



Deep Data Needs and Challenges in Precision Health

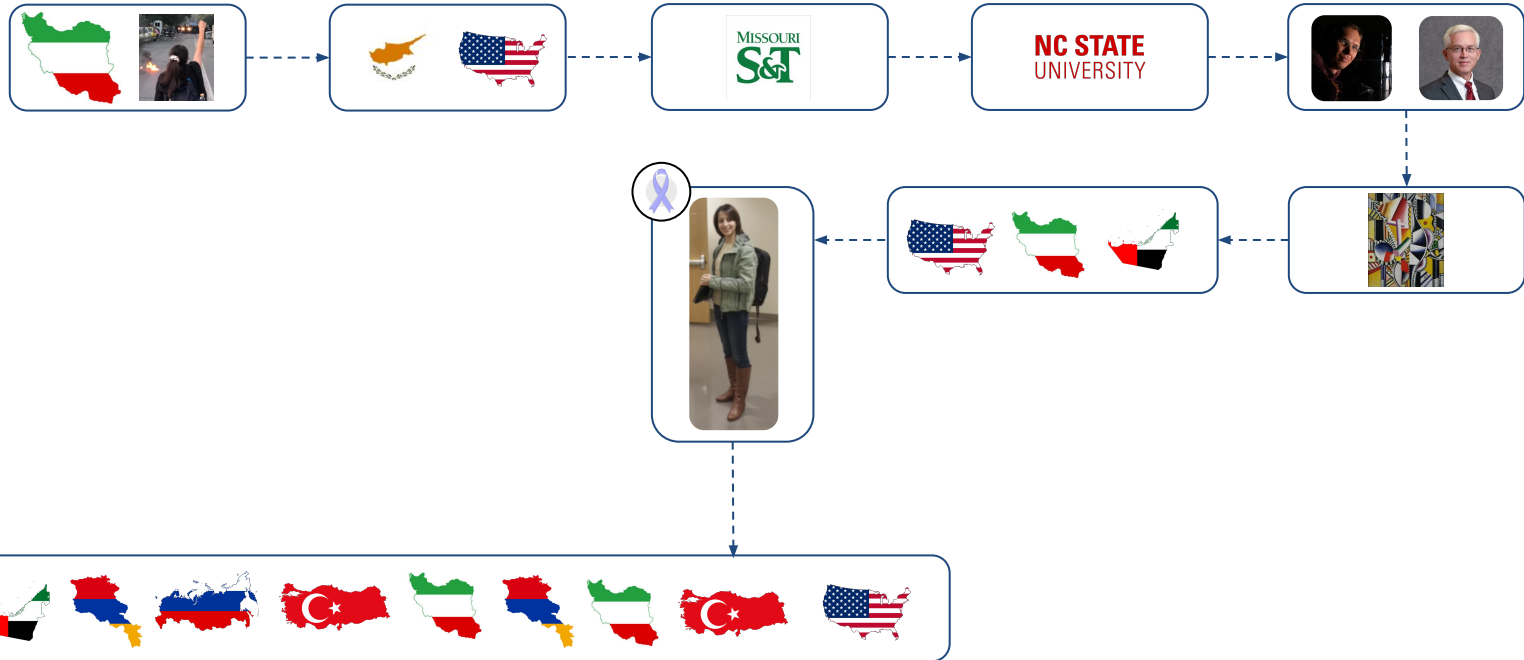
Amir Bahmani, PhD

Outline

- **Personal Story 1:** International Students
- **Precision Medicine:** Needs and Challenges
- **Disruptive Solutions:**
 - Deep Data Collection and Analysis at Scale
 - **Real Time Health Framework**
 - Personal Health Dashboard (PHD)
 - Real-time Alert System
- **Personal Story 2:** Collaboration
 - Optimizing Computation and Storage
 - **Hummingbird:** Performance Prediction
 - **Swarm:** Federated Learning
- **Personal Story 3:** Serendipity
 - Unlocking Deep Data for Precision Medicine Applications
 - **Serverless System:** Stanford Data Ocean

Timeline

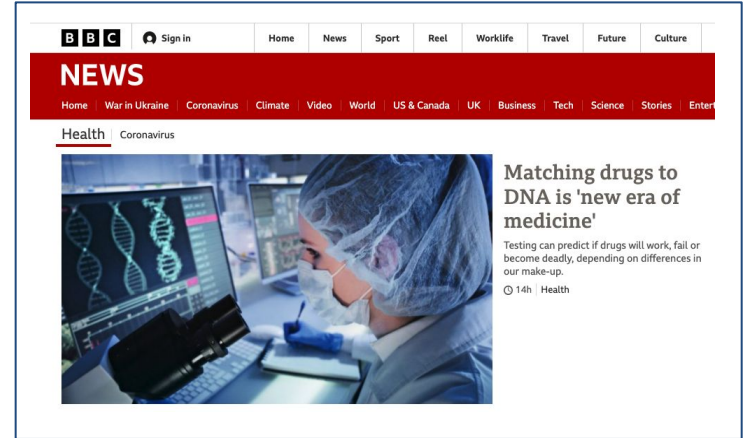
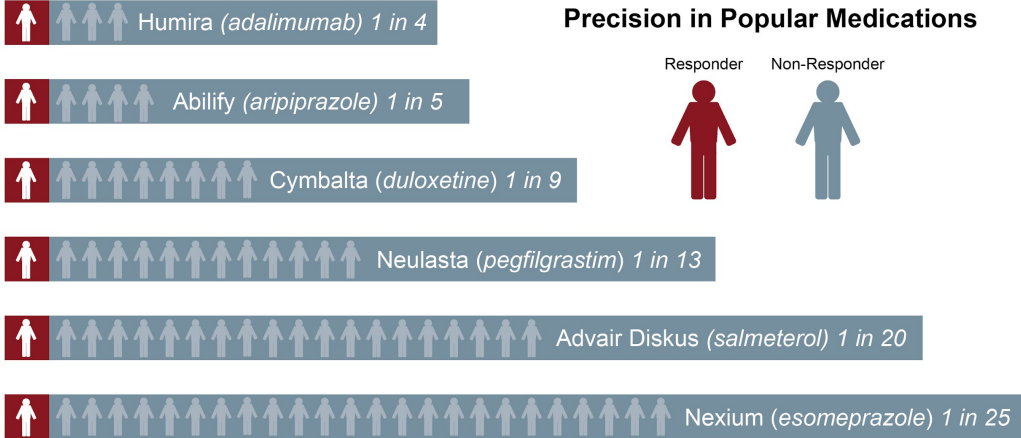
Life of an International Student



From **Shallow** Medicine to **Deep** Medicine

“The top ten highest-grossing drugs in the United States help between 1 in 25 and 1 in 4 of the people who take them.”

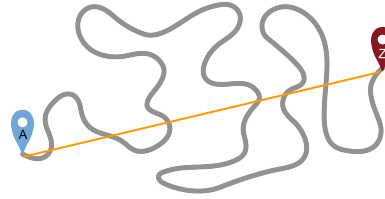
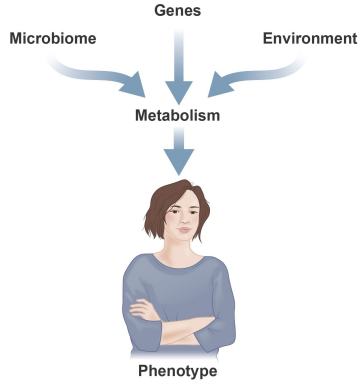
Schork, Nicholas J. "Personalized medicine: time for one-person trials." *Nature* 520.7549 (2015): 609-611.



“Some drugs are completely ineffective or become deadly because of subtle differences in how our bodies function.” BBC News, 29 March 2022.

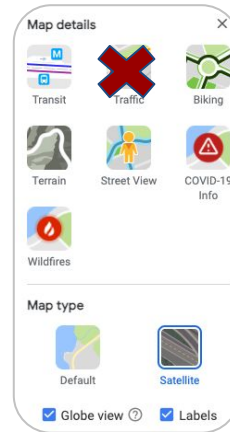
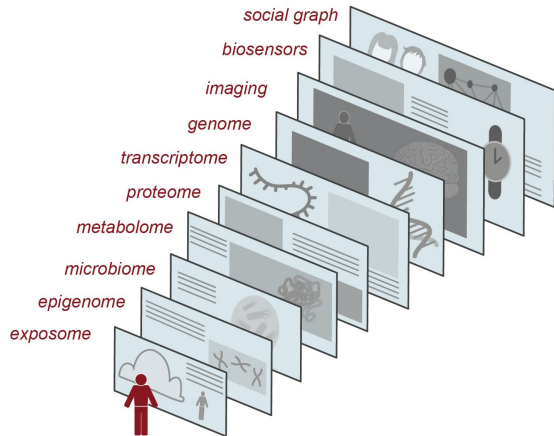
Source: <https://www.bbc.com/news/health-60903839>

From **Shallow** Medicine to **Deep** Medicine (cont'd)



“It is more important to know what sort of person has a disease than to know what sort of disease a person has”

~ Hippocrates, a Greek physician, 2500 years ago



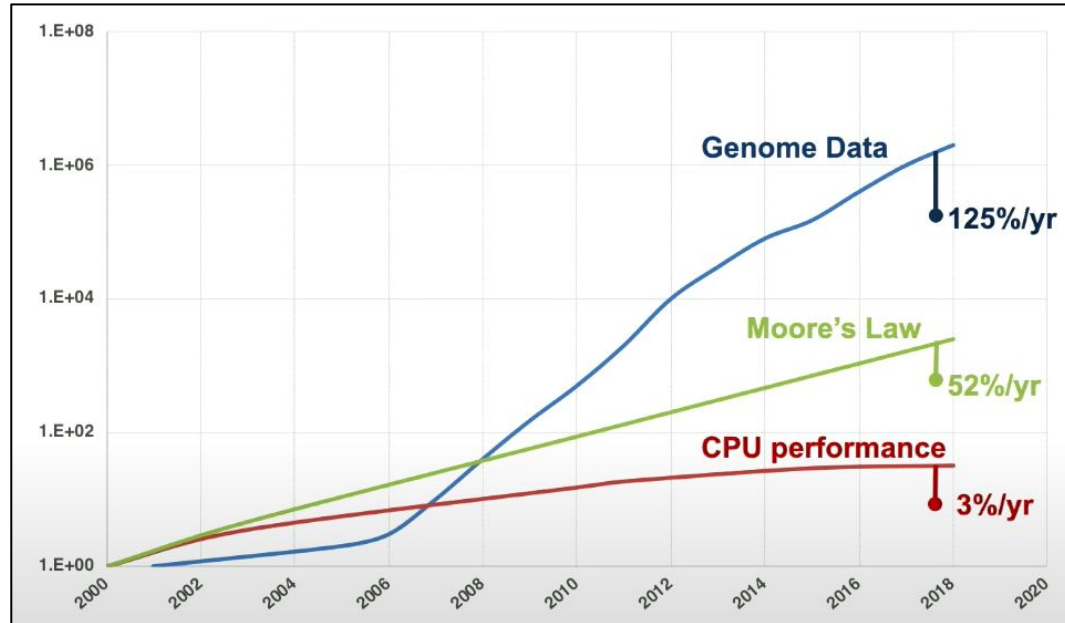
A Tsunami of Data is Approaching the Healthcare Industry



