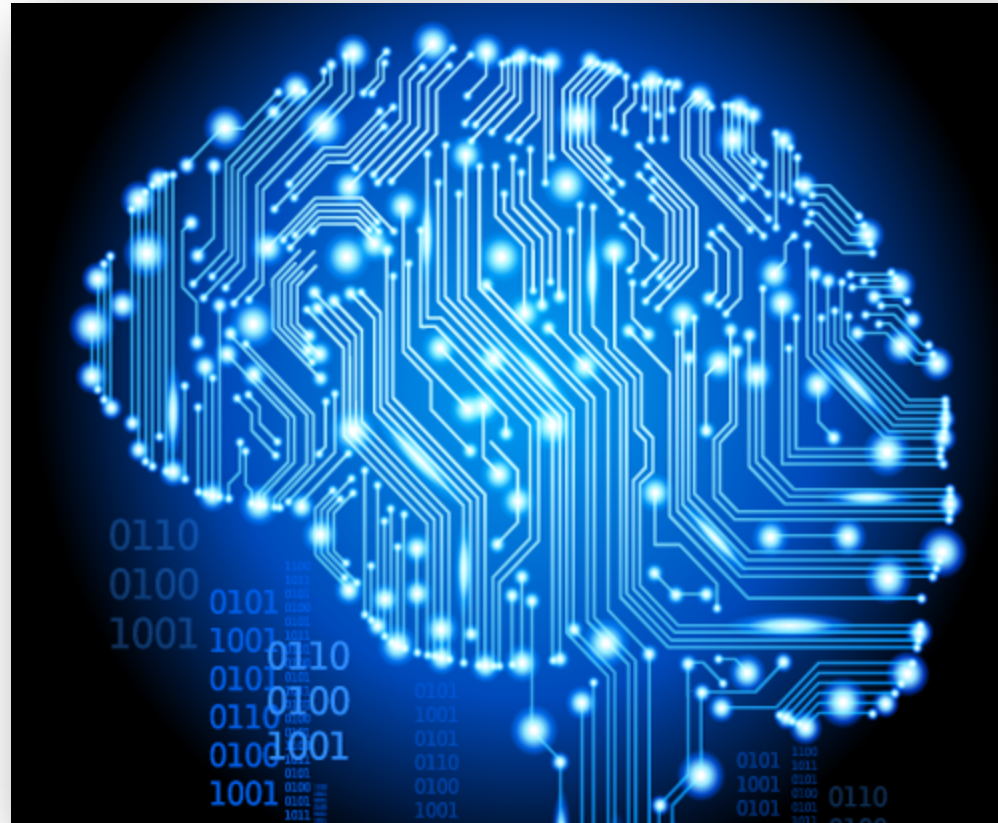


# Session IV: Deep Learning/CNN Methods



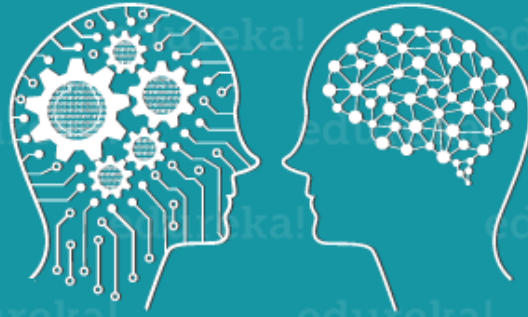
# ARTIFICIAL INTELLIGENCE

Engineering of making Intelligent  
Machines and Programs



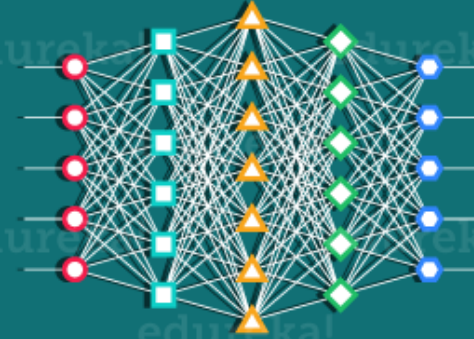
# MACHINE LEARNING

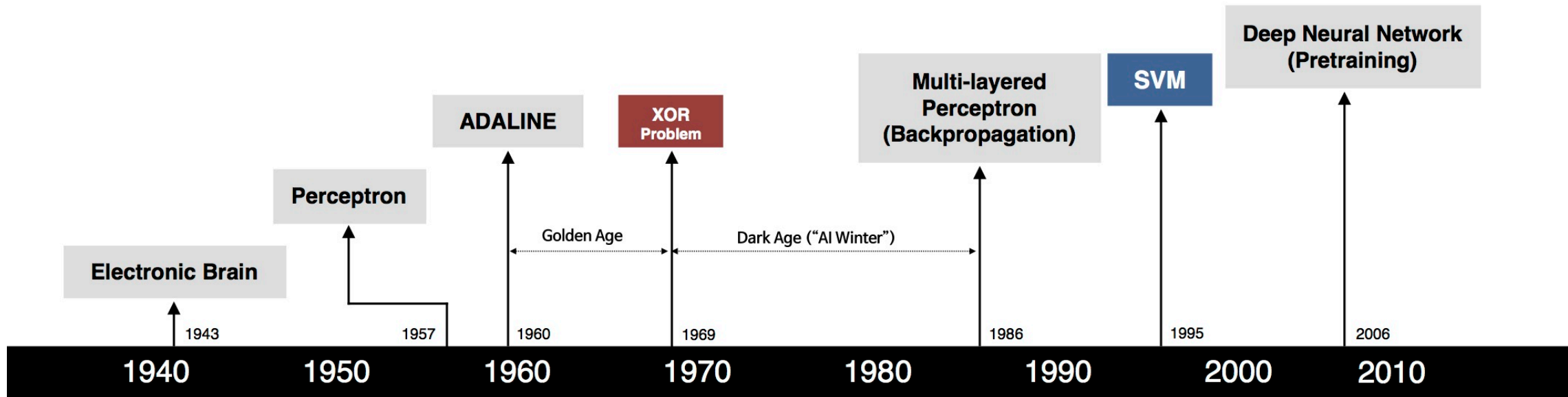
Ability to learn without being  
explicitly programmed



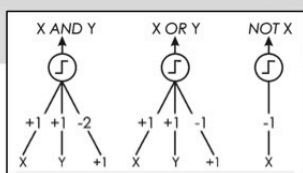
# DEEP LEARNING

Learning based on Deep  
Neural Network





S. McCulloch – W. Pitts



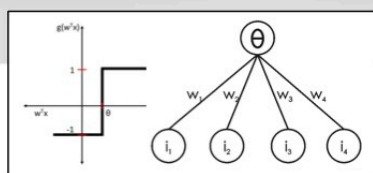
- Adjustable Weights
- Weights are not Learned



F. Rosenblatt



