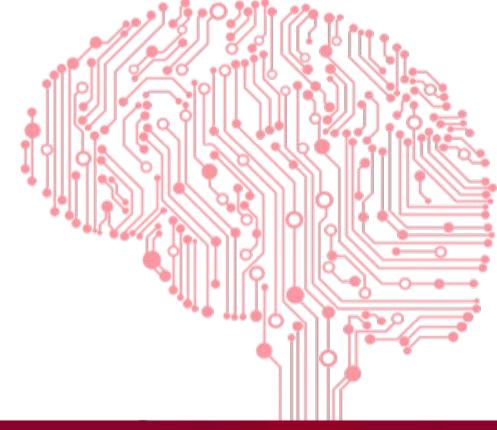
# Lightning Introductions

Research Interfaces between Brain Science and Computer Science

**December 3-5, 2014** 

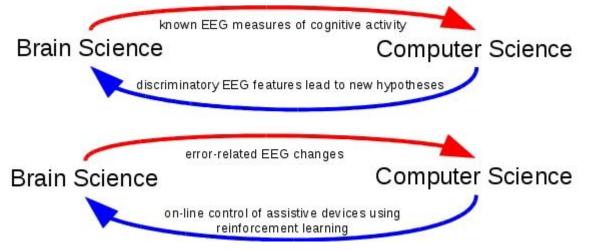




#### Charles Anderson / Colorado State



**Brain-Computer Interfaces** 





www.cs.colostate.edu/eeg



# Sanjeev Arora / Princeton





Computational complexity, designing algorithms for NP-hard problems, provably correct and efficient algorithms for ML (esp. unsupervised learning)

(Panel Moderator: Computing and the brain.)



# Satinder Singh Baveja / Michigan





- Reinforcement Learning: Architectures for converting payoffs/rewards to closed-loop behavior in Als and Humans
- Optimal rewards theory, or, Where do reward (functions) come from?

 Computationally Rational models for explaining animal behavior and for deriving brain mechanisms.



#### Andrew Bernat / CRA



How might computer science and brain science get the resources they need?



#### Matt Botvinick / Princeton



- Cognitive/computational neuroscience
  - fMRI, behavioral methods, neurophysiology
  - Computational modeling (deep learning, reinforcement learning, graphical models)
- Perspective: Computation as a Rosetta Stone
  - •A common language in which to understand both behavior/cognition and neural function





#### Randal Burns / JHU

CS->Brain: Data-Intensive Web-services

- scale infrastructure to capture high-throughput imaging (1TB/day)
- integrated visualization and analytics
- semantic/spatial queries of brain structure/function





BRAIN->CS: Inspiration for new data organization and indexing techniques

OPEN CONNECTOME PROJECT

COLLECTIVELY REVERSE-ENGINEERING THE BRAIN ONE SYNAPSE AT A TIME.



# Miyoung Chun / Kavli Foundation



The benefits are *mutual*:

- Understanding the brain will have tremendous impacts on
  - hardware (e.g. neuromorphic computing) and
  - software (e.g. image recognition, machine learning, algorithms).
- Computer Science developments are crucial to
  - analyze the data and discover the brain's circuits, and
  - aggregate, share, and collectively analyze diverse and heterogeneous neuroscience datasets.





### Christos Davatzikos / UPenn



UPENN
Center for Biomedical
Image Computing and
Analytics

- --Brain Image Analysis
- --Machine Learning and Imaging Pattern Analysis



#### Susan Davidson / UPenn





**CS**→ **Brain:** Novel information/analysis challenges

- "Mesoscale data" but lots of it: do databases help?
- Integrating many different types of data (e.g. image, genomic, hospital/patient)
- Privacy and security issues
- Reproducibility, data provenance, data publishing

**Brain**→ **CS:** Implications for computation/information organization and retrieval?



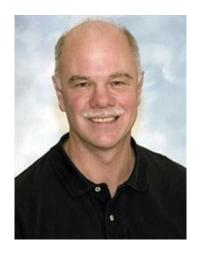
#### Ann Drobnis / CCC





Continued open engagement across the disciplines







#### Jim Duncan / Yale

#### Biomedical Image Analysis

- model-based strategies
- Bayesian/machine learning approaches
- application areas of interest include neuro-, cardiovascular and biological (microscopy) problems



