



Dr. Debdoot Sheet

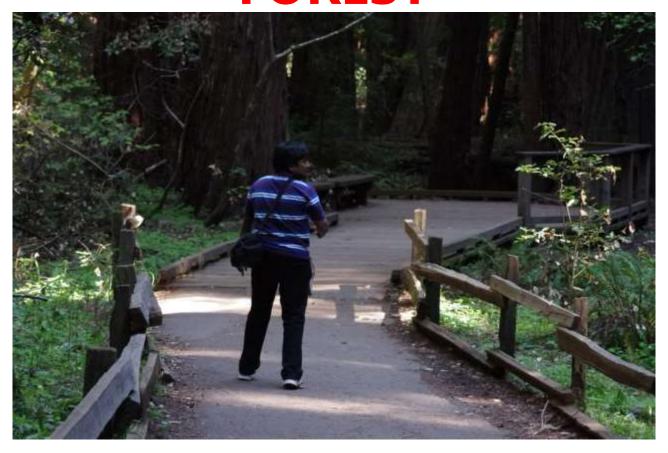
Assistant Professor
Department of Electrical Engineering
Indian Institute of Technology Kharagpur

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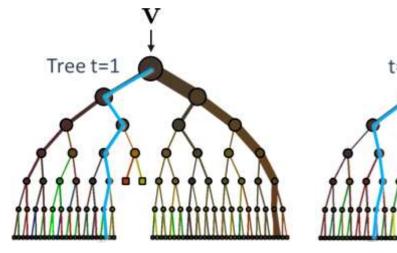
NOT ABOUT WALKING IN A FOREST

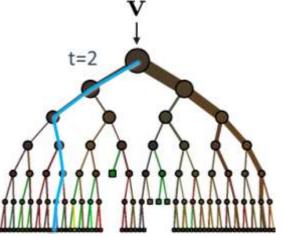


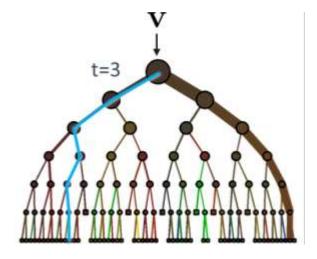
1 Mar 2015



IS ALL ABOUT









Overview

- Historical Perspective
- Decision Tree
- Random Forest
- Application Scenarios
- Computational Complexity
- Variable Importance
- What's hot about them in ML Research?



Historical Perspective

Decision Trees and Random Forests / Debdoot Sheet / MLCN2015

Decision Trees

- L. Breiman, J. Friedman, C. J.
 Stone, and R. A. Olshen,
 Classification and Regression
 Trees. Chapman and Hall/CRC
 (SIAM), 1984.
- J. R. Quinlan, *C4.5: Programs* for Machine Learning. **1993**.

Random Forests

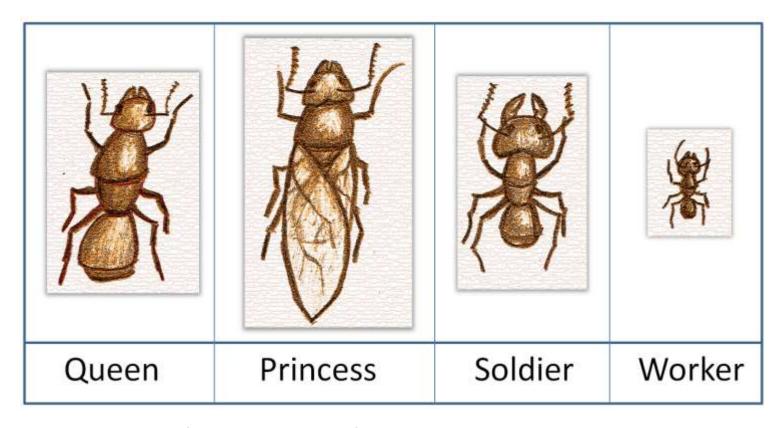
- Y. Amit and D. Geman., "Shape quantization and recognition with randomized trees," *Neural Computation*, vol. 9, pp. 1545–1588, **1997**.
- T. K. Ho, "The random subspace method for constructing decision forests," *IEEE T-PAMI*, vol. 20, no. 8, pp. 832–844, **1998**.
- L. Breiman, "Random forests," *Machine Learning*, vol. 45, no. 1, pp. 5–32, 2001.



