christina Yau^{2, 3, 7, 21}, Toshinori Hinoue^{4, 21}, Denise M. Wolf^{5, 21}, Alexander J. Lazar^{6, 21}, Esther Drilf^{7, 21}, Ronglai Shen^{7, 21}, Alison M. Taylor^{8, 9, 21}, Andrew D. Cherniack^{8, 9, 21}, Vesteinn Thorssson^{10, 21}, Rehan Akbani^{8, 21}, Reanne Bowlby^{11, 21}, Christopher K. Wong^{12, 21}, Maciej Wiznerowicz^{13, 14, 15}, Francisco Sanchez-Vega¹⁶, A. Gordon Robertson¹¹, Barbara G. Schneider¹⁷, Michael S. Lawrence^{8, 18} ... Peter W. Laird^{4, 22} R, B

In Statistics



“premature summarization is the root of
all evil in statistics and
data science”



Single Cells -> Communities -> Phenotype

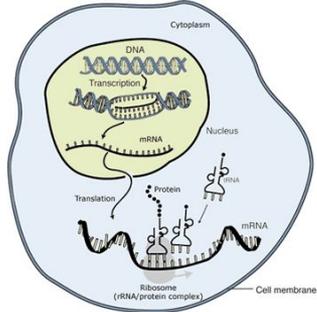
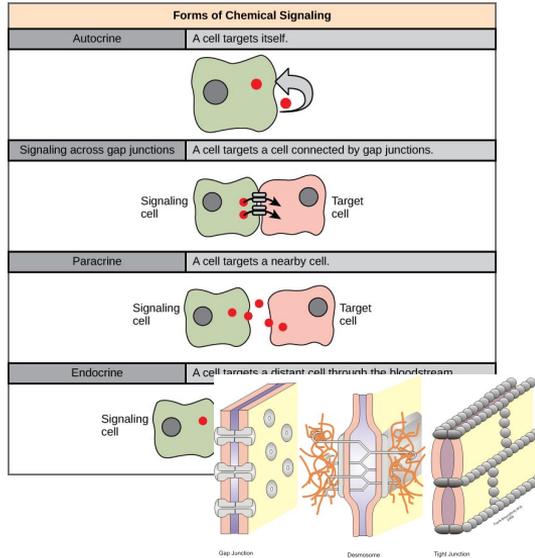
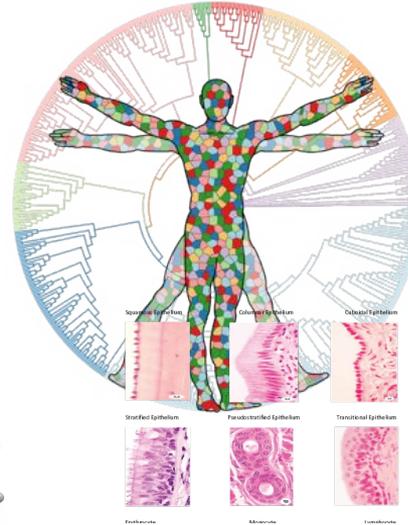


Image adapted from: National Human Genome Research Institute.

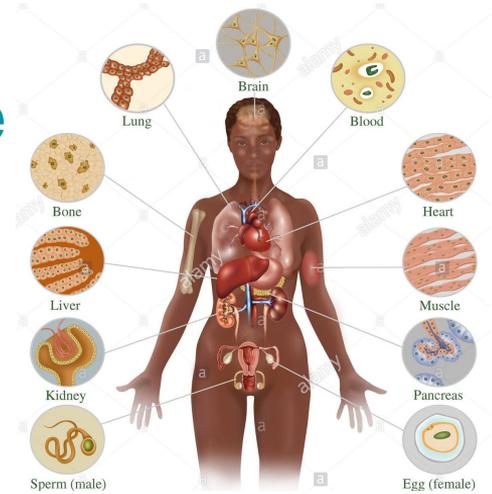
'Omics DNA Chromatin
RNA Protein
Glycosylation-
metabolites etc



Connected by signaling
(paracrine, endocrine
Gap junctions, autocrine)



Composed of organized
Cells types, polarity



Human Phenotype
defined by
Systems,
Organs that are
composed of Cell
Communities

Emerging Needs

Infrastructure

- Representation of each data multi-view , unified language, Cell /tissue type specific Ontologies,
- Representation/Visualization of anatomy

Benchmarking

- Methods for integration of different scales /merging later / mapping at pheno level
- Datasets to enable identification of DNA chromatin structure-> histone marks ->

Education

- As disciplines work together, Nomenclature dictionaries /common terms
- Education/Conference across discipline, especially in spatial biology - biologists learn from other fields and not reinvent GIS/weather/ecology

Products from meeting for multi-platform analysis

Datasets

Online- Bioc package

**Open source
resources**

Code

Code for all contributed talks

Glossary/Language - Google Sheet (Data/Methods/Education) -
Resource available as Awesome-sc list

White Paper

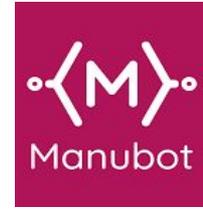


Optimistic Timeline for White Paper

- Week 1 (June 26):
 - theme leaders push **outline** to Manubot to manage theme overlaps
 - Glossary of terms signed off
- Week 2 (July 3): **full section** written (~ 1 page + 1 Figure)
- Week 4 (July 17): **first draft** distributed to all for comments
- Week 6 (July 31): **comments back** from *all* co-authors
- Week 8 (August 14): finalise and submission