# Naomi Feldman / Maryland





Cognitive models of language acquisition and processing

- Constructing cognitive models that draw on corpora and analysis techniques from automatic speech recognition
- Using models of early language acquisition to inform zero- and lowresource speech technologies



# Vitaly Feldman / IBM Research



IBM Research Foundations of machine learning: models, algorithmic and statistical complexity, robustness

Learning processes in nature:

Concept representation and learning in the brain Learning via evolution



#### Charless Fowlkes / UC Irvine



Applying computer vision and machine learning techniques to build robust, reusable tools for biological image and shape analysis Understanding computational role of feedback, mid-level representation and ecological statistics in visual processing





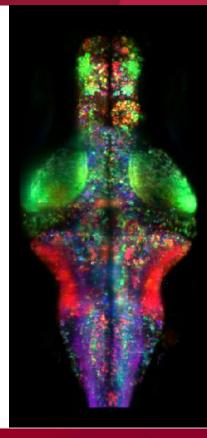
# Jeremy Freeman / HHMI



Using large-scale data analytics, interactive visualization, and experimental design to map brain activity in mice, fish, and flies.



Developing open source technologies for a **modern**, **scalable**, and **collaborative** neuroscience.





#### Shafi Goldwasser / MIT





Complexity Theory,
Cryptography,
Property Testing, Fault Tolerance
in Distributed Computing,
Randomness







#### Polina Golland / MIT

#### Biomedical Image Analysis

- Anatomical and functional variability from non-invasive imaging
- Functional organization of the brain
- Models of pathology
- Joint modeling of imaging and genetics



# Greg Hager / JHU



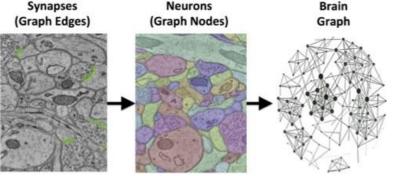


Are there universal "motifs" to brain structure and function? Do they lead us toward new models for computational perception and cognition?

Synapses

Neurons

Brain





# Jim Haxby / Dartmouth / CIMeC(Trento)









- MVPA decoding neural representations from fMRI
- Hyperalignment building a common model of representational spaces in human cortex
- HyperCortex a functional brain atlas based on a high-dimensional common model of neural representational spaces

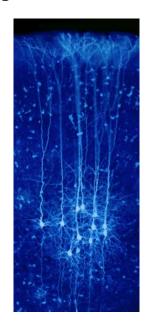


# Sean Hill / Human Brain Project





An open collaborative platform for large-scale data-intensive brain research bridging brain structure and function from genes to cognition - using semantic/spatial search, multimodal data integration, provenance tracking, analysis, machine learning, visualization, modeling and simulation.





# Vasant Honavar / Penn State University





- What principles govern (in both brains and machines):
  - Learning from experience
  - Learning from multimodal, multi-scale data?
  - Eliciting causal from disparate observations and experiments?
- How can we infer brain network structure and predict behavior from brain activity?



# Konrad Koerding / Northwestern





How might computer science and brain science benefit from one another?



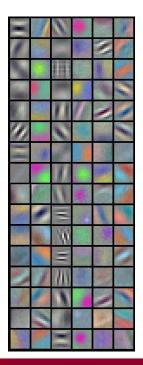
#### Yann LeCun / NYU & Facebook Al Research



facebook



- What are the underlying principles of learning, natural and artificial?
- How does the brain perform unsupervised learning?
- What is the neural basis of reasoning and planning?
- What are the essential architectural components?