DECISION TREE



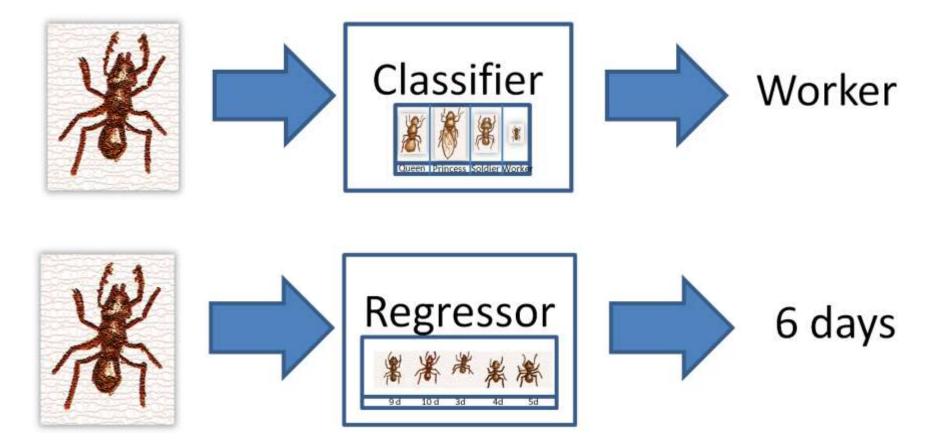
Problem Statement



Formica rufa (Red wood ant)