At Stanford Center for Genomics and Personalized Medicine (SCGPM), we have collected over 2 PB of Data around Dr. Michael Snyder

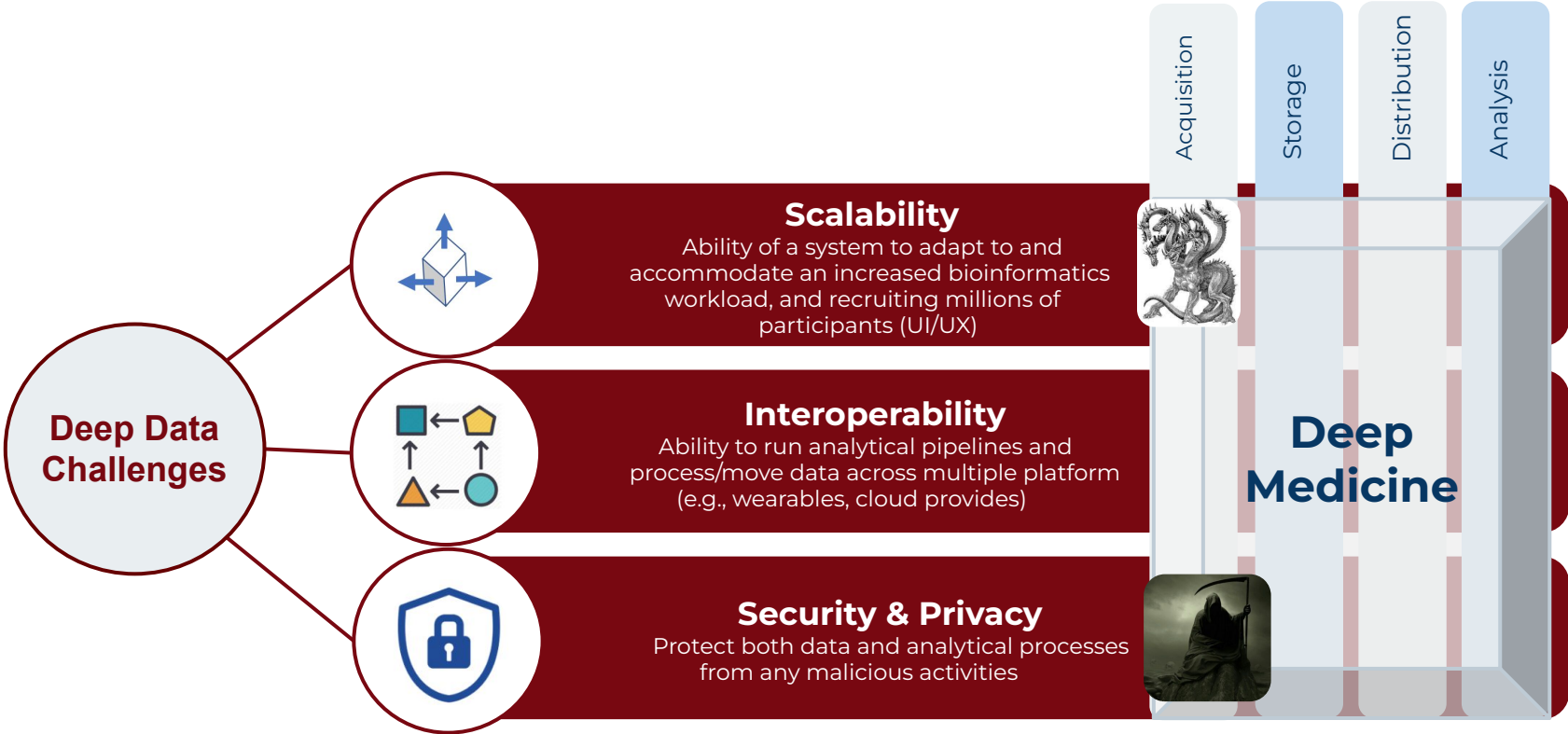
Cost of Standard Storage Per Month for **1PB**
(pay-as-you-go) as of March 2022

GCP	\$20000 US West1
Azure	\$19100 US West
AWS	\$21000 US West (Oregon)



Source: <https://www.usenix.org/conference/atc19/presentation/turakhia>

Deep Data Needs and Challenges

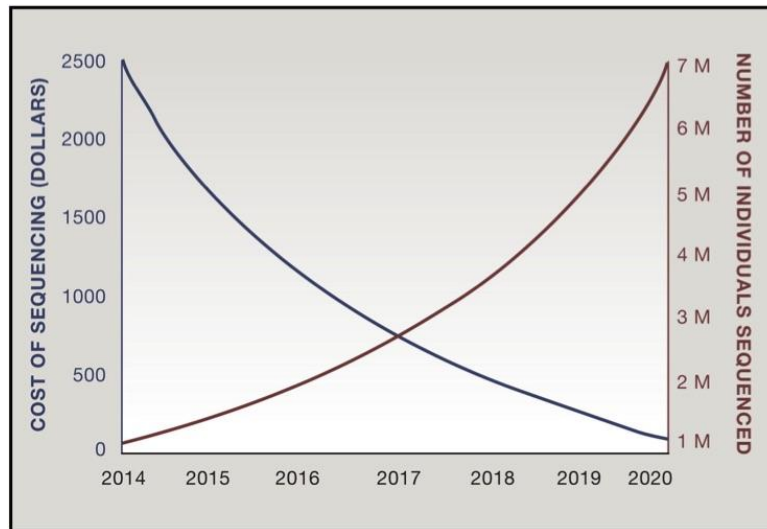


The Gap

How can I test my scalable system with a real-world application?



Computer Scientists/Engineers



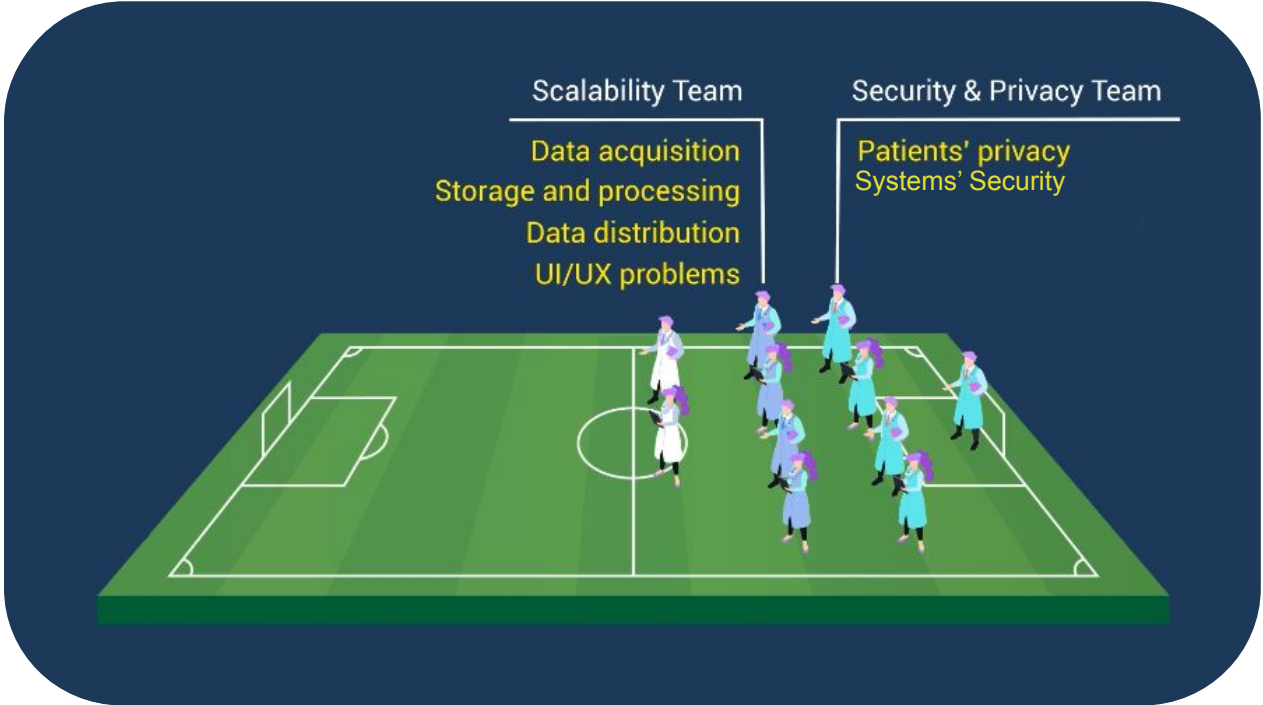
Topol, Eric J. "Individualized medicine from prewomb to tomb." *Cell* 157.1 (2014): 241-253.

How am I going to scale my gene model?



