

# CS 5043: Advanced Machine Learning

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# What is Machine Learning?

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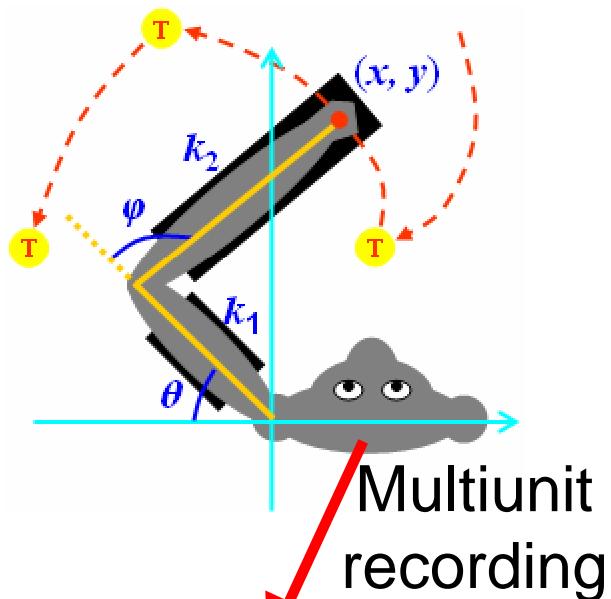
- Fundamentally: using data to automatically construct a model
- The model must be predictive!
  - I.E.: to be useful, it must produce meaningful output given novel situations.