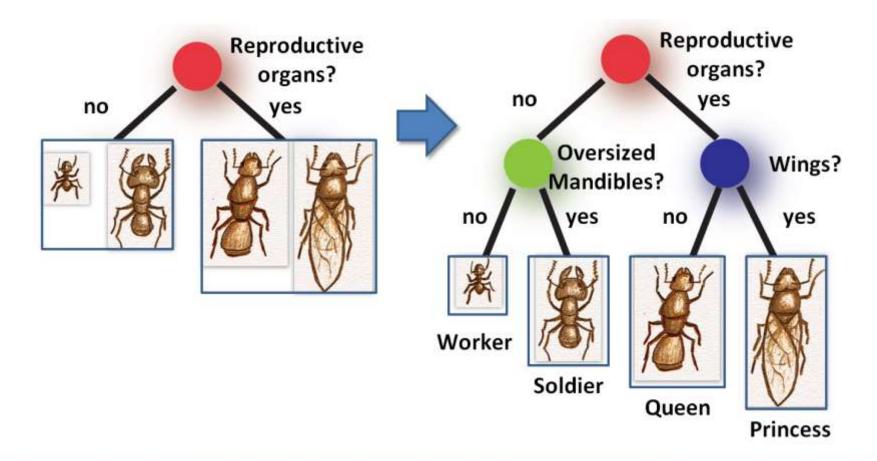


Classification vs. Regression





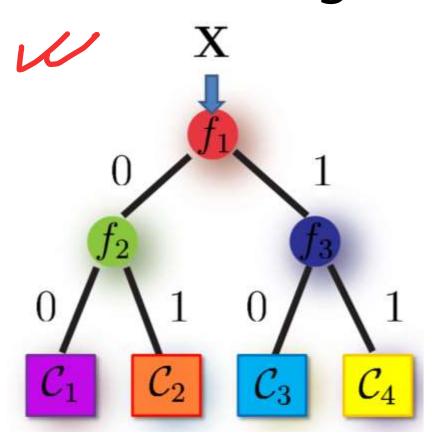
Decision Tree

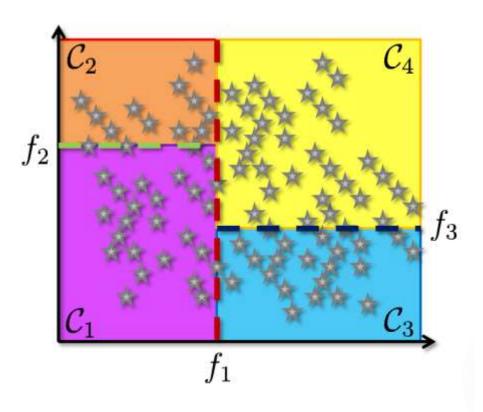






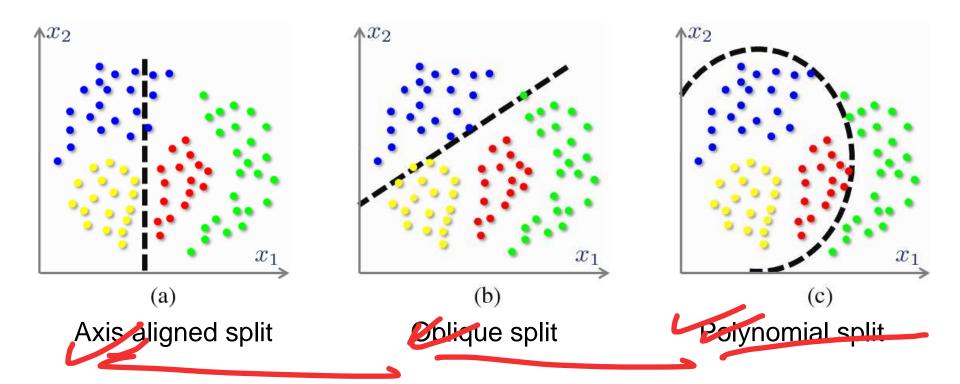
Forming a Decision Tree





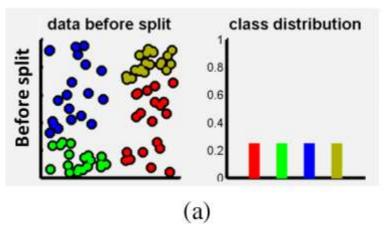


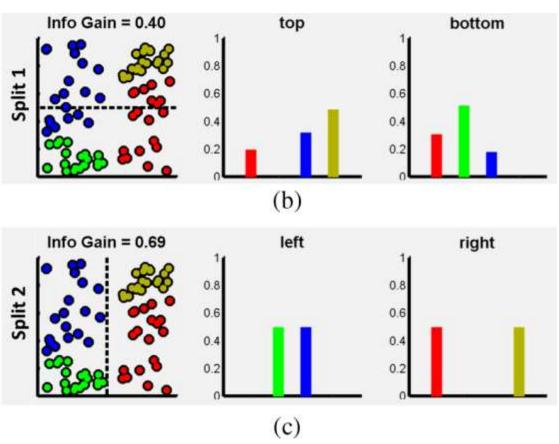
Step 1: Split Function at Node





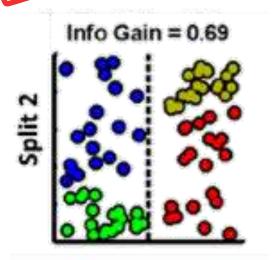
Step 2: Assessing Purity of Split







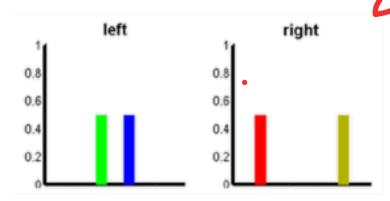
Sost function for Split Purity