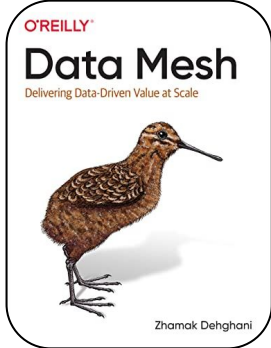
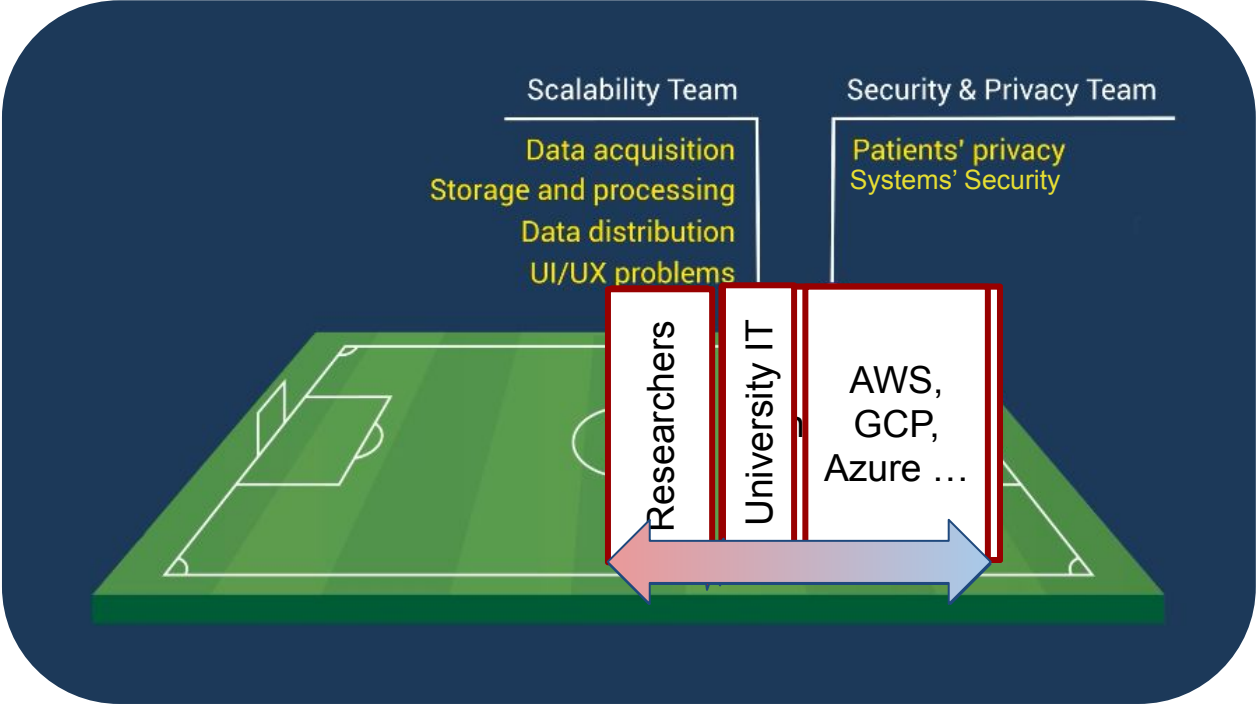
Biologists/Physicians

Problem #1: Team Formation in Schools of Medicine



Problem #2: Global Governance

Maintain Dynamic Equilibrium b/w Domain Autonomy and Global Governance



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Wearables can track many things



Fitbit Charge



Fitbit Ionic



Apple Watch



Garmin



SensOmetrics



Dexcom G6



Biostrap



Motiv Ring



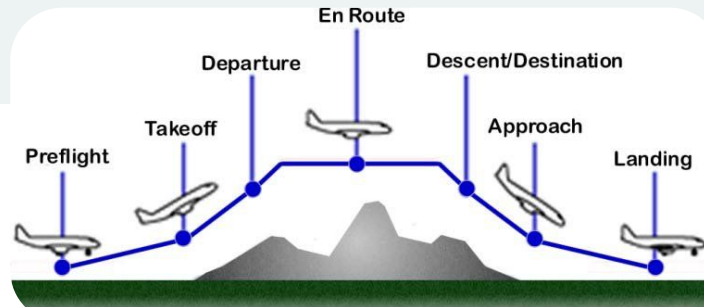
Oura Ring



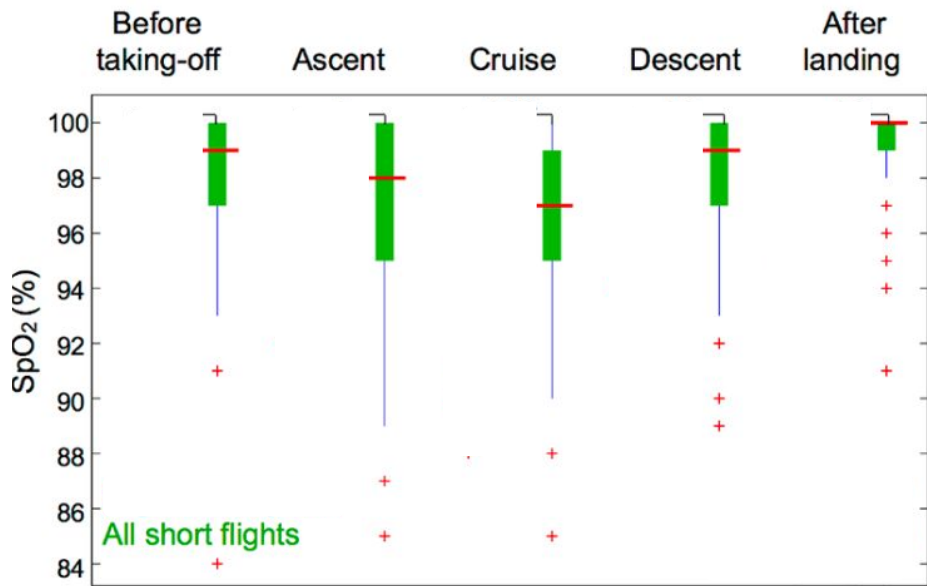
Empatica E4

Heart Rate, Heart Rate Variability, Blood Oxygen Level, Sleep, Skin Temperature, Steps, Respiration Rate, etc.

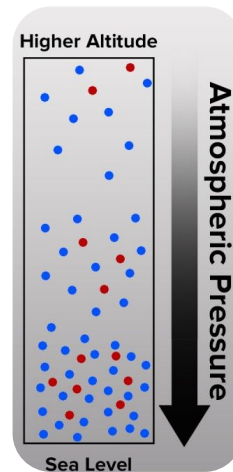
Flight Study: SpO2



Wu, Paul, et al. "Fuzzy multi-objective mission flight planning in unmanned aerial systems." 2007 IEEE Symposium on Computational Intelligence in Multi-Criteria Decision-Making. IEEE, 2007.

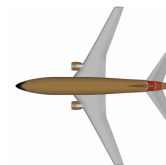


* Digital Health: Tracking Physiomes and Activity Using Wearable Biosensors Reveals Useful Health-Related Information (PLOS BIOLOGY 2017)



O₂ Oxygen ●
N₂ Nitrogen ●

"The current FAA regulations for limiting cabin pressures to 8,000-foot equivalent altitudes allow for mildly hypoxic conditions."

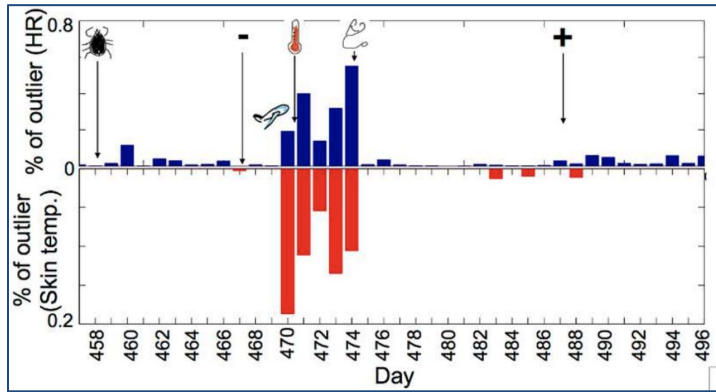


Source: https://www.faa.gov/data_research/research/m_ed_humanfacs/cer/media/HealthEffectsVulnerablePasengers.pdf

Source: <https://www.koaa.com/news/covering-colorado/oxygen-and-altitude-why-do-you-get-so-winded-when-you-climb>

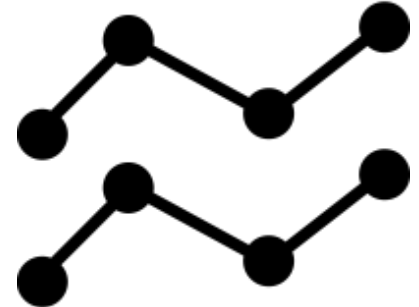
Wearables Data <> Infectious Disease

Lyme Disease



* Digital Health: Tracking Physiomes and Activity Using Wearable Biosensors Reveals Useful Health-Related Information (PLOS BIOLOGY 2017)

Heart rate / HRV /
Temperature / SpO2



Infectious Diseases

2017



Need: Wearable Data
Acquisition at Scale

Personal Health Dashboard (PHD)



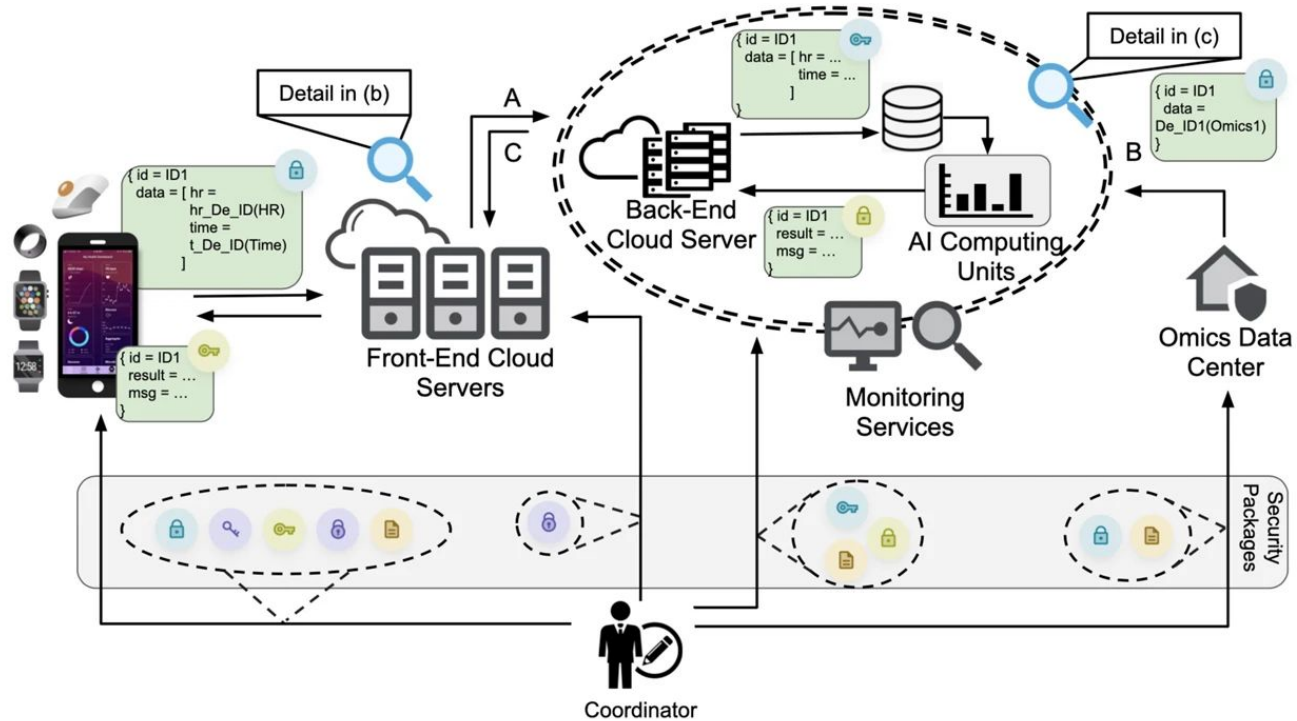
A scalable, secure, and interoperable platform for deep data-driven health management