# Brain-Machine Interfaces

Estimate of intended movement

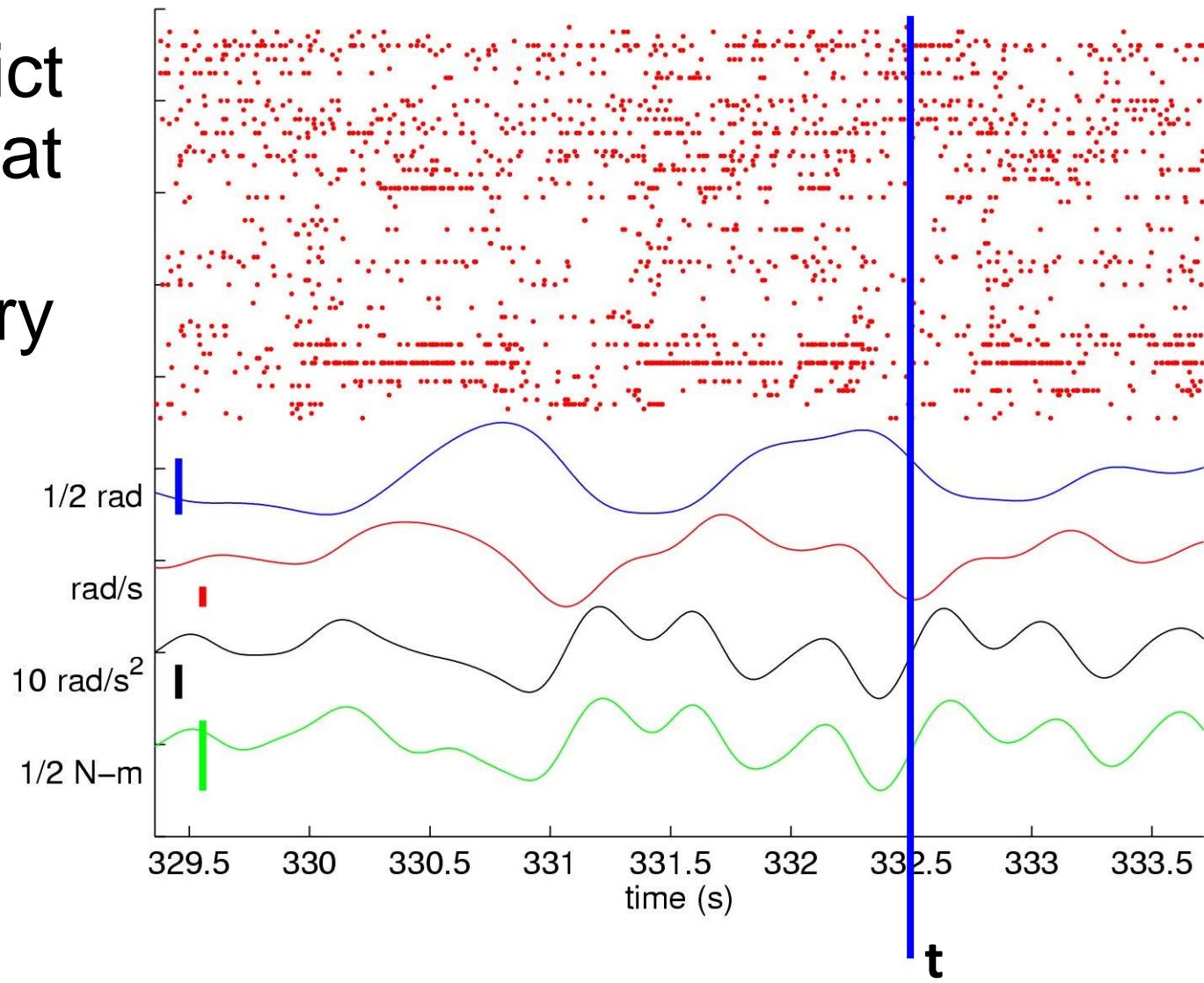
Predictive model

Command  
prosthetic arm



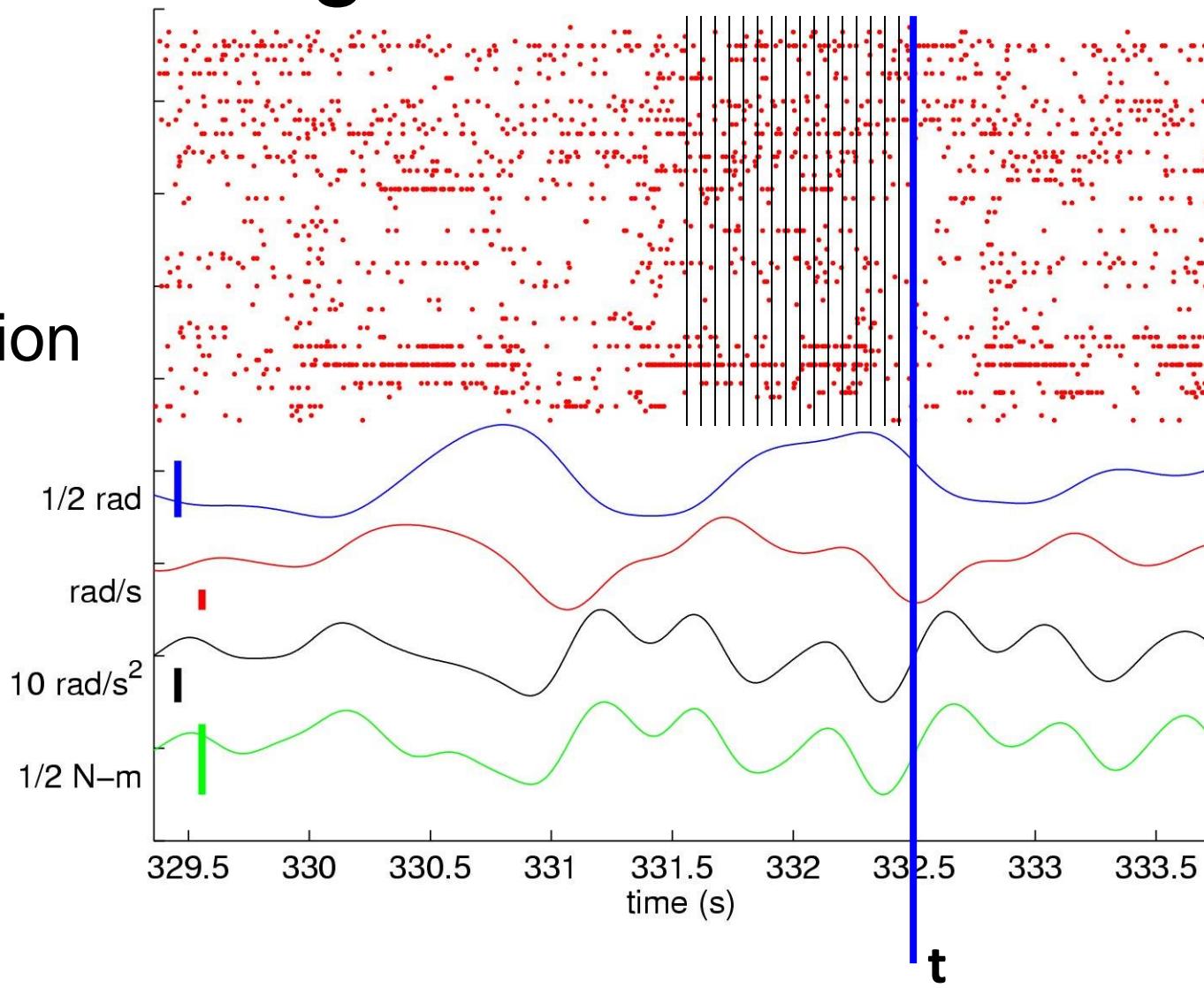
# Decoding Arm State

Want to predict arm motion at time  $t$  given recent history of spiking behavior