1 hr

Entropy of class distribution

$$H(S) = -\sum_{c \in C} p(c) \log(p(c))$$

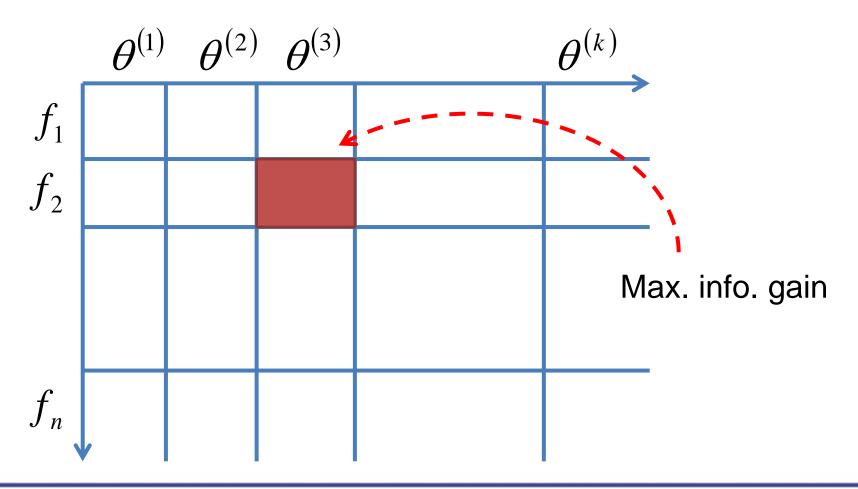


Information Gain

$$I = H(S) - \sum_{i \in \{L,R\}} \frac{|S^i|}{|S|} H(S^i)$$

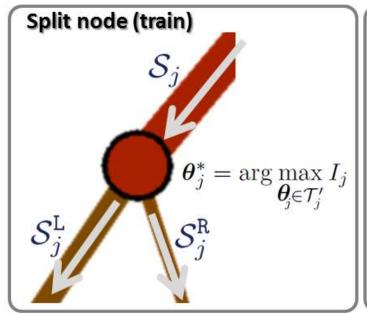


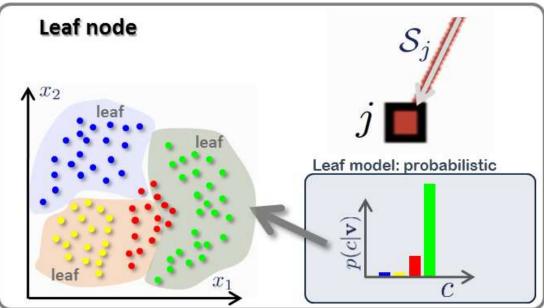
Step 3: Selecting Optimum Split





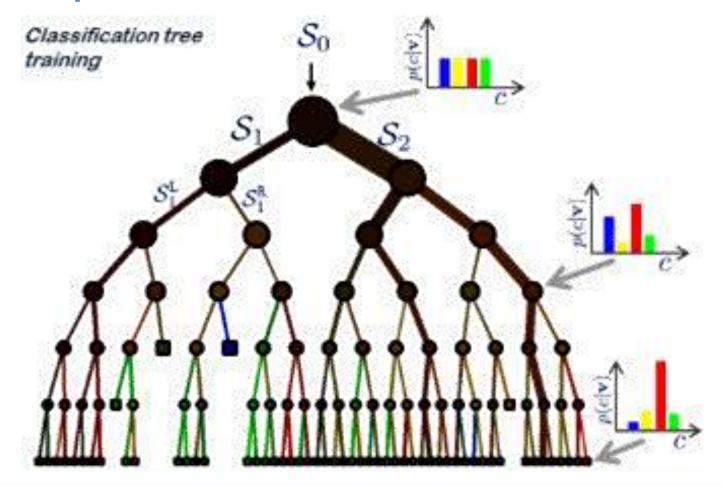
Step 4: Stopping Criteria





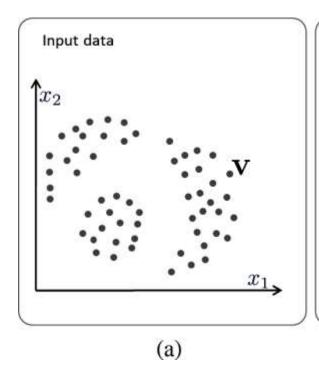


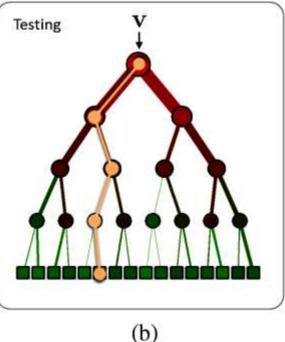
Step 5: Leaf Prediction Model

