

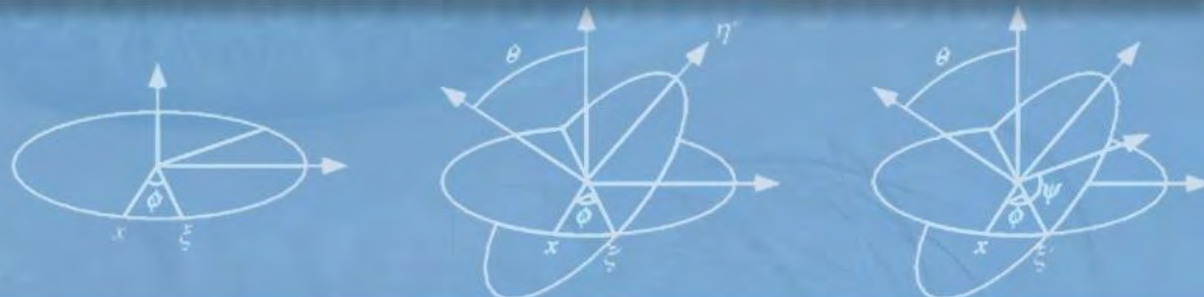


JHU vision lab

Opportunities for Accelerating Discovery: Mathematical and Algorithmic Issues

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THE DEPARTMENT OF BIOMEDICAL ENGINEERING

The Whitaker Institute at Johns Hopkins



JOHNS HOPKINS
MATHEMATICAL INSTITUTE
for DATA SCIENCE

AI 3.0

- AI 1.0: build mathematical/physical/biological/symbolic models
- AI 2.0: use big data to learn black-box (deep) models
- AI 3.0: develop new generation of AI methods that integrate mathematical/physical/biological/symbolic models & data-driven models.