# Richard Lewis / Michigan





- Computational cognitive science:

  Coordinated modeling + human experiments
- Computational rationality/bounded optimal control approaches to language, eyemovements, memory, choice..
- Reinforcement learning: optimal rewards



#### Chris Martin / Kavli Foundation



- The question is not how the two fields will benefit each other, but how to incentivize and build upon the substantial overlap that already exists!
- By embracing and adopting the ideas coming from both sides, this room already embodies that hybrid approach.
- 100 years from now will there be (or should there be) a distinction between Computer Science and Neuroscience? When your futuristic handheld neuromorphic device gets depressed who will you take it to?



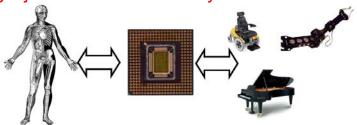


#### Sandro Mussa-Ivaldi / Northwestern





- How are representations of the world geometry and dynamics formed and updated through processes of sensory motor learning?
- Computational primitives and dimensionality reduction in sensorymotor control
- Human/Computer interfaces
- Synergistic interaction between human and machine learning
- Human/machine interactions for the recovery of motor functions following injuries to the nervous system





# Sheila Nirenberg / Cornell





 Understanding the codes neurons use and the transformations they perform

 Using this understanding to build neuroprosthetics, brain/machine interfaces, and robots



#### Aude Oliva / MIT





To further develop and fully engage with artificial systems at human-cognitive levels, we must understand cognition itself, and how it is mediated by the brain.



# Bruno Olshausen / UC Berkeley

- Theories of sensory coding
- Natural scene statistics
- Sparse representation
- Hierarchical representation and inference in cortical circuits





# Christos Papadimitriou / UC Berkeley





Algorithms and Complexity.

Game Theory and Economics.

Evolution.

How can algorithmic thinking help make progress in the sciences



#### Pietro Perona / Cal Tech



- Computer Vision
- Machine Learning
- Visual perception
- Neural computation
- Behavior





## Hanspeter Pfister / Harvard





HARVARD
School of Engineering
and Applied Sciences

- Visualization / Graphics / Vision
- Connectomics

Reconstructing brain circuits at the nanoscale using CS methods will allow deduction of interesting general organizational principles that in the long term will benefit CS.



#### Tal Rabin / IBM Research





- Cryptography Research
  - MultipartyComputations
  - Threshold and Proactive Security



# Rajesh Rao / U. Washington





Computer Science
Models, Algorithms, Devices

Predictive coding
Bayesian inference and learning
Acting under uncertainty
Brain-Computer Interfaces

Efficient approximate algorithms for AI, ML, and Robotics Novel Computer Interfaces

Brain Science
Neural Mechanisms





## Giulio Sandini / Italian Institute of Technology



On the left



Robotics, Brain and Cognitive Sciences

**Interests**: the development of <u>sensorimotor coordination</u> and <u>social interaction</u> by studying humans and building artificial systems:

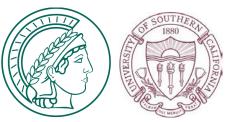
- Developmental Robotics
- Motor Cognition (Interaction, Prediction and Communication)
- Multimodal Sensory Integration.
- Motor Rehabilitation and Social Inclusion

**CS-BS Mutual benefits:** by sharing questions and space and discussing the "big picture" (how to assemble the brain puzzle from a topological and functional perspective)



#### Stefan Schaal / MPG / USC





Research interests include topics of statistical and machine learning, neural networks, computational neuroscience, functional brain imaging, nonlinear dynamics, nonlinear control theory, and biomimetic robotics. Applications to problems of artificial and biological motor control and motor learning, focusing on both theoretical investigations and experiments with human subjects and anthropomorphic robot equipment



# Nicolas Schweighofer USC





#### Computational Neurorehabilitation:

Brain -> CS: Models of motor learning and recovery of the lesioned brain

CS -> Brain: With predictive models, optimize re-training of the lesioned brain



#### Fei Sha / USC





Machine learning, with recent focus on

- How to derive useful representations from data automatically?
- How to robustly cope with changing environments w/o human intervention?

#### Perspective

The comparative study between artificial and biological learning systems will advance both fields significantly.