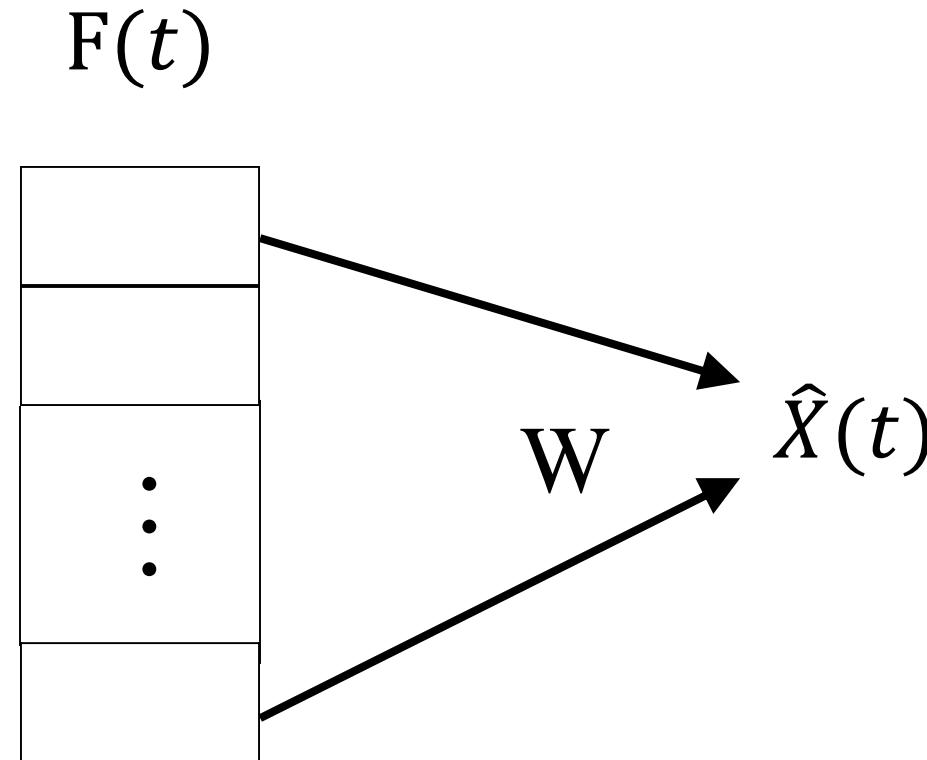
# Decoding Arm State

50ms bins: 20  
descriptors of  
neural activation  
for each cell



# Wiener Filter

Each feature ( $F_i$ ) is a count of spikes by a neuron for a 50 ms bin



$$\hat{X} = g_W(F(t)) = W^T F(t)$$

Column vector encoding spike counts for  $N$  cells at  $T$  taps up to time  $t$

# Data Types

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- Continuous
  - Probabilities
- Categorical (enumerated)
  - Binary
- Structured
  - Bind together arbitrary primitive data types
    - A class in an Object-Oriented sense does this
    - Heterogeneous objects are possible, too
  - Vectors, matrices

# Data Types

- In some cases, we want to be able to acknowledge the fact that some data values are unknown or uncertain

# Classes of Models

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Defined by the data type of the output (and possibly the input). Very broadly:

- Categorical output: classifier models
- Continuous output: regression-type models

# Classes of Machine Learning Problems

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## Supervised learning

- Training set contains only input / output (labels) pairs
- Outputs could be continuous, probabilistic or categorical

# Classes of Machine Learning Problems

## Semi-Supervised learning

- Part of the training set contains input / output pairs
- The rest of the training set contains only inputs
- Using all of the data can yield a better model than if we only used the labeled data

# Classes of Machine Learning Problems

## Unsupervised learning

- The training set contains only inputs
- Fundamental question: what is the structure of these inputs?
  - Most common case: algorithm assigns categorical labels to each sample
  - But we can also ask continuous questions. For example: how different are two samples?