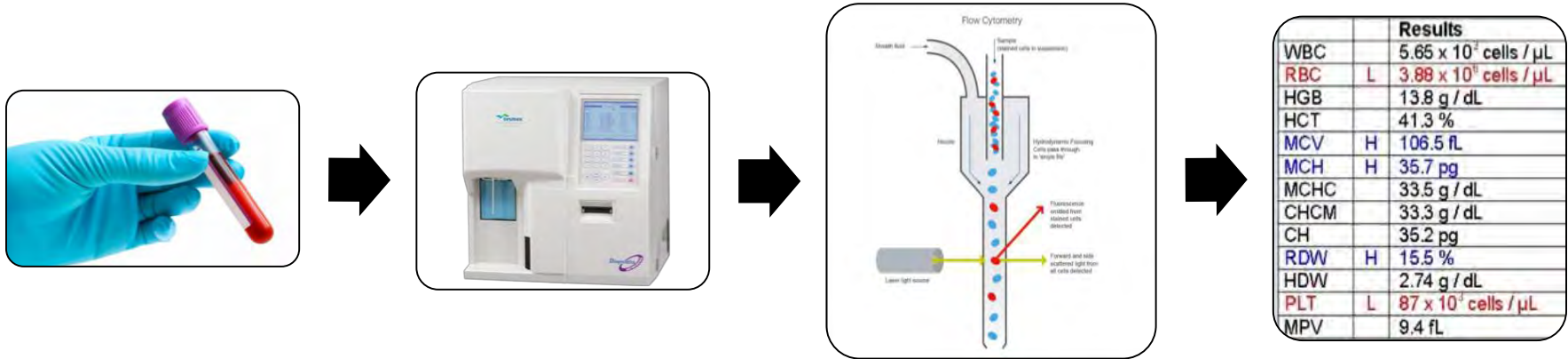
- Explainable
- Robust
- Scalable



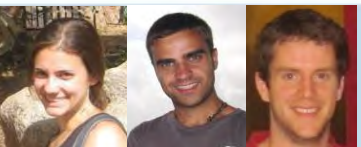
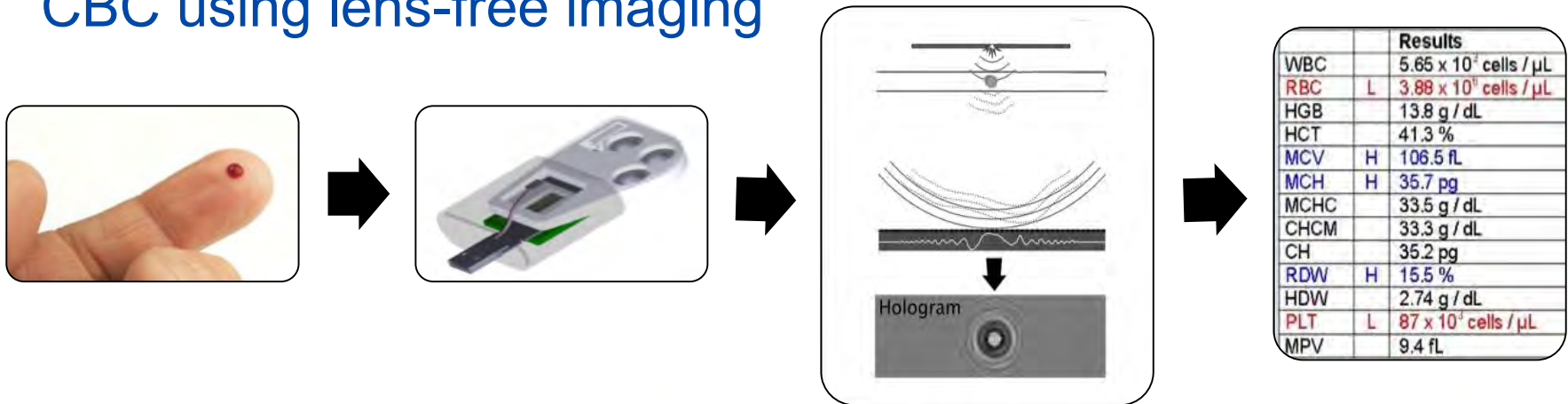
- Safe
- Adaptable
- Lifelong

Example: AI for Point of Care Testing

- Complete Blood Count (CBC) using hematology analyzer

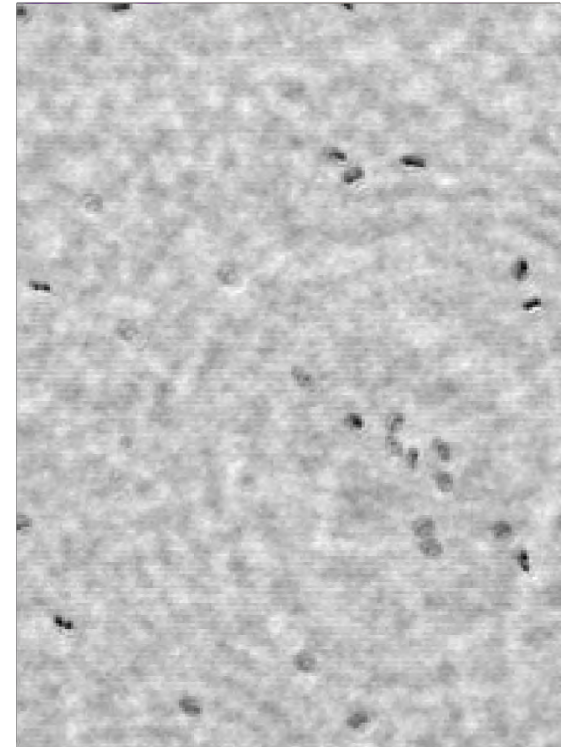
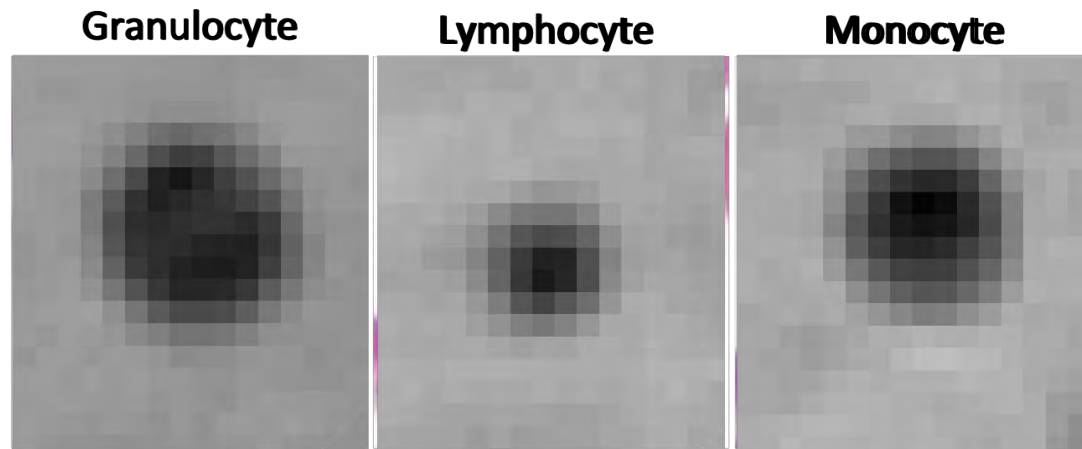


- CBC using lens-free imaging



Example: AI for Point of Care Testing

- Number of training images/videos, patients, chips needed to achieve accurate/repeatable performance?