Amir Bahmani^{1,2,3,7}, Arash Alavi ^{1,2,3,7}, Thore Buergele ^{3,7}, Sushil Upadhyayula^{1,3,4}, Qiwen Wang^{1,3,4}, Srinath Krishna Ananthakrishnan³, Amir Alavi³, Diego Celis^{1,3,4}, Dan Gillespie ³, Gregory Young^{1,3}, Ziyi Xing^{1,2}, Minh Hoang Huynh Nguyen^{1,2}, Audrey Haque^{1,2}, Ankit Mathur^{1,3,4}, Josh Payne ^{1,3,4}, Ghazal Mazaheri^{1,3}, Jason Kenichi Li^{1,3,4}, Pramod Kotipalli^{1,3,4}, Lisa Liao^{1,3,4}, Rajat Bhasin³, Kexin Cha^{1,3}, Benjamin Rolnik^{1,3}, Alessandra Celli¹, Orit Dagan-Rosenfeld¹, Emily Higgs¹, Wenyu Zhou^{1,2}, Camille Lauren Berry^{1,3}, Katherine Grace Van Winkle^{1,3}, Kévin Contrepois ¹, Utsab Ray^{1,2,3}, Keith Bettinger^{1,2}, Somalee Datta⁵, Xiao Li^{1,6}  & Michael P. Snyder ^{1,2,3} 

Nature Communications 2021



PHD Overview



PHD Case Study 1: COVID-19

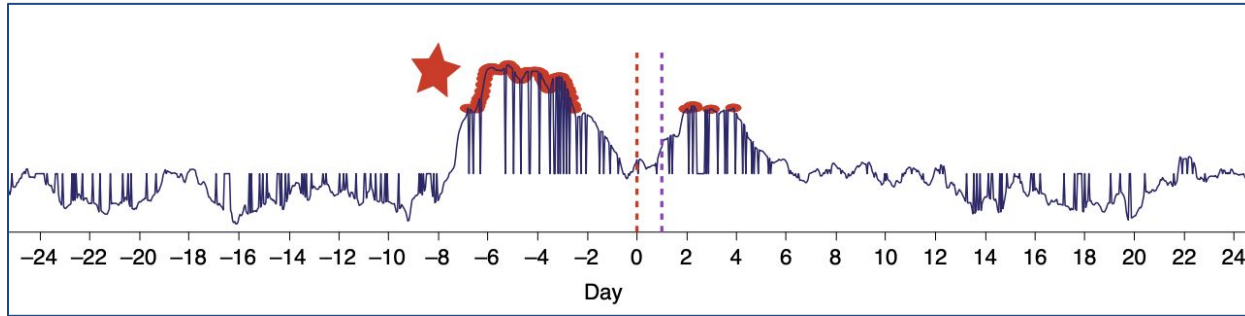


Pre-symptomatic detection of COVID-19 from smartwatch data

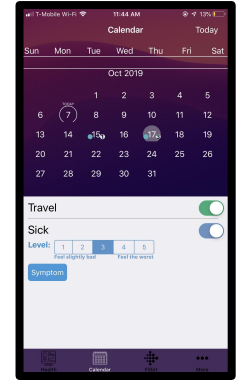
Tejaswini Mishra ^{# 1}, Meng Wang ^{# 1}, Ahmed A Metwally ^{# 1}, Gireesh K Bogu ^{# 1},
Andrew W Brooks ^{# 1}, Amir Bahmani ^{# 1}, Arash Alavi ^{# 1}, Alessandra Celli ¹, Emily Higgs ¹,
Orit Dagan-Rosenfeld ¹, Bethany Fay ¹, Susan Kirkpatrick ¹, Ryan Kellogg ¹, Michelle Gibson ¹,
Tao Wang ¹, Erika M Hunting ¹, Petra Mamic ¹, Ariel B Ganz ¹, Benjamin Rolnik ¹, Xiao Li ²,
Michael P Snyder ³

[Nature Biomedical Engineering 2021]

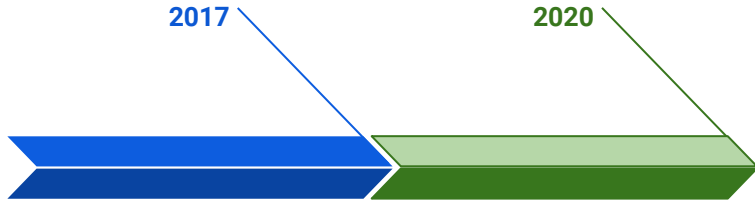
PHD Case Study 1: COVID-19



* Pre-symptomatic detection of COVID-19 from smartwatch data
(NATURE BIOMEDICAL ENG 2020)



MyPHD Phase 1

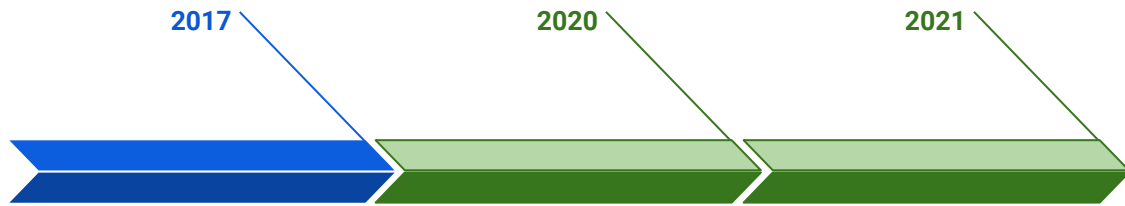
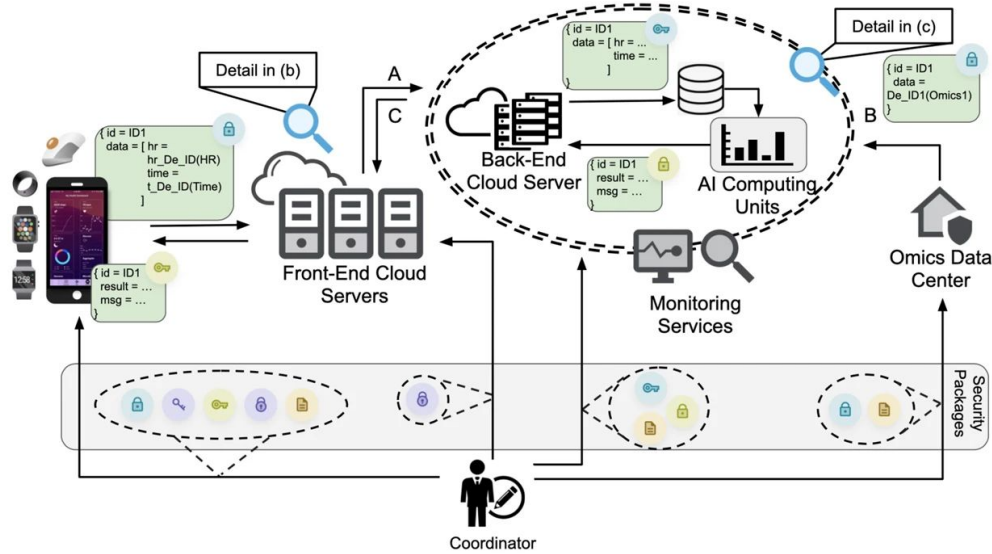


Need: Wearable Data Acquisition at Scale

Need: Retrospective COVID-19 Detection Algorithms + Data Collection Platform

- 5 months
- 30+ COVID-19 patients
- 5300+ MyPHD users
- Algorithm successfully detects COVID-19 up to 10 days and a median of 4 days before first reported symptom.

PHD: Secure Data Collection and Analysis at Scale



Need: Wearable Data Acquisition at Scale

Need: Retrospective COVID-19 Detection Algorithms + Data Collection Platform

Need: Secure Data Collection and Analysis at Scale MyPHD

PHD COVID-19 Alert System



Real-time alerting system for COVID-19 and other stress events using wearable data

[Arash Alavi](#), [Gireesh K. Bogu](#), [Meng Wang](#), [Ekanath Srihari Rangan](#), [Andrew W. Brooks](#), [Qiwen Wang](#), [Emily Higgs](#), [Alessandra Celli](#), [Tejaswini Mishra](#), [Ahmed A. Metwally](#), [Kexin Cha](#), [Peter Knowles](#), [Amir A. Alavi](#), [Rajat Bhasin](#), [Shrinivas Panchamukhi](#), [Diego Celis](#), [Tagore Aditya](#), [Alexander Honkala](#), [Benjamin Rolnik](#), [Erika Hunting](#), [Orit Dagan-Rosenfeld](#), [Arshdeep Chauhan](#), [Jessi W. Li](#), [Caroline Bejikian](#), [Vandhana Krishnan](#), [Lettie McGuire](#), [Xiao Li](#), [Amir Bahmani](#) ✉ & [Michael P. Snyder](#) ✉

[Nature Medicine 2022]

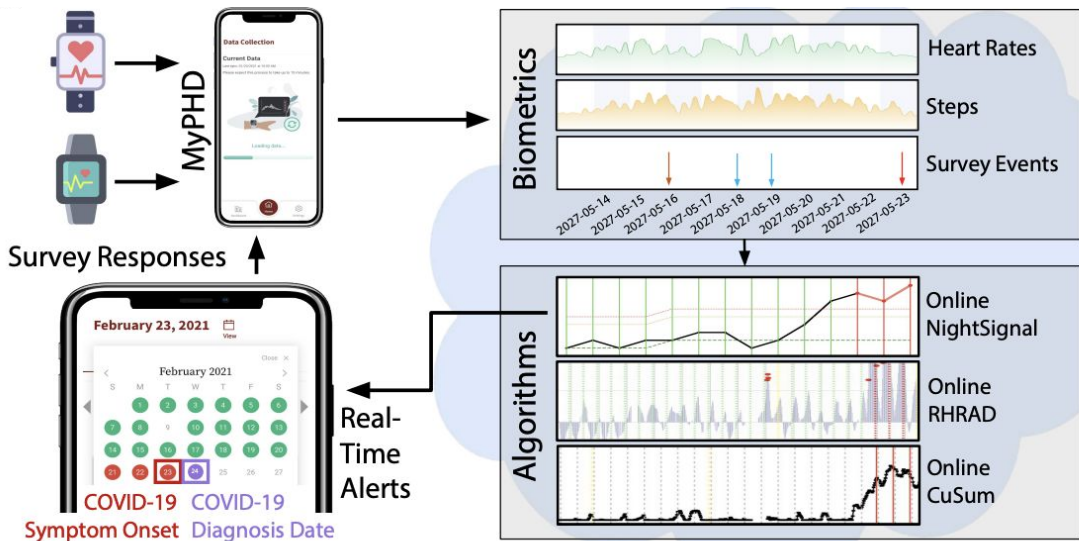


NightSignal



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MEDICINE

Real-time Infectious Disease Alerting



80%

early detection for pre-symptomatic and asymptomatic COVID-19 cases with a median of 3 days before symptom onset