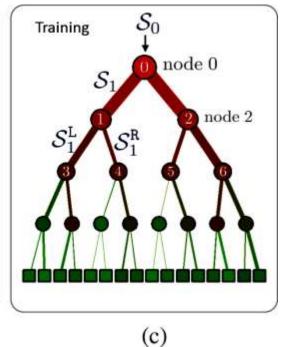




Deploying a Decision Tree





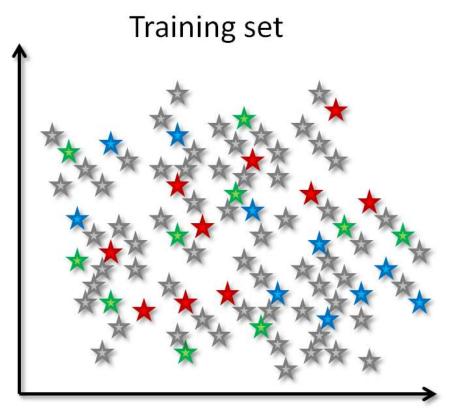




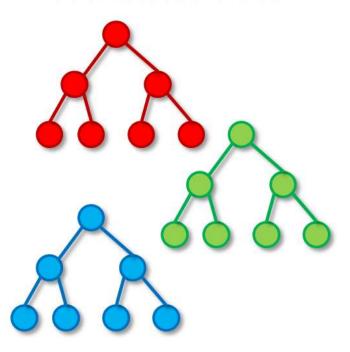
RANDOM FOREST



Growing Multiple Trees in a Forest



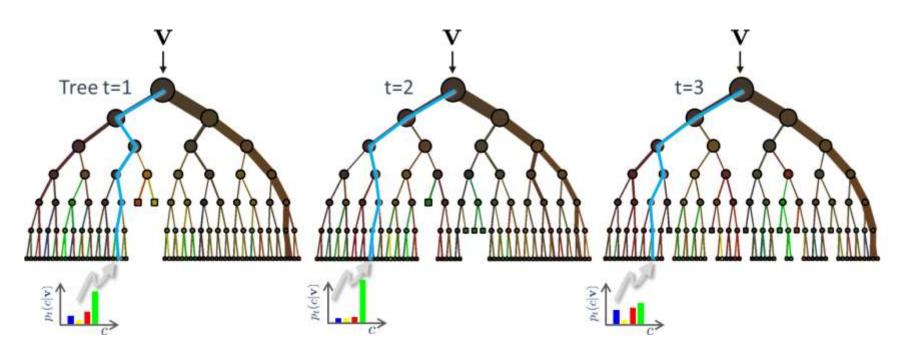
Decorrelated trees



Bagging – Bootstrapped Aggregation



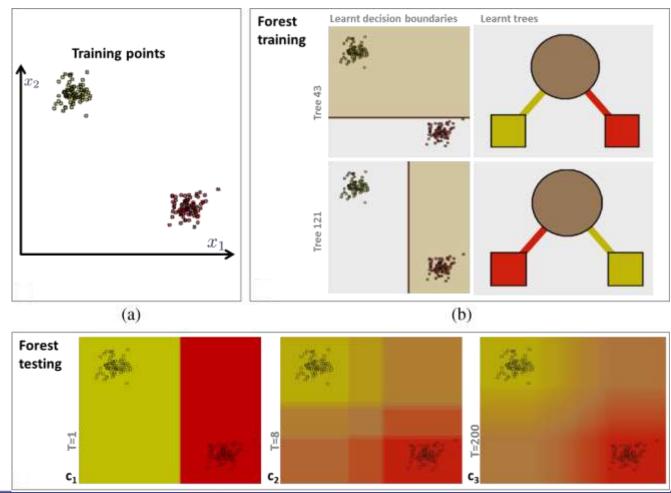
Ensemble Prediction Model



$$p(c|\mathbf{v}) = \frac{1}{T} \sum_{t}^{T} p_t(c|\mathbf{v})$$

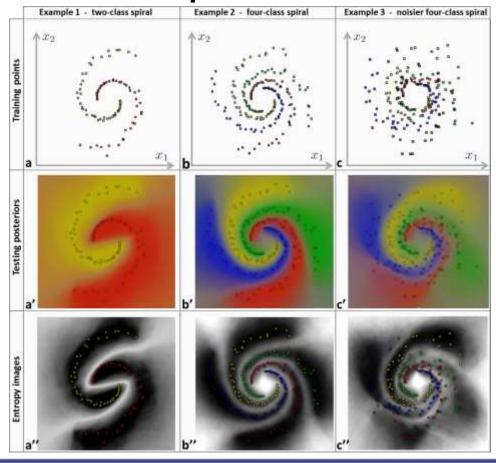


What do we gain by using a Forest?