- Data cleaning/annotation is very costly
- To succeed, exploiting biological knowledge was essential

ARO MURI: Semantic Information Pursuit

MURI: Semantic Information Pursuit for Multimodal Data Analysis



[Overview](#) [Research](#) [People](#) [Publications](#) [Meetings](#)

FACULTY INVESTIGATORS

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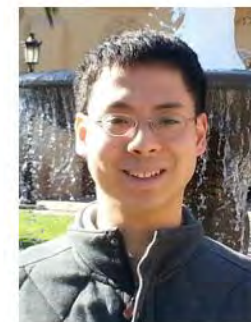
DONALD GEMAN



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ARNAUD DOUCET



MARK GIROLAMI



JOSEF KITTLER



SIMONE SEVERINI



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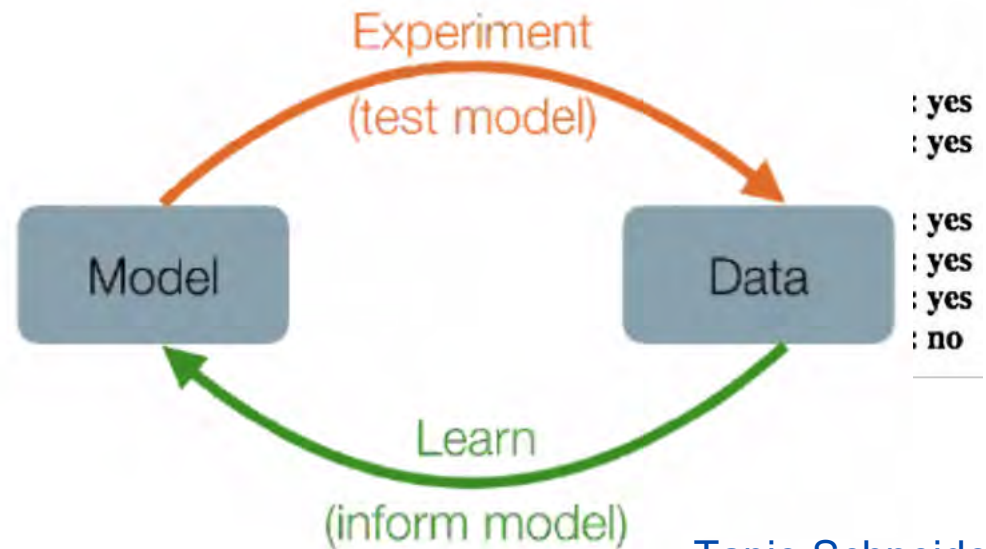


ARO MURI: Semantic Information Pursuit

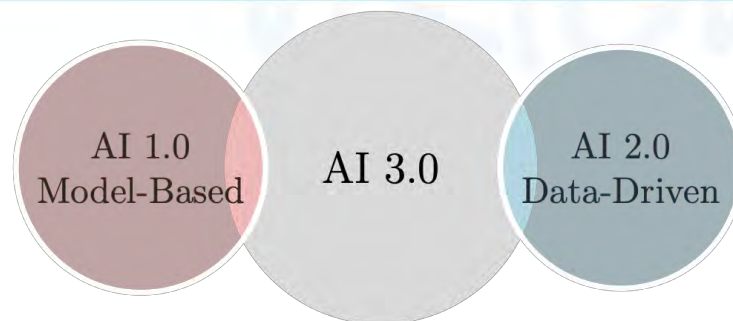
- Develop an information-theoretic framework for characterizing semantic information content in complex **multimodal** data
- Semantic information = expected number of queries
- Can we use it to determine what experiment to conduct next?



1. Q
2. Q
3. Q
4. Q
5. Q
6. Q
:



Tapio Schneider



- Integration of generative models, graphs, geometry, topology & deep learning
- Integration of mechanistic models & ML for accelerating discovery in biology, neuroscience & personalized medicine
- Integration of PDEs with deep learning for accelerating scientific discovery (e.g., physics, chemistry, geosciences)
- Integration of dynamical systems and reinforcement learning for designing the next generation of autonomous system