2017

2020

2021

2021

Need: Wearable Data Acquisition at Scale

Need: Retrospective COVID-19 Detection Algorithms + Data Collection Platform


Need: Secure Data Collection and Analysis at Scale
MyPHD

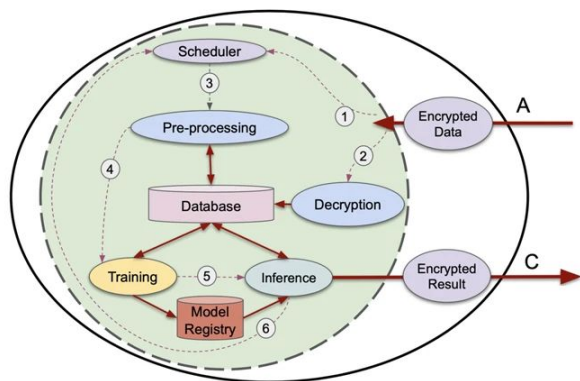
Need: First Real-time Alert System



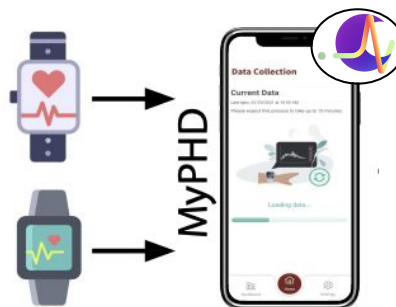
Stanford
MEDICINE

COVID-19 Phase 2: Technological Advancements

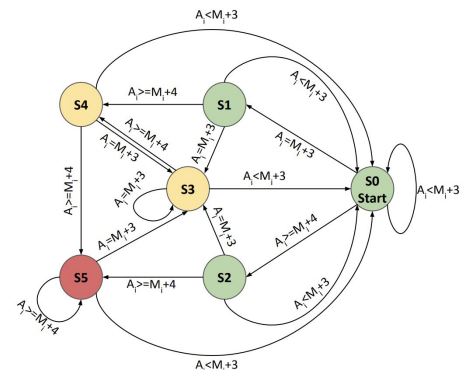
	NightSignal 	Anomaly detection (EllipticEnvelope)	CuSum Statistical Algorithm
Sensitivity	80%	69%	72%
Specificity	87.8%	87.6%	82.1%
Complexity	LightWeight	Heavy	Heavy
Sensitive to resolution	No	Yes	Yes



Machine Learning Cluster
it scales any AI/ML algorithms for wearable/multi-omics datasets on the cloud



Lightweight Algorithm
NightSignal can be executed on the phone
Low cost & useful for countries/places w/ slow Internet speeds



A_i : Average resting heart rate overnight for night i
 M_i : Median of averages of resting heart rate overnight for all nights upto night i

Alert State Machine

MyPHD Supports Multiple Research Studies

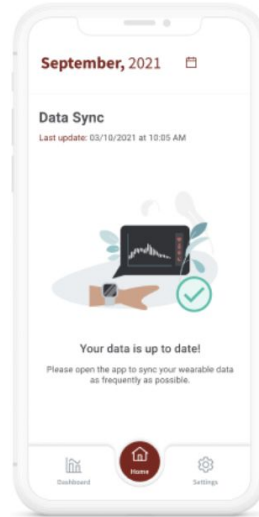


3:18

Studies

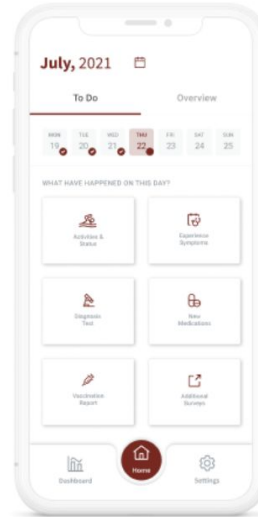
Studies Selection

- COVID-19 Wearable Study
- Penn/CHOP Transplant Study
- Stress and Resilience Study
- Crohn's Exposome Study
- Aviation Study
- Personalized Baseline Study
- Fiber Cognition Study
- Crash Course Study
- Autism Study
- iPOP Study



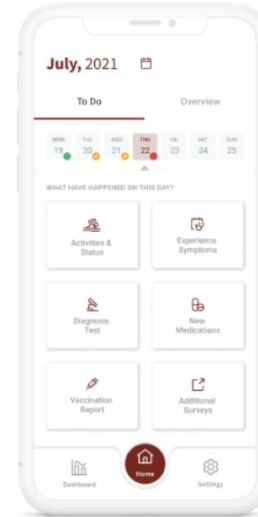
Package 1

- Wearable data collection
- Survey collection
- Data analysis
- Customized requests



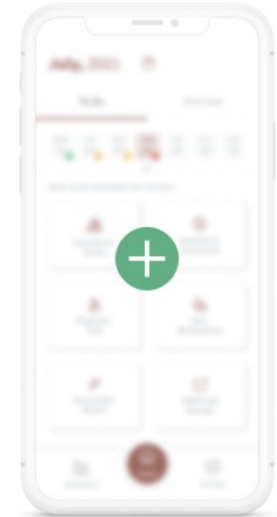
Package 2

- Wearable data collection
- Survey collection
- Data analysis
- Customized requests



Package 3

- Wearable data collection
- Survey collection
- Data analysis
- Customized requests



Package 4

- Wearable data collection
- Survey collection
- Data analysis
- Customized requests

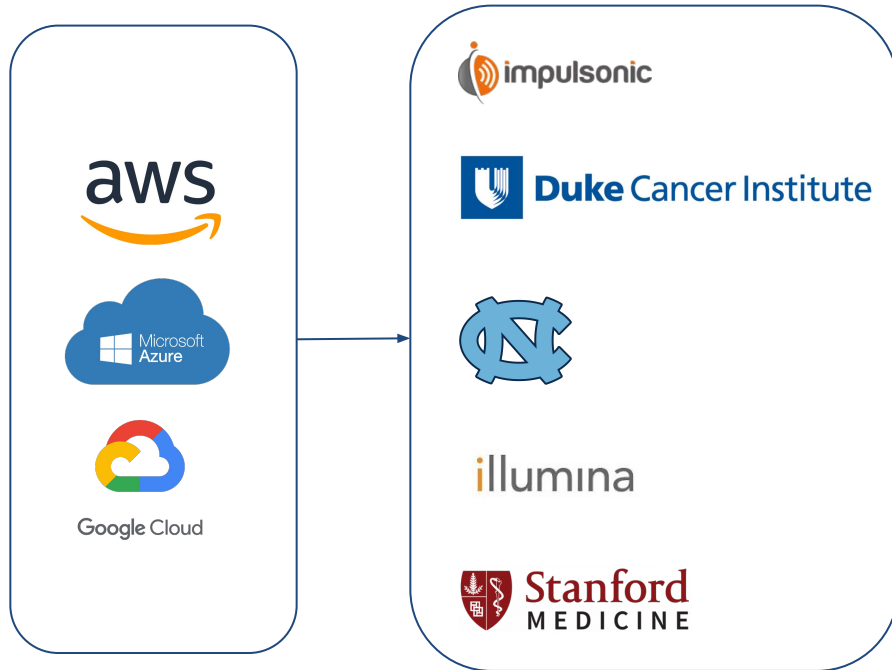
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Timeline

Collaboration

John C. Maxwell:
“Collaboration is multiplication.”



Problem #3: Understand the behaviour of our solution!

Think about scalability at an early stage



\$25,000

100X



\$250

1. Excessive Cost
2. If your implementation only works for B747, then you are restricted to only special runways

Hummingbird: Efficient Performance Prediction



<https://www.weforum.org/agenda/2020/11/formula-one-f1-innovation-ventilators-fridges/>



Hummingbird: efficient performance prediction for executing genomic applications in the cloud FREE

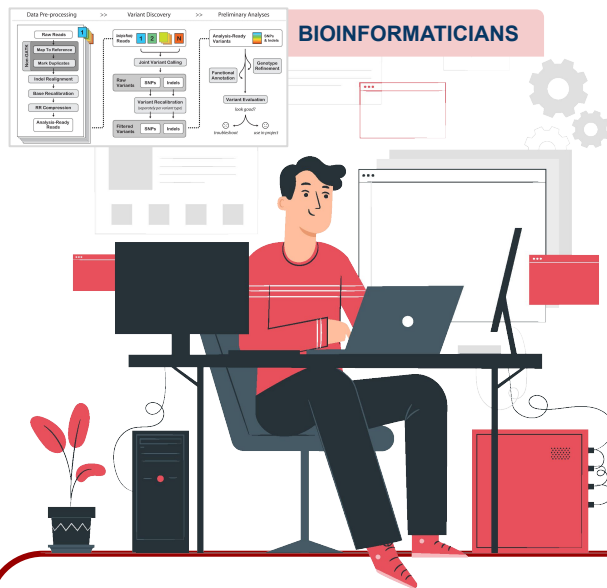
Amir Bahmani, Ziye Xing, Vandhana Krishnan, Utsab Ray, Frank Mueller, Amir Alavi, Philip S. Tsao, Michael P. Snyder ✉, Cuiping Pan ✉ [Author Notes](#)

[Bioinformatics 2021]



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Optimizing Computation and Storage



For a computational pipeline, what machine type is the:

- 1) Cheapest
- 2) Fastest
- 3) Cost-efficient



Optimizing Computation and Storage (cont'd)

This is the problem we're solving...

What do I choose?

