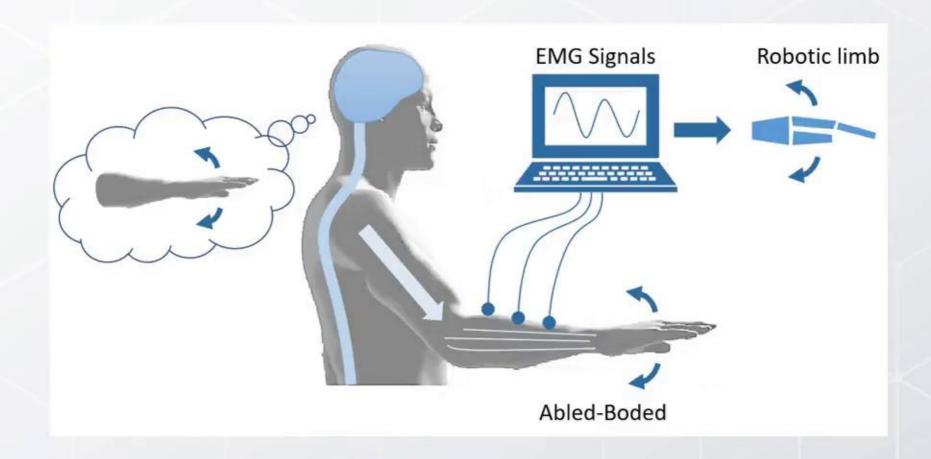
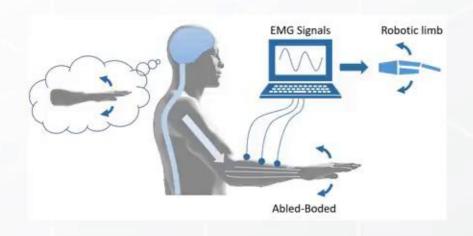
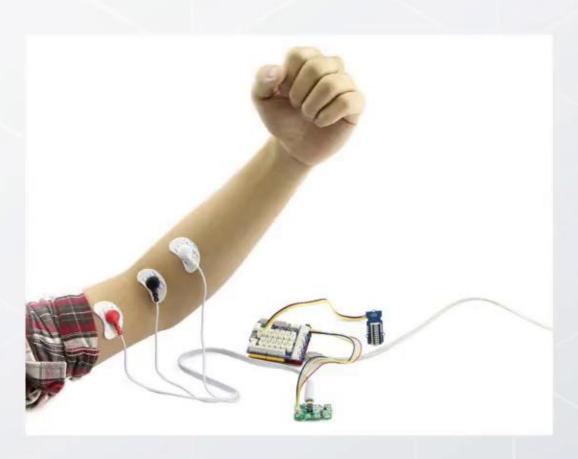
Feature Selection in Machine Learning

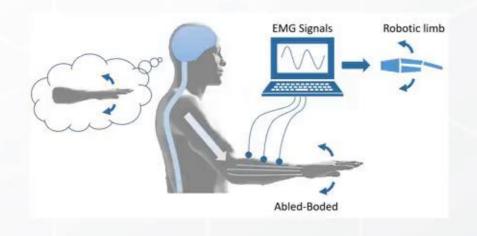
Guohui Song
Old Dominion University

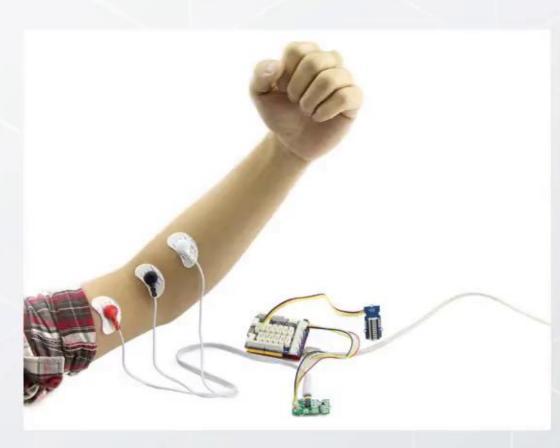


A controller converts EMG signals into hand movement.





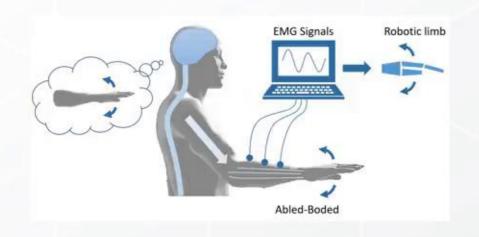


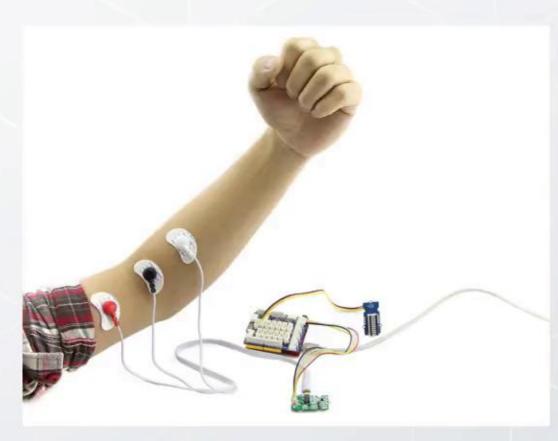




















How to choose the sensor locations?

Disease Discoveries Using Biomarkers

• Identifying biomarkers that are indicative of a specific disease, such as breast cancer or Alzheimer's disease.

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You might have data on the expression levels of 20,000 genes for each patient sample.



- filter methods: use statistical tests to measure the relation between each feature and the output.
 - Pearson correlation, mutual information, Chi-Square etc.
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- filter methods: use statistical tests to measure the relation between each feature and the output.
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 - Fast but not very accurate. They do not consider interactions among features.
- wrapper methods: search the space of all possible subsets of features.
 - o Forward selection, Backward elimination, Recursive feature elimination, Exhaustive feature selection.
 - Slow but more accurate. They consider interactions among features.

- embedded methods: embed the feature selection process into the model learning process.
 - Lasso, Ridge regression, Elastic net, Random forest, Gradient boosting.
 - o Combine the benefits of both the wrapper and filter methods. Include interactions of features but also maintain reasonable computational costs

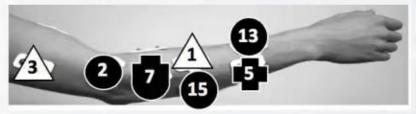
- embedded methods: embed the feature selection process into the model learning process.
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We assign a weight to each feature and find the best weights under the constraints of sparseness.

Multiscale LASSO Feature Selection on EMG Data

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Multiscale LASSO

Movement	Selected sensors	CV MSE	Time (m)
Finger	7,12	7.2E-2	47
Wrist	11,15	3.9E-2	70

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Multiscale LASSO

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It does not work well if the data is too few comparing to the number of features.



• When the data lacks of quality/quantity, we need to incorporate prior knowledge.

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 - o toss the coin 10 times and get 7 heads

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- Prior comes from the domain knowledge
- o Prior is in particular useful for limited data
- o Prior will be "corrected" by the data
- Do not eliminate the possibility unless absolutely impossible

Thank you for your attention!

Questions?

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