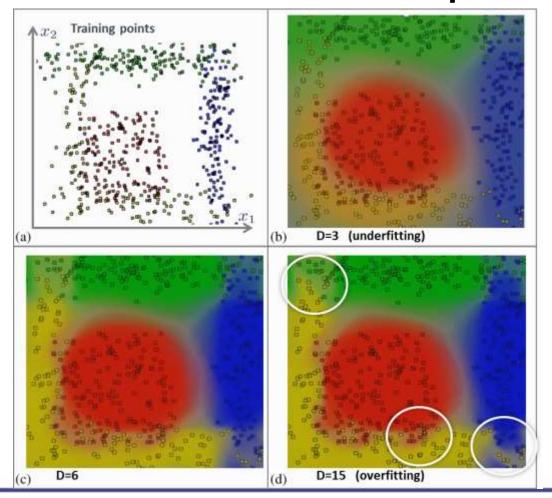


Noise Resilience and Topology Independence



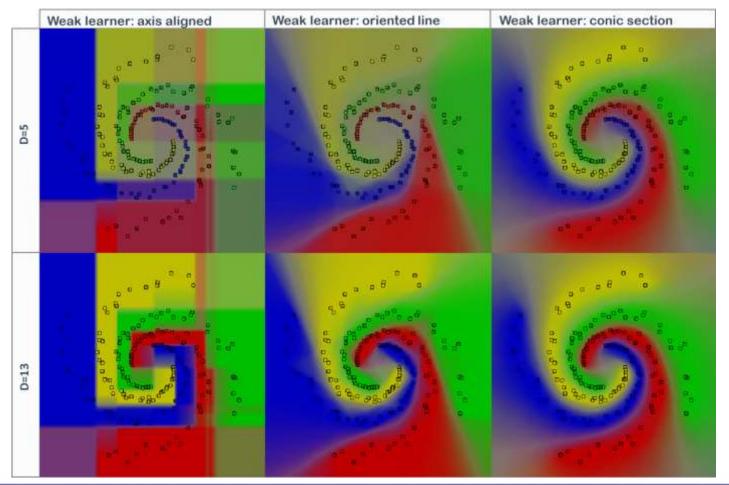


Effect of Tree Depth



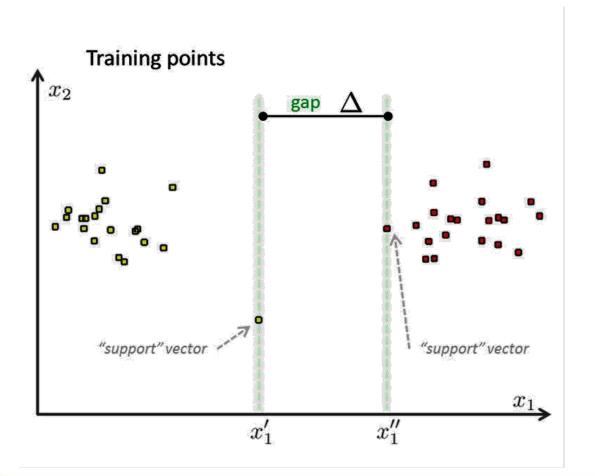


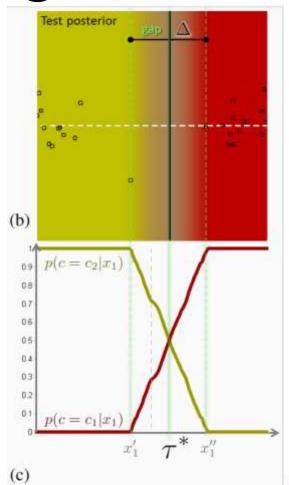
Effect of Split Function





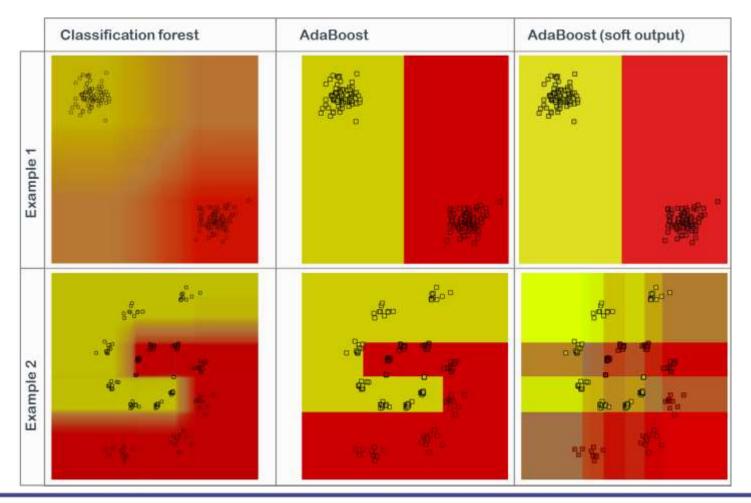
Classification Margin





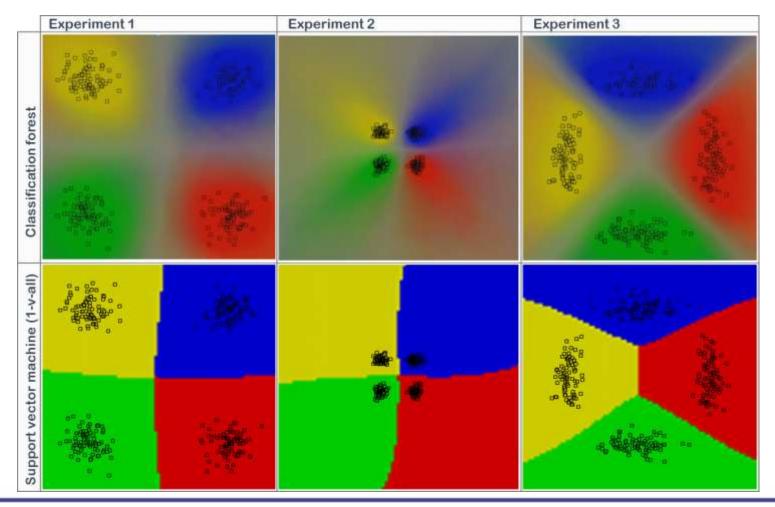


Random Forest vs. AdaBoost



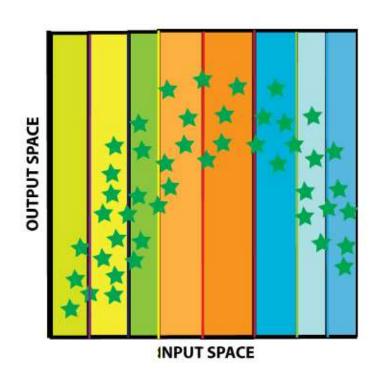


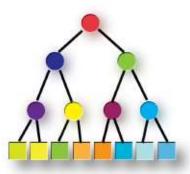
Random Forest vs. SVM

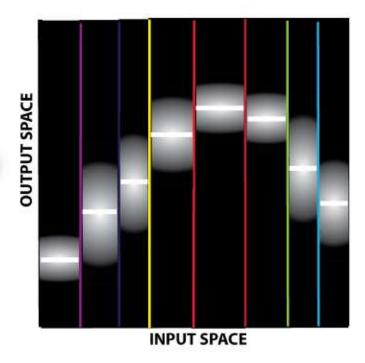




Regression Forest

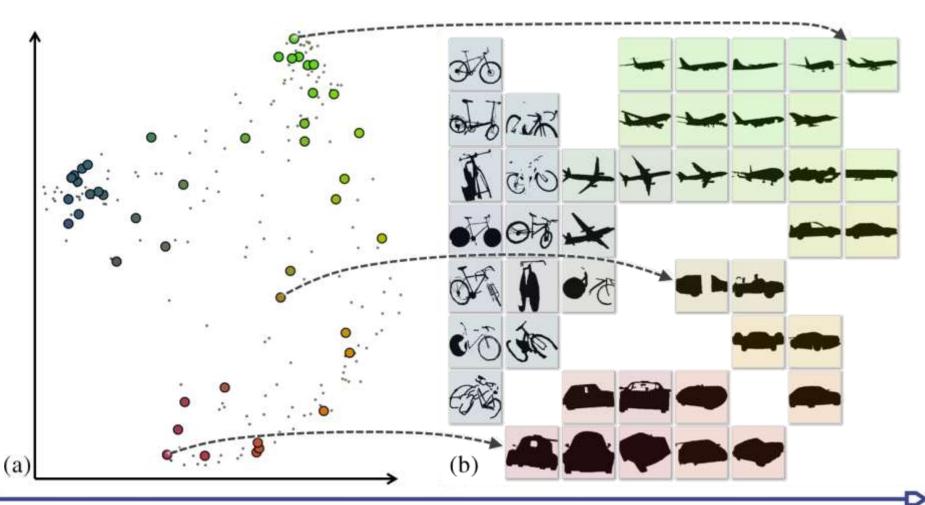








Manifold Forest



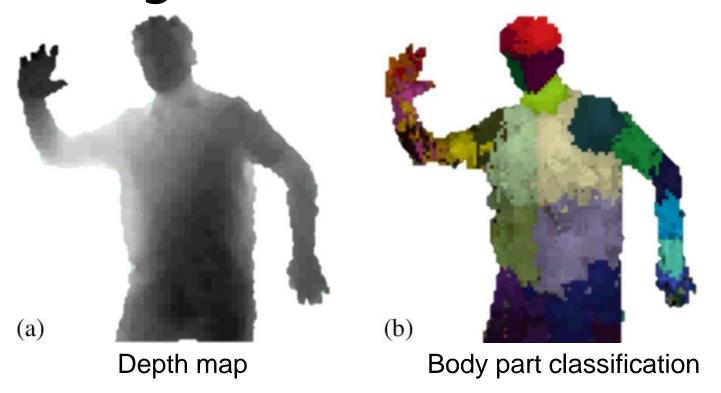


After the Brainstorming (Break)!