# Classes of Machine Learning Problems

## Reinforcement learning

- The training set contains inputs and an evaluation (reinforcement) of the output that is generated
- Most common case: a single evaluation can be a function of the sequence of outputs that is generated
  - How much time did it take to solve a problem?
  - How much energy did you use while solving the problem?
- Learning algorithm: for a given input, what is the output that maximizes the reinforcement over time?

# Our Topics

- Decision trees: ensemble methods, random forests, boosting
- Combatting overfitting with advanced regression: ridge regression, Tikhonov regression, lasso, elastic nets
- Dimensionality reduction: Kernel PCA, local linear embedding, ISOmap, multidimensional scaling
- Semi-supervised learning

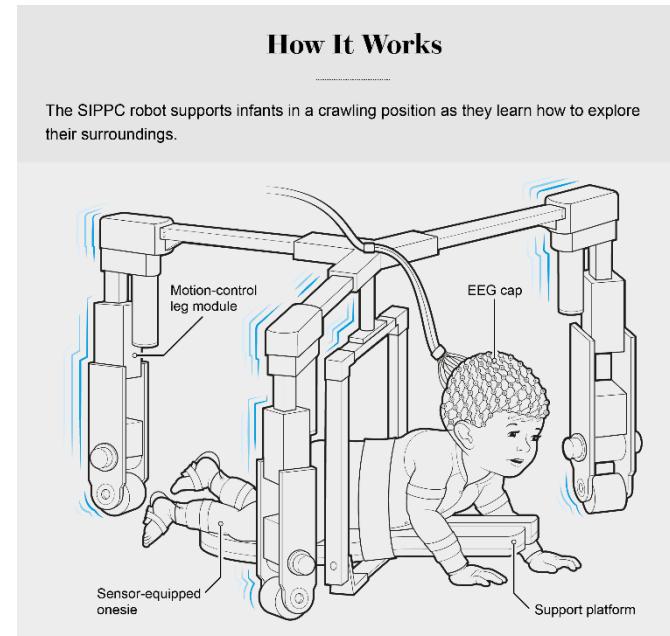
# Our Topics

- Deep Learning
  - Dropout and other normalization techniques
  - Convolutional neural networks
  - Recurrent neural networks
  - Autoencoders
- Evaluation in ML: metrics, cross-validation, statistics, addressing the multiple comparisons problem
- Reinforcement learning

# Real-Time Activity Recognition for Assistive Robotics



OU Crawling Assistant  
(Kolobe, Fagg, Miller, Ding)