<https://birsbiointegration.github.io/whitePaper/>

Goal: White Paper



Manubot for white paper

339 commits 5 branches 0 packages 0 releases 1 environment 20 contributors

Branch: master New pull request Create new file Upload files Find file Clone

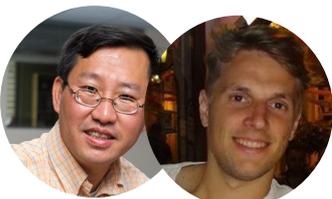
BIRSBiointegration Merge pull request #4 from ejfertig/patch-2 Latest commit a90bb72

github/workflows	GitHub Actions: cache manubot files in ci/cache
build	upgrade manubot to fix webpage subprocess handling
ci	Export environment variables needed for gh-pages readme
content	Update metadata.yaml
output	GitHub Actions workflow for building and deployment
webpage	GitHub Actions workflow for building and deployment
.appveyor.yml	.appveyor.yml: note about skipping branches with PR
.gitignore	Dependency upgrade on 2019-06-03 with multiple ref file su...
LICENSE-CC0.md	Dual license code and data under CC0
LICENSE.md	Switch CC BY license to markdown
README.md	slight re-work of the readme
SETUP.md	Simplify setup by creating branches later
USAGE.md	metadata: use list for author.funders
screenshot_pull_reque...	Add files via upload

#manubot channel

Pull requests managed by Casey Greene, organisers and theme leaders

White Paper



1. Spatial Transcriptomics: [#seqFish_theme](#)



2. RNA - DNA: [#scNMT-seq_theme](#)



3. Targeted Proteomics: [#scTarg_Proteomics_theme](#)



4. Summary methods: [#summary_Analyses_theme](#)

- [01.abstract.md](#)
- [02.introduction.md](#)
- [10.current-tech.md](#)
- [20.interp-challenges.md](#)
- [30.case-studies.md](#)
- [32.scNMT.md](#)
- [35.scRNA.md](#)
- [37.spatial.md](#)
- [40.common-methods....](#)
- [50.software.md](#)
- [60.benchmarking.md](#)
- [70.discussion.md](#)

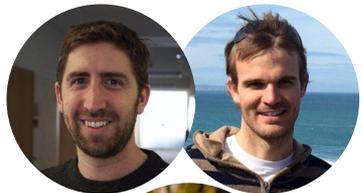
White Paper



1. Interpretation challenges: [#interpretation_theme](#)



2. Software infrastructure: [#software_theme](#)



3. Benchmarking: [#benchmark_theme](#)



4. Future Directions: [#future_theme](#)

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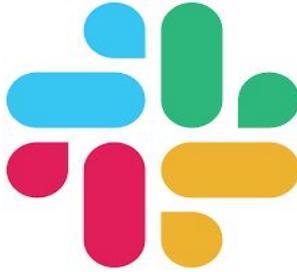
Communication will be key in the coming weeks!

Live



Zoom

Communication



Slack

Datasets, code, paper



Github



BIRSBioIntegration

<https://github.com/BIRSBioIntegration>

Monitor these tools and make good use of them!

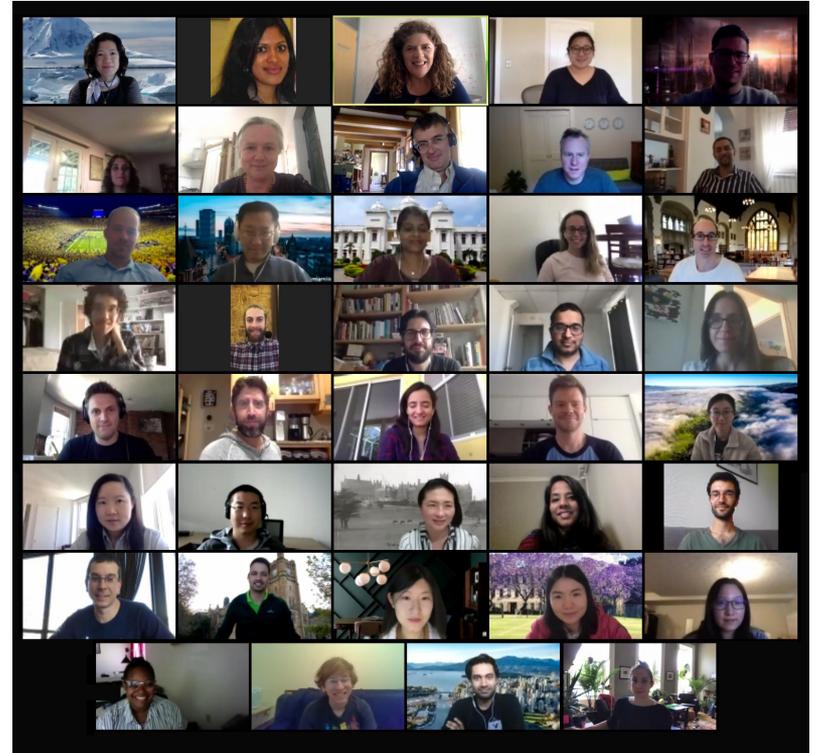
Thank you for staying up late & waking up early

Interest in;

- Follow up meeting in Banff (deadline for application is Sep/Oct)
- Designing our own benchmarking expt and asking \$\$ from CZI?
- Other ideas. Please suggest.

A **first poll** will be distributed to state your authorship contribution.

Go to **#information** channel lists all important links



On behalf of the (fully zoomed) organizers - Thank You



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Banff International Research Station
for Mathematical Innovation and Discovery

Scientific Program Coordinator: Chee Chow
Program Assistant: Dominique Vaz
Station Manager: Linda Jarigina-Sahoo
Technology Manager: Brent Kearney

 @BIRS_Math