APPLICATION SCENARIOS



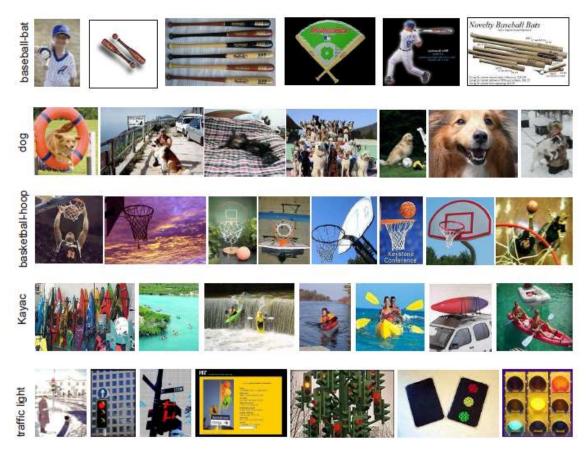
Gaming – Kinect for Xbox 360



J. Shotton, A. Fitzgibbon, M. Cook, T. Sharp, M. Finocchio, R. Moore, A. Kipman, and A. Blake, "Real-time human pose recognition in parts from a single depth image," in *Proc. CVPR*, 2011.



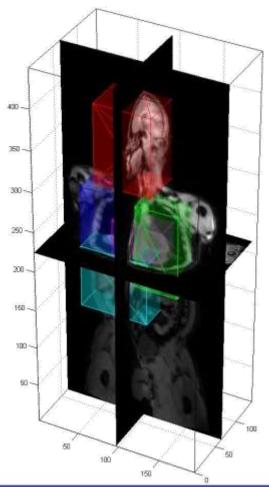
Vision – Scene Classification

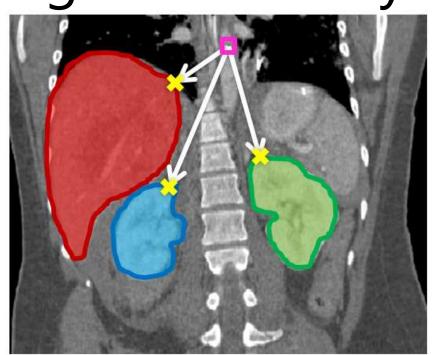


Bosch, A., Zisserman, A., & Muoz, X. "Image classification using random forests and ferns", ICCV 2007.



Medical – Digital Anatomy

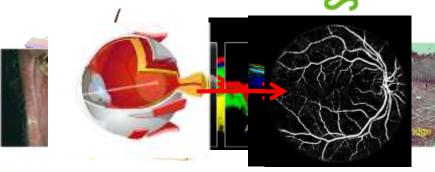




A. Criminisi, D. Robertson, E. Konukoglu, J. Shotton, S. Pathak, S. White, and K. Siddiqui, "Regression Forests for Efficient Anatomy Detection and Localization in Computed Tomography Scans", *Medical Image Analysis*, 2013



Medical – **Computational Histology**







D. Sheett cetal, "Dietritul control reportal bresoelis ishation kisha astrigin in ing astriging a like in ing a like in ing a like in ing a like ing a lik estricins execution of the political extension of ce tathisticadiliptivitisi cpt approximation to the control of the co Bolotrassocium dia dillege (1180 Belgie 20 na), 18(1), 2014

D. Sheet, et al., "Transfer Learning of Tissue Photon Interaction in Optical Coherence Tomography towards In vivo HI BOHOR SOFTHE WITH SOFTH PROPERTY OF THE S Cuntrast Imaging for Label-free Retinal Angiography", Int. Symp. Biomed. Imaging (ISBI), 2014

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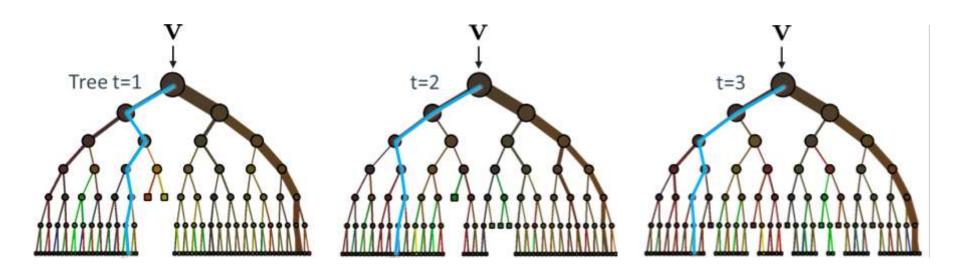
IFRAMET G 595 BV); 2012.