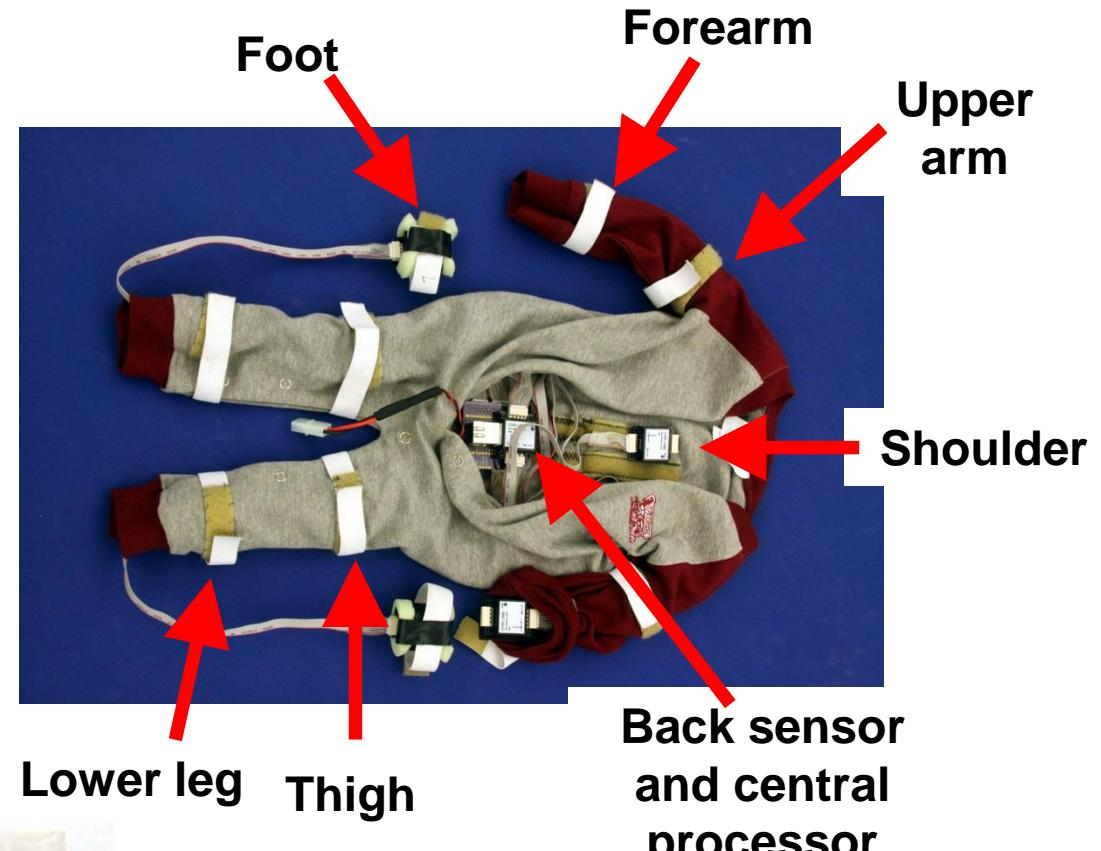


Scientific American (Oct 2016)