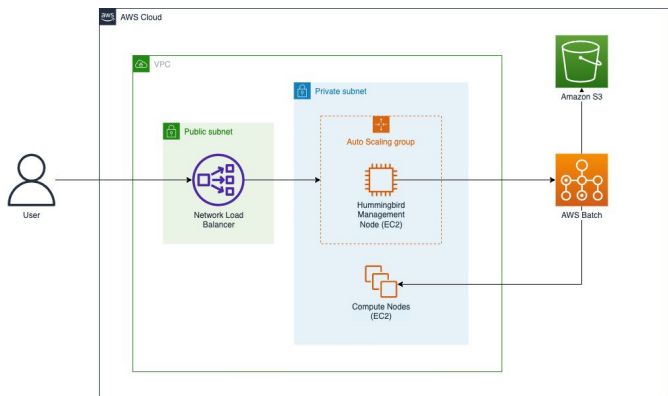
- Machine Configuration is complex (*CPU, Memory, I/O*)
- Even highly trained informaticians cannot optimize their configurations adequately

The screenshot shows the AWS Management Console interface for selecting an EC2 instance type. The page title is "Description" and there is a search bar. The instance types listed are:

Instance Type	Family	vCPU	Memory	On-Demand Linux pricing	On-Demand Windows pricing	Free tier eligible
t2.micro	t2	1	1 GiB	0.0116 USD per Hour	0.0162 USD per Hour	Yes
t2.small	t2	1	2 GiB	0.023 USD per Hour	0.032 USD per Hour	No
t2.medium	t2	2	4 GiB	0.0464 USD per Hour	0.0644 USD per Hour	No
t2.large	t2	2	8 GiB	0.0464 USD per Hour	0.0644 USD per Hour	No
t2.xlarge	t2	4	16 GiB	0.0928 USD per Hour	0.1288 USD per Hour	No

A red circle highlights the instance type selection area, and a red arrow points from the text "What do I choose?" to this area. The "Free tier eligible" status is visible for the t2.micro and t2.xlarge instances.

Hummingbird – a tool for effective prediction of performance and costs of genomics workloads on AWS - 2022



**Goutam
Nistala**

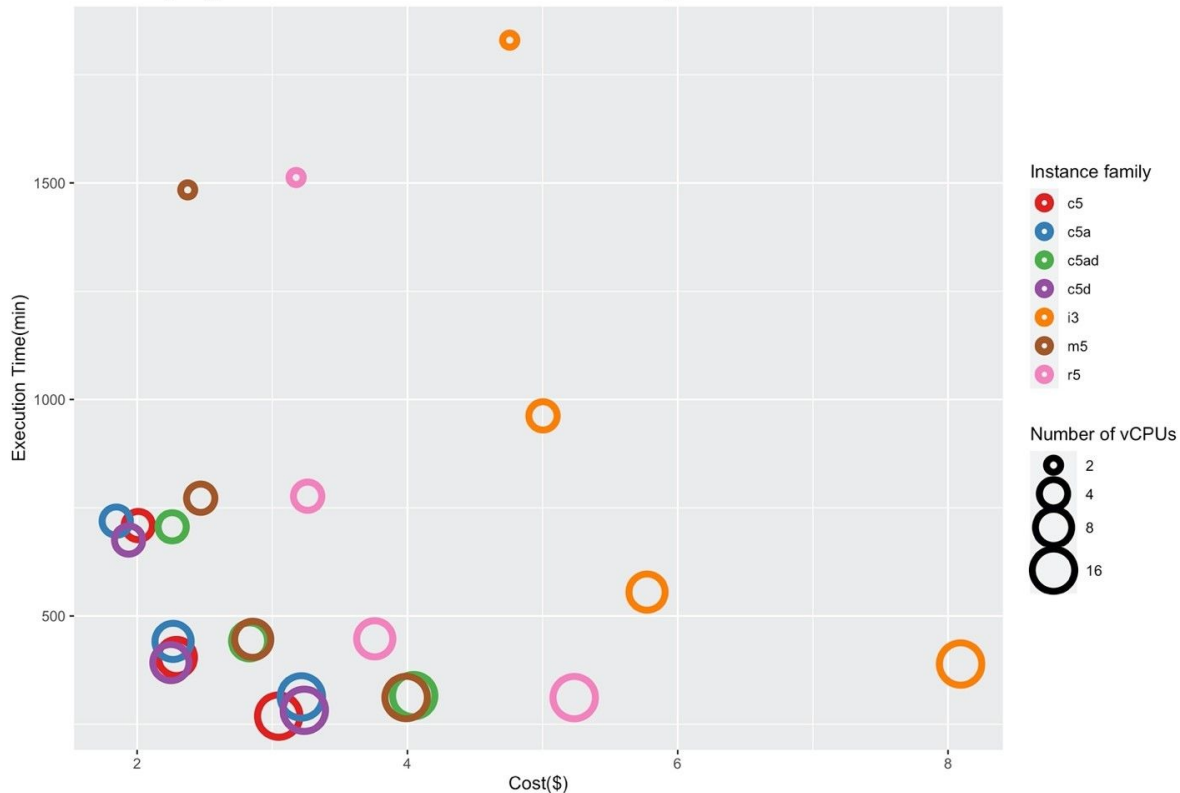


**Paul
Saxman**



**Sujaya
Srinivasan**

GATK HaplotypeCaller Runtime vs cost for various instance types



Problem #4: Centralized Databases Interoperability and Federated Computing

The NEW ENGLAND
JOURNAL of MEDICINE

Perspective

The Illusion of Inclusion — The “All of Us” Research Program and Indigenous Peoples’ DNA

Keolu Fox, Ph.D.

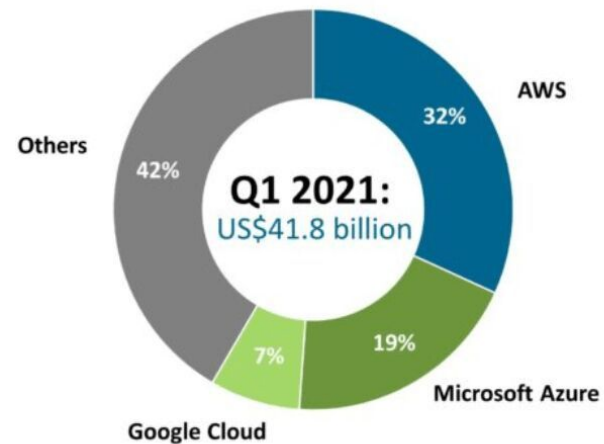
Article Figures/Media Metrics

5 References 43 Citing Articles Letters 1 Comment

July 30, 2021
N Engl J Med
DOI: 10.1056/NEJMp2107001

Related
CORRECTION
The “A

RAW DATA, INCLUDING DIGITAL SEQUENCE INFORMATION DERIVED FROM HUMAN genomes, have in recent years emerged as a top global commodity. This shift is so new that experts are still evaluating what such information is worth in a global market. In 2018, the direct-to-consumer genetic-testing company 23andMe sold access to its database containing digital sequence information from approximately 5 million people to GlaxoSmithKline for \$300 million. Earlier this year, 23andMe partnered with Almirall, a Spanish drug company that is using the information to develop a new anti-inflammatory drug for autoimmune disorders. This move marks the first time that 23andMe has signed a deal to license a drug for development.



Source: Public cloud market share statistics in 2022 by Enterprise Engineering Solutions, Inc. (EES)

Swarm: A Federated Cloud Framework



RESEARCH ARTICLE

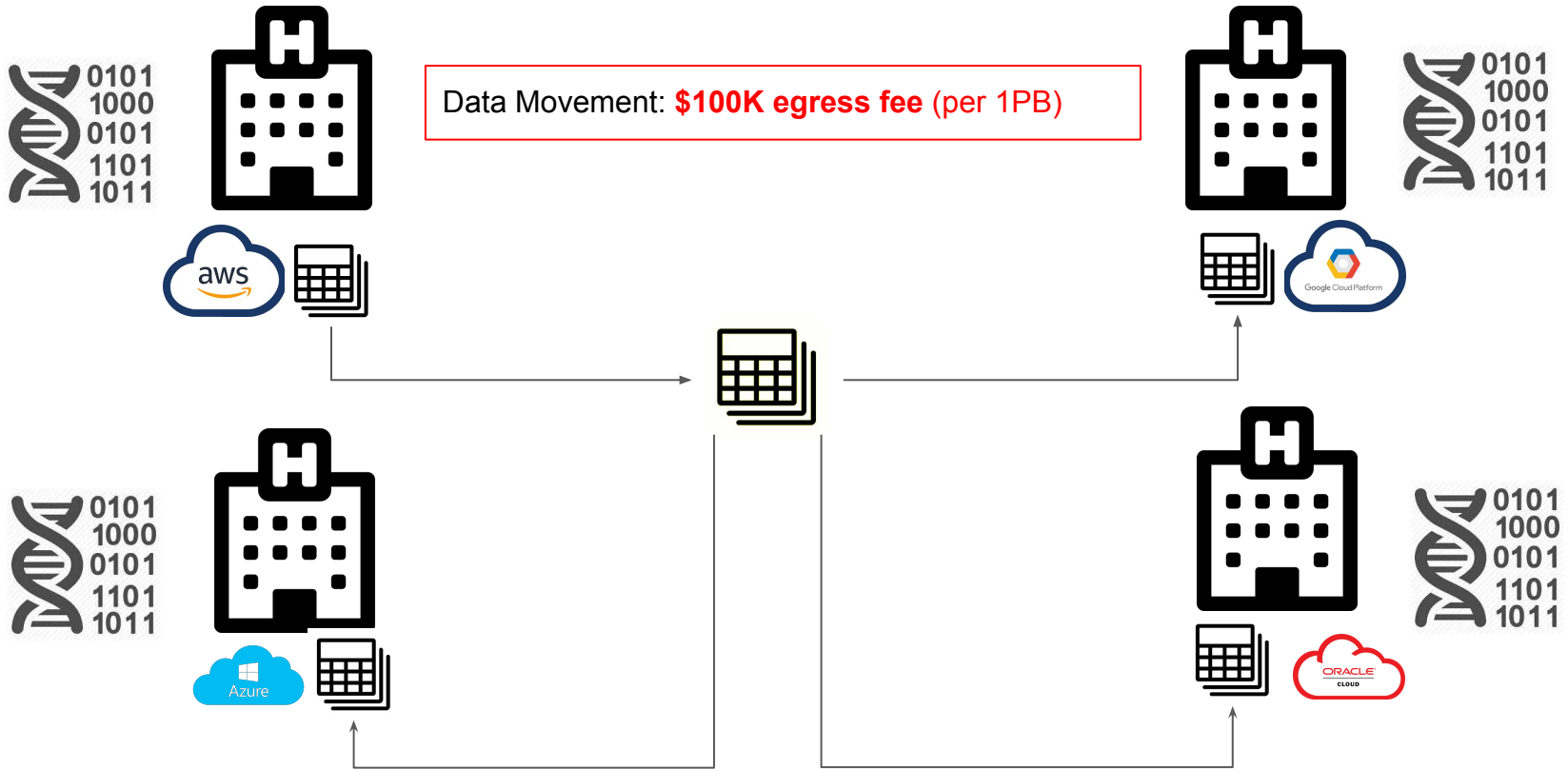
Swarm: A federated cloud framework for large-scale variant analysis