ENGINEERING DESIGN PERSPECTIVE



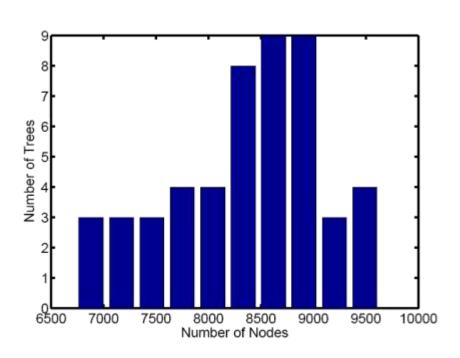
Understanding Computations



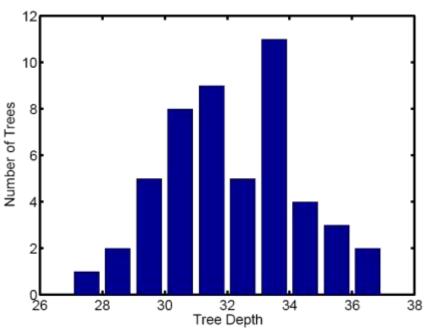


Computational Complexity

Training Complexity

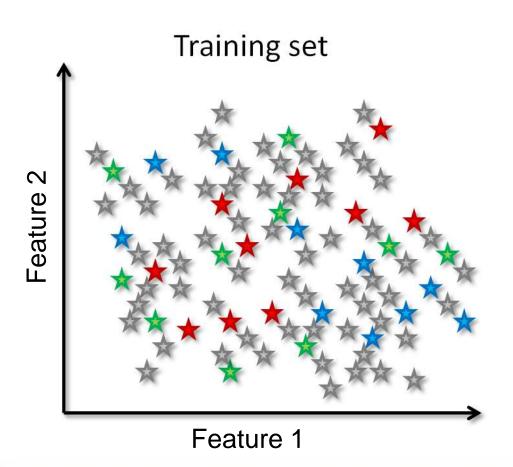


Testing Complexity

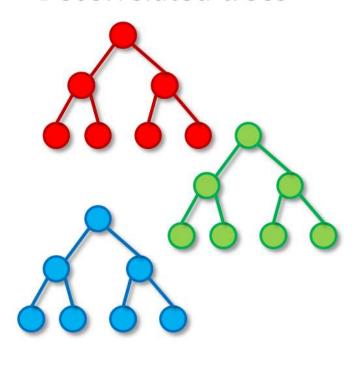




Features and their Role

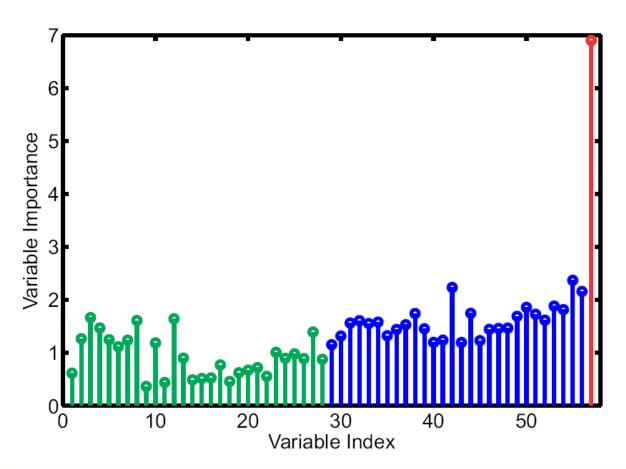


Decorrelated trees





Variable Importance



Genuer, R., Poggi, J.-M., Tuleau-Malot, C., (2010). Variable selection using random forests. *Pat. Recog. Letters.* **31**(14):2225-2236



WHAT'S HOT IN RESEARCH?



Research Challenges in 2015

- Architecture
 - Online learning
 - Incremental learning
 - Long term memory
 - Parallel distributed architectures
 - Split functions, cost functions, stopping criteria
 - Domain adaptation

- Engineering and Application
 - Computational complexity
- Statistics and Science
 - Consistency of forests
 - VC dimension
 - De-correlated trees





Take Home Message

Reading

- L. Breiman, J. Friedman, C. J.
 Stone, and R. A. Olshen,
 Classification and Regression
 Trees. Chapman and
 Hall/CRC, 1984.
- L. Breiman, "Random forests," *Machine Learning*, vol. 45, no. 1, pp. 5–32, 2001.
- A. Criminisi and J. Shotton, Decision Forests for Computer Vision and Medical Image Analysis, Springer, 2013.

Toolboxes and Packages

- randomForest in R
- TreeBagger in Matlab
- sklearn.ensemble.RandomFo restClassifier in Python-Scikit-learn

Conferences

- Int. Conf. Comp. Vis. (ICCV)
- Eur. Conf. Comp. Vis. (ECCV)
- Asian Conf. Comp. Vis. (ACCV)
- Comp. Vis. Patt. Recog. (CVPR)
- Med. Image Comp., Comp. Assist. Interv. (MICCAI)