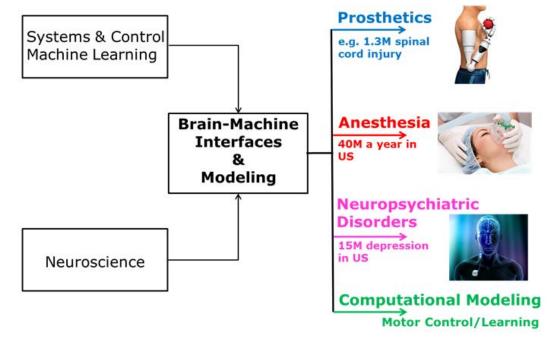


## Maryam Shanechi / USC



USC Viterbi

School of Engineering
Electrical Engineering
Department



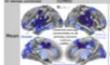


## Hava Siegelman / UMass- Amherst





- 20 year fMRI DB to brain motif: heirarchical abstraction
- Computing beyond the Turing limit via plastic RNN's



- Mini-circuitry: network theory u/ constraints
- Modeling memory Reconsolidation + applications in HCI HRI
- Memory-aid Technology: multi-tasking control



## Ambuj Singh / UC Santa Barbara





- Network-based analysis of fMRI and DTI/DSI
- Discovery of significant fragments for learning/disease
- Noise/robustness

- Spatial network processes
- Integration of heterogeneous datasets (genetic, DTI/DSI, fMRI..)
- Population-based adaptation & diversity



#### **James Smith**





- High performance computer architecture and organization
- Spiking Neural Networks

Temporal communication and computation

Biological plausibility, including training

Application to machine learning



#### Sara Solla / Northwestern





Background: theoretical physicist, machine learning research at Bell Labs, theoretical neuroscience research at Northwestern.

Research interests: neural control of movement, encoding of sensory input, Bayesian decision making, dimensionality reduction, dynamics of large and noisy systems, learning and adaptation.

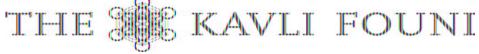
- The neural code and its use in the complex computations performed by the brain can inspire novel paradigms of biomimetic computation.
- Machine learning tools excel at feature extraction and classification. Does the brain implement similarly generic as opposed to task specific statistical strategies?



### Sharif Taha / Kavli Foundation



Computer science and brain science will benefit from one another, but only by actively listening the needs of both communities.





## Bertrand Thirion / Neurospin



INVENTEURS DU MONDE NUMÉRIQUE

- fMRI
- Machine learning,
- Open-source software,
- encoding and decoding,
- functional connectivity,
- Individual Brain Charting



### Francisco Valero-Cuevas / USC



USC Viterbi
School of Engineering

What physics-based problems does the brain really face and solve when controlling the body?

How do we produce versatile function with physiological sensors and actuators that are noisy, delayed and nonlinear?

These are critical questions we need to solve so that, when we interface with the nervous system, we can have a principled and effective approach to BMI and neurorehabilitation.

I address these questions by combining mechanics and neuroscience.





#### Les Valiant / Harvard

Computational models and theories of cortex, and their validation.

**Computational Complexity** 

Machine Learning

**Evolution** 



## Helen Vasaly / CCC



Bringing two separate communities together





## Santosh Vempala / Georgia Tech





High-dimensional Algorithms, Randomness, Optimization, Foundations of ML

What is a (mathematically) plausible explanation of the cortex's ability to learn? Not an answer by far, but <a href="https://www.nee.google.com/here">here</a> is something...



## Ragini Verma / UPenn



Structural and Functional Connectomics

&

All things Diffusion Imaging



## Joshua Vogelstein / JHU





- Data Intensive Brain Sciences
- Big Graph Statistics
- Wide Data Computational Statistics
- Democratizing Brain Data & Methods



#### Ross Whitaker / Utah





- Medical image analysis
- Scientific computing
- Analysis/visualization high dimensional data and uncertainty



#### Rebecca Willett / Wisconsin





Machine learning provides fundamental insight into how salient information is represented and used to make predictions and decisions.

Brain science provides a working model of how small, local decisions can be lead to complex but robust behaviors.

Understanding each of these can significantly further the state-of-the-art in the other.