Amir Bahmani ^{1,2,3} , Kyle Ferriter^{2,3} , Vandhana Krishnan ^{2,3}, Arash Alavi^{2,3}, Amir Alavi^{2,3}, Philip S. Tsao ^{4,5}, Michael P. Snyder ^{1,2,3*}, Cuiping Pan ^{5*}

[PLOS Computational Biology 2021]



 **Stanford**
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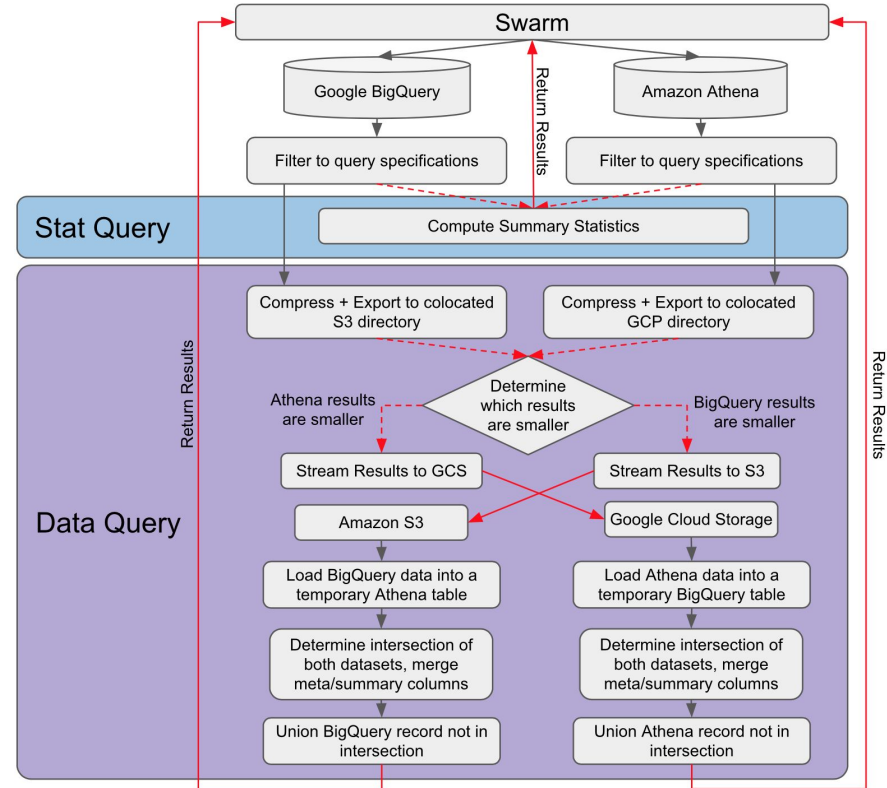


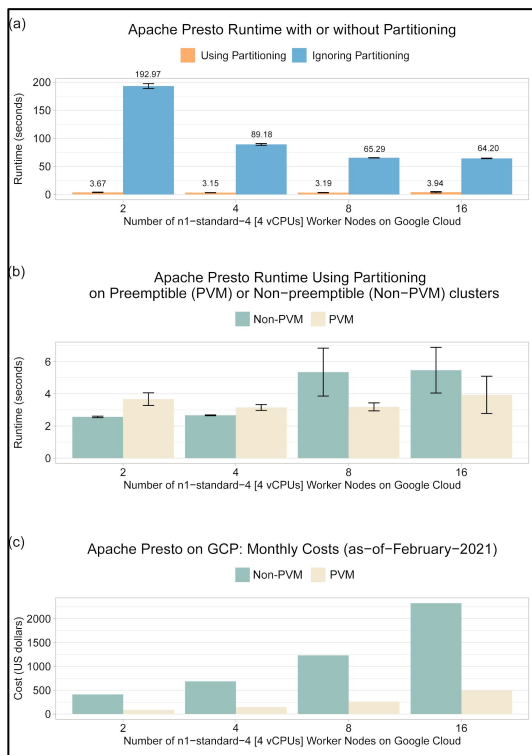
Data Redundancy: **4x Storage Fees**

Swarm Overview

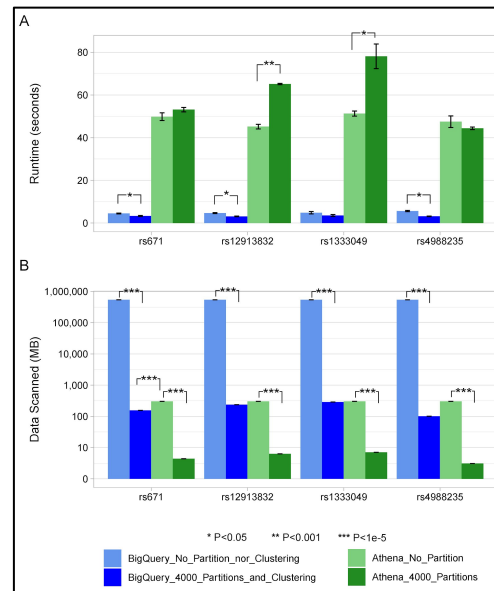
- Minimal data motion
- Reduction of costs, delays, and security and privacy risks.
- Applications:
 - Searching for particular genes or variants
 - Annotating large VCF tables
 - Allele frequency discrepancies between subject populations

Swarm: Federated computation promoting **minimal data motion** and facilitates crosstalk between genomic datasets stored on various cloud platforms





Average execution time and amount of data processed for computing allele frequency for an input set of rsIDs in Google BigQuery, Amazon Athena, Apache Presto



Description	rsID	Chr	Pos
Attention-deficit/hyperactivity disorder (ADHD)	rs671	12	112241766
Blue Eye Color (BEC)	rs12913832	15	28365618
Coronary Heart Disease (CHD)	rs1333049	9	22125503
Lactose Intolerance	rs4988235	2	136608646

- Minimize data motion: **500GB vs. 4KB**
- Facilitates model training without the need of sharing raw data, and therefore strengthens privacy protection.

Outline

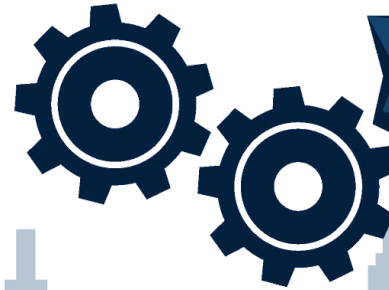
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Timeline

Serendipity: Meeting Late Prof. James L. Morrison



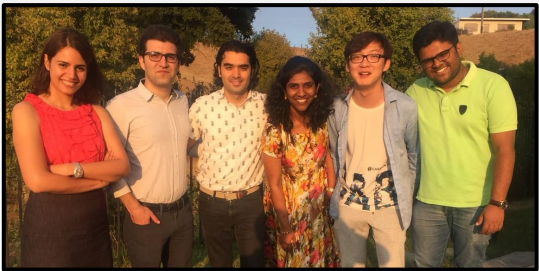
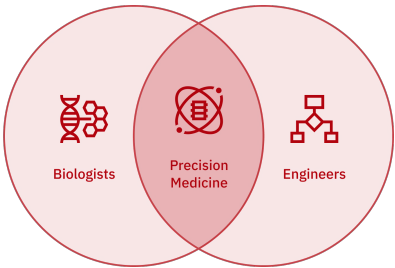
Precision Medicine and Education



SUCCESS

Stanford Data Ocean (SDO)

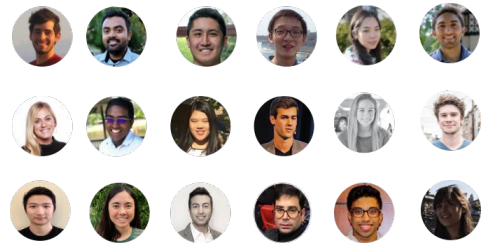
Educating a New Generation of Computer Scientists and Biologists



Summer Internship 2018



Summer Internship 2019



Summer Internship 2020

SPRING 2021

Cloud Computing for Biology and Healthcare

GENE222 / CS273C / BIOMEDIN222

Do you want to apply your coding skills in Precision Medicine?

Learn about Big Data in Precision Medicine in simple terms and learn how Cloud Computing and AI can transform it!

Contact: gene222-help@stanford.edu



SPEAKER
Mike Snyder
Chair of Genetics at Stanford University School of Medicine



SPEAKER
Mark Russinovich
CTO of Microsoft Azure



SPEAKER
Eric Topol
Founder & Director of Scripps Research Translational Institute



SPEAKER
Eric Schmidt
Former CEO & Chairman, Google Co-Founder, Schmidt Futures

...AND MANY MORE GUEST SPEAKERS THROUGHOUT THE QUARTER!

THE FUTURE OF HEALTH CARE STARTS HERE.



Stanford MEDICINE | Healthcare Innovation Lab

Steps:

- 1) Download datasets
 -) Large datasets?
- 2) Configuring the computing environments
 -) Download the code
 -) Install dependencies
 -) Configure input parameters/environmental variables
 -) Computing capabilities (e.g., GPU)?



Article | [Published: 18 November 2020](#)

Pre-symptomatic detection of COVID-19 from smartwatch data

[Tejaswini Mishra](#), [Meng Wang](#), [Ahmed A. Metwally](#), [Gireesh K. Boqu](#), [Andrew W. Brooks](#), [Amir Bahmani](#), [Arash Alavi](#), [Alessandra Celli](#), [Emily Higgs](#), [Orit Dagan-Rosenfeld](#), [Bethany Fay](#), [Susan Kirkpatrick](#), [Ryan Kellogg](#), [Michelle Gibson](#), [Tao Wang](#), [Erika M. Hunting](#), [Petra Mamic](#), [Ariel B. Ganz](#), [Benjamin Rolnik](#), [Xiao Li](#) & [Michael P. Snyder](#)

Nature Biomedical Engineering 4, 1208–1220 (2020) | [Cite this article](#)

103k Accesses | **78** Citations | **1391** Altmetric | [Metrics](#)

Access counts - The number of times an article has been accessed on SpringerOpen or BioMed Central

Data availability

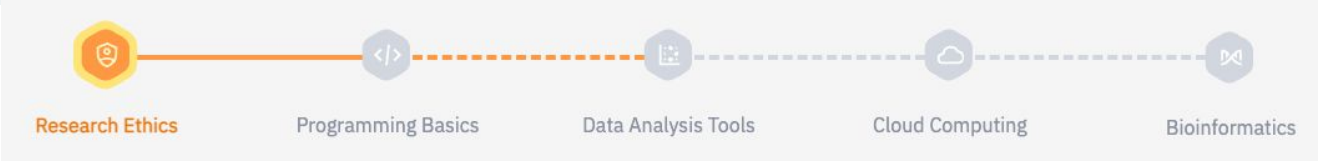
The de-identified raw heart rate, steps and sleep data used in this study can be downloaded from the study data repository (https://storage.googleapis.com/gbsc-gcp-project-ipop_public/COVID-19/COVID-19-Wearables.zip). Processed data, including algorithm outputs and the data used for plotting the figures are provided as Supplementary Data 1.