# Kinematic Capture Suit

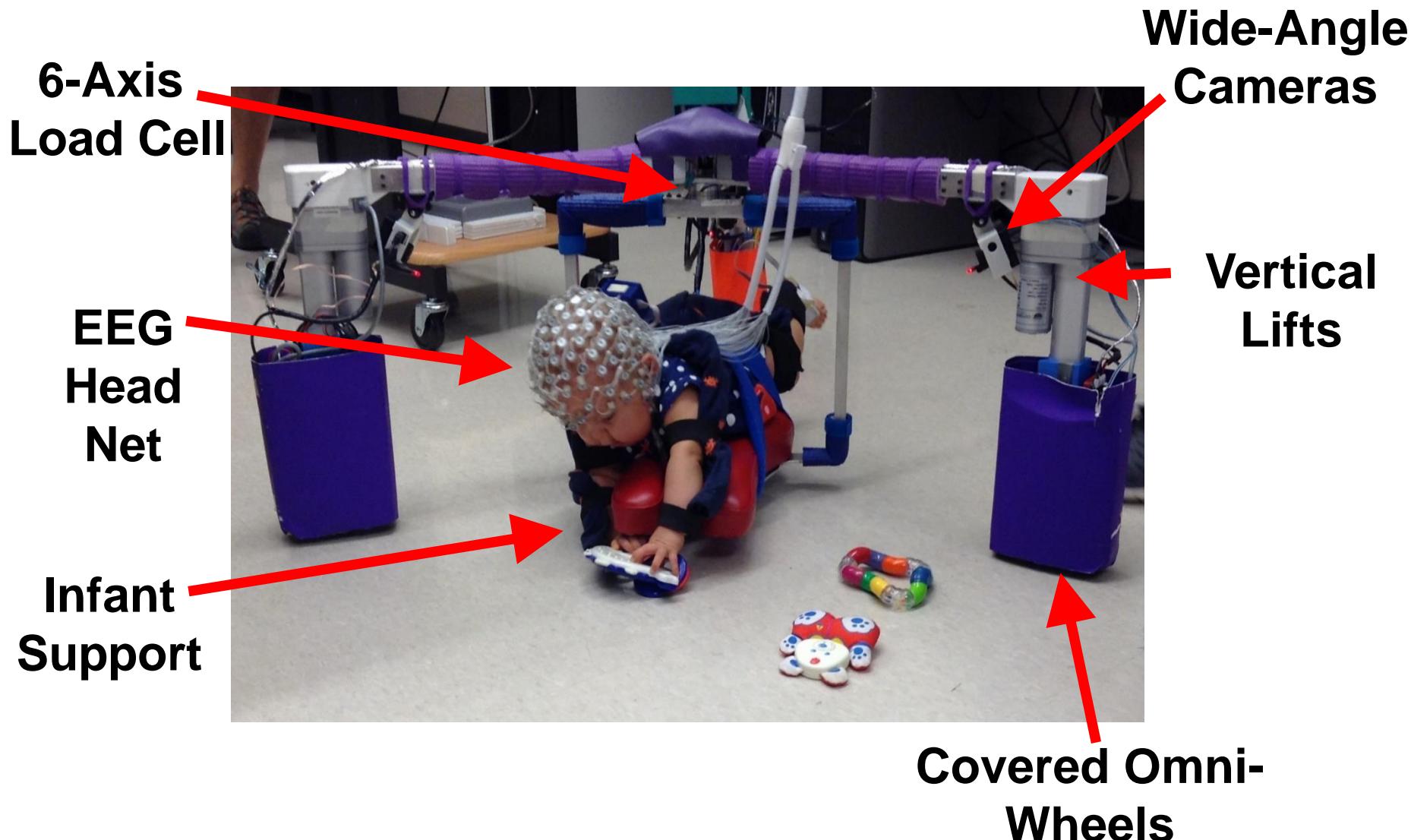
IMU-based kinematic suit

- 12 sensors mounted in suit
- Real-time reconstruction of body posture
- Recognition of crawling-like actions



Southerland (2012)

# SIPPC Crawling Assistant



# Infant-Robot Interaction

Three modes of interaction:

- **Force control**: robot velocity is linearly related to ground reaction forces
- **Power steering**: small ground reaction forces produce a substantial robot movement
- **Gesture-based control**: recognized crawling-like movements produce robot movement

# Machine Learning Questions

- Predict robot motion from kinematic data
- Predict visual attention from kinematic and robot data
- Predict limb motion from EEG data
- Predict visual attention from EEG data
- ...

# Other Challenges

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- Noisy data
- Unknown / invalid data values
- Data does not provide a complete “view” of the world
- Shift in the statistics of the data over time (non-stationarity)
- Small data sets
- Measuring model performance

# To “Solve” a Model Building Problem We Must Answer:

- What is the nature of the data that we have?
- How much data do we have?
- What is the prediction problem?
- How do we measure performance of a model?
- How to select an appropriate model and learning algorithm?
- How to choose parameters?
- How to convince ourselves (and others) that we have a useful model?

# What I am assuming about you...

- Programming skills
- Able to jump into Python, including the “Object-Orientedness” of it
- Know or can learn unix command-line tools

# Resources

- Course web page:  
<http://www.cs.ou.edu/~fagg/classes/mlfds>
- Text: Aurélien Géron (2017) *Hands-On Machine Learning with Scikit-Learn and TensorFlow (Concepts, Tools, and Techniques to Build Intelligent Systems)*, O'Reilly Media
- May also be useful: Nikhil Buduma (2017) *Fundamentals of Deep Learning*, O'Reilly Media
- Web resources: documentation, tutorials, papers (linked from the schedule or announced on Canvas)

# Computing Environment

Setting up a ML environment (especially one based on TensorFlow) can be a bear ...

- We are providing a pre-configured compute cluster on Amazon Web Services (AWS)
- Key tools: Python, Scikit-learn, TensorFlow (Deep Learning), Jupyter (Interactive Development Environment)
- Other software: editors (emacs, vi, gedit)
- Will also house our common data sets

# Computing Environment

AWS machines are nice, but cost us based on the resources we use

- Right now, we have one AWS machine configured (1 processor; 2 GB of memory; 32 GB of swap).
- This machine will increase in resources as needed
- Additional machines will be added as needed
  - They will share a common user accounting system and file system
  - We will experiment with different machine configurations (some GPUs and AWS “Spot Instances”)

# Configure your Cluster Account

- Install SSH on your local machine
- Generate a public key; email this and your desired username to me
- Once your account is created: SSH to the machine and configure Jupyter
  - The Jupyter server runs on the cluster
  - The front end runs in the browser of your local machine
  - A ***SSH tunnel*** allows the two to communicate securely

# Homework Assignments

- First half of the semester
- Explore different ML methods and data sets

# Projects

- Last half of the semester
- Topic / data set are your choice, but must be approved
- Several in-class presentations
- Final paper

# Grading

- In-class participation: 10%
- Homework: 40%
- Project work: 50%

# Proper Academic Conduct

- Homework assignments are to be done on your own
  - No communication of solutions in any form
- Projects: I am still deciding whether these will be group or individual projects

# For Next Time

- Chapters 1 and 2 from the book
- We will discuss Pandas and start in on Scikit-learn



# Advanced Machine Learning for Data Science

## Software Tools

- Python
- Scikit-Learn
- XG-Boost
- TensorFlow

## Reading

- Hands-On Machine Learning with Scikit-Learn & TensorFlow (O'Reilly)
- XG-Boost
- Online tutorials