Code availability

Code for the algorithms used in this manuscript is available at <https://github.com/mwgrassgreen/WearableDetection> (RHR-Diff and CuSum) and <https://github.com/gireeshkbogu/AnomalyDetect> (HROS-AD).



Stanford Data Ocean (SDO): Bioinformatics



Learn fundamental concepts

Follow a study roadmap that leads to your goal.

Bioinformatics	Clinician	Cloud Engineer
I want to learn essentials to have a solid foundation in precision medicine.	I want to quickly learn how to get insights from patients' data at work.	I want to get into cloud computing in healthcare settings.
18 Modules	9 Modules	8 Modules
2 Diplomas	2 Diplomas	2 Diplomas

Research Ethics Programming Basics Data Analysis Tools Cloud Computing Bioinformatics

01 Introduction to Genomics
Learn how to process and analyze Variant Call Format (VCF) file using Amazon Athena.

02 Introduction to Epigenetics
Understand changes caused by modification of gene expression, rather than alteration of genetic code.

04 Introduction to Proteomics
Get familiar with the large-scale study of proteins in an organism.

03 Introduction to Transcriptomics
Understand the study of transcripts/mRNAs and their functions.

05 Introduction to Statistics II
Concepts that comes handy when you start doing data analysis.

06 Introduction to Microbiome
Get familiar with the study of the genetic material of all the microbes.

08 Introduction to Clinical Data
Analyze clinical data that includes health assessments, diagnoses, illness and documentation of care delivery.

07 Introduction to Metabonomics
Examine recent trends: metabolites changes in body fluids or tissues to learn analytical techniques.

09 Introduction to Imaging
Get into medical imaging, and study examples on how to process MRI images.

10 Introduction to Wearable Data
Learn how to analyze data collected from wearables devices.

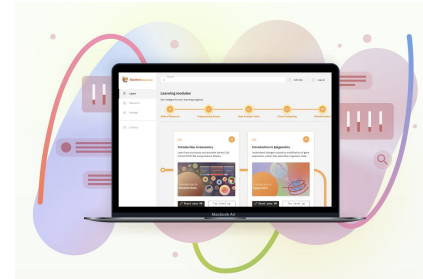
11 Introduction to Exposome
Analyze the environmental exposures that an individual is subjected to from conception to death.

Bioinformatics
Diploma for complex training modules.

John Doe
I awarded this diploma as a proof of the learning completion by Stanford Data Ocean on August 29, 2022.

You've completed training for **Bioinformatics!**
All the training diploma as proof of learning completion by Stanford Data Ocean, and includes to demonstrate your qualifications.

📄 [Download Diploma](#)



EARLY ACCESS



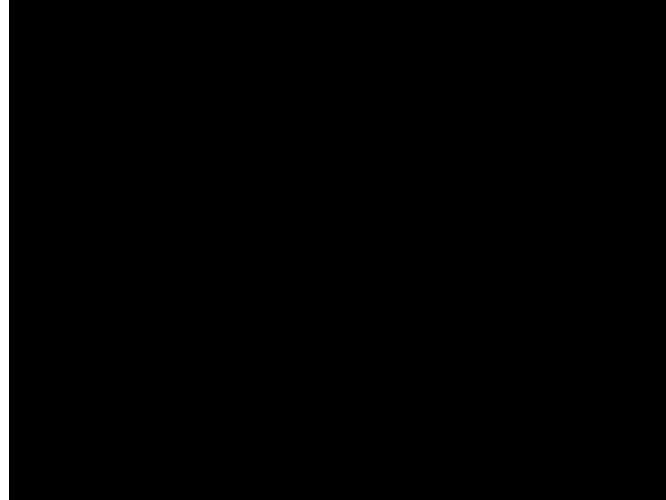
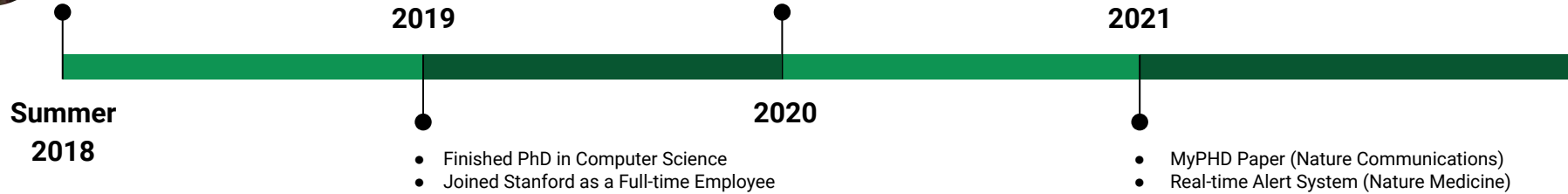
Internship + Course Impact



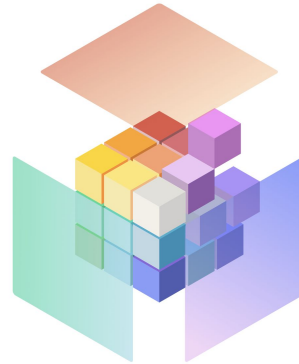
Personal Health Dashboard

- Started Working on the Security and Privacy Concerns of MyPHD Apps

- Helped Scaling the COVID-19 Study on MyPHD Apps (Nature Biomedical Engineering)
- Presented at the CyberSecurity Festival



Visit our website and learn
about our projects! We are
always looking for
collaborators!



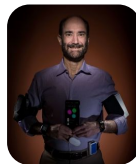
Special Thanks to:

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Stanford
Associate Chief of Staff for Precision
Medicine, VA Palo Alto Health Care
System



Late James L. Morrison,
Professor of Educational
Leadership, UNC

Peter Knowles
Pramod Kotipalli
Jessi W Li
Qiwen Wang
Quentin Hall
Rajat Bhasini
Ramesh Nair
Shrinivas Panchamukhi
Srinath Krishnan
Sujaya Srinivasan
Sushil Upadhyayula
Tadeu Perillo
Tao Wang
Tejaswini Mishra
Utsab Ray
Vandhana Krishnan
Wenyu Zhou
Xiao Li
Ziye Xing



Google Cloud



**MILLION
VETERAN
PROGRAM**



**Stanford
MEDICINE**



Deep Data
Research Center



MyPHD Platform



SDO Platform



Swarm



PHD



COVID19
Phase 1

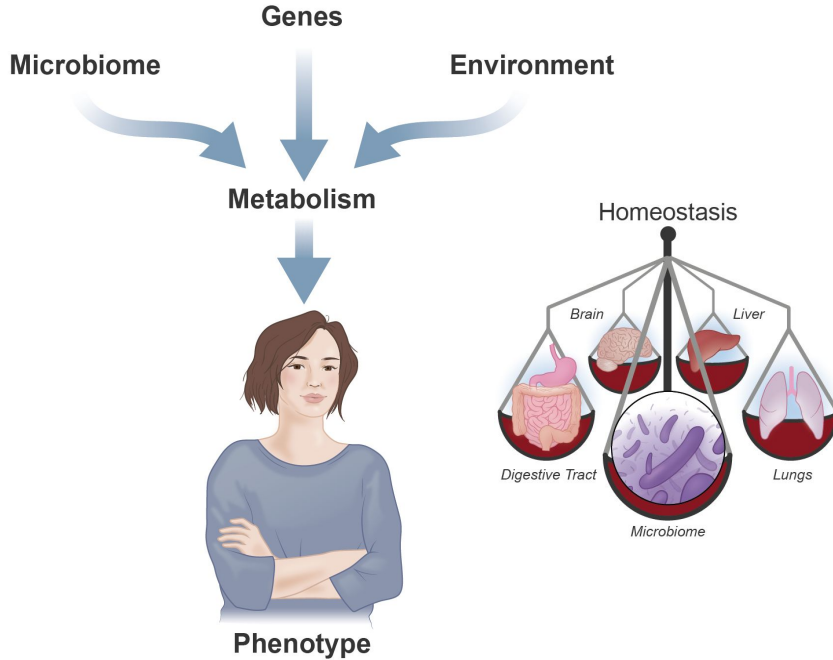


Alert
COVID-19
Phase 2

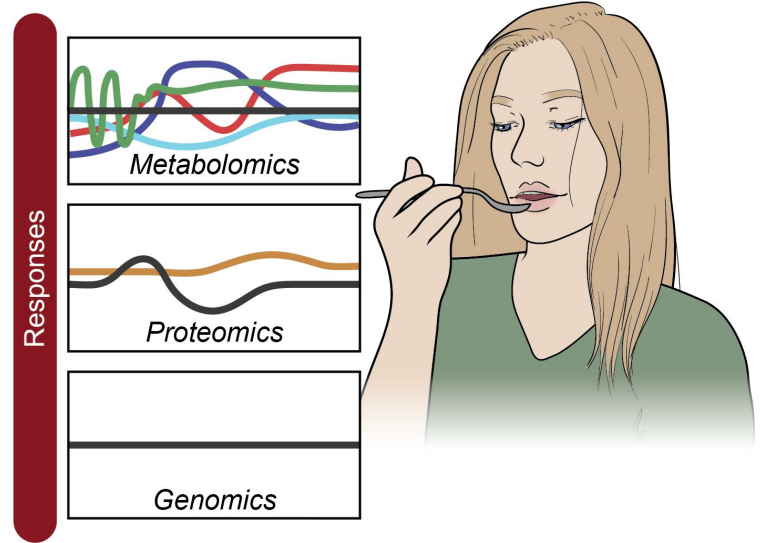


Hummingbird

Find the effective resolution+window



Metabolism: Conversion of food/fuel to energy to run cellular processes



Metabolomics is more time sensitive than other “omics”

Latency-sensitive vs. Throughput-oriented

Applications can be grouped into two broad classes:

Latency-sensitive workload
(User-facing front-end applications)

A job might require a latency of 100 seconds for uploading daily heart-rate data



Throughput-oriented workload
(Internal batch analytics frameworks)

A batch job might require a throughput of 10000 genomes per day