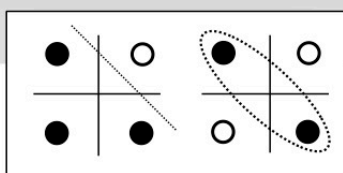
B. Widrow – M. Hoff



- Learnable Weights and Threshold



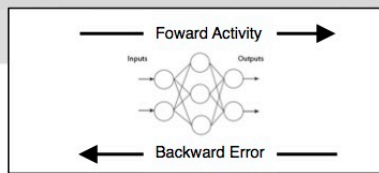
M. Minsky – S. Papert



- XOR Problem



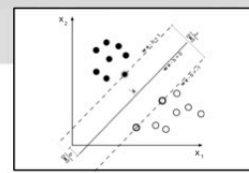
D. Rumelhart – G. Hinton – R. Williams



- Solution to nonlinearly separable problems
- Big computation, local optima and overfitting



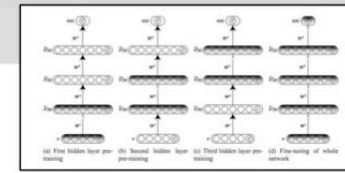
V. Vapnik – C. Cortes



- Limitations of learning prior knowledge
- Kernel function: Human Intervention

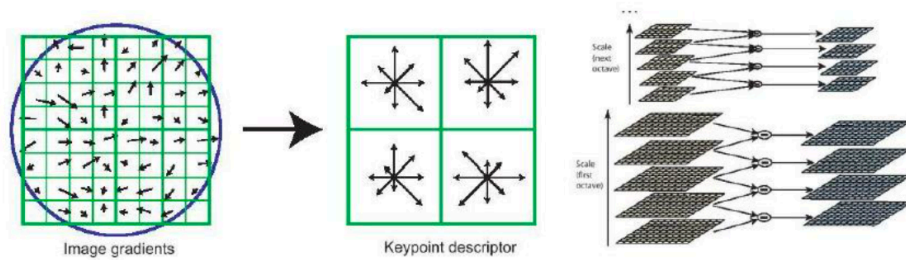


G. Hinton – S. Ruslan

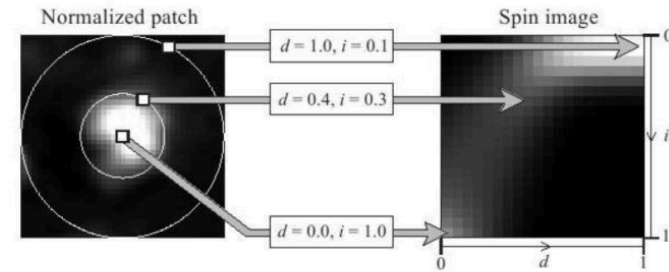


- Hierarchical feature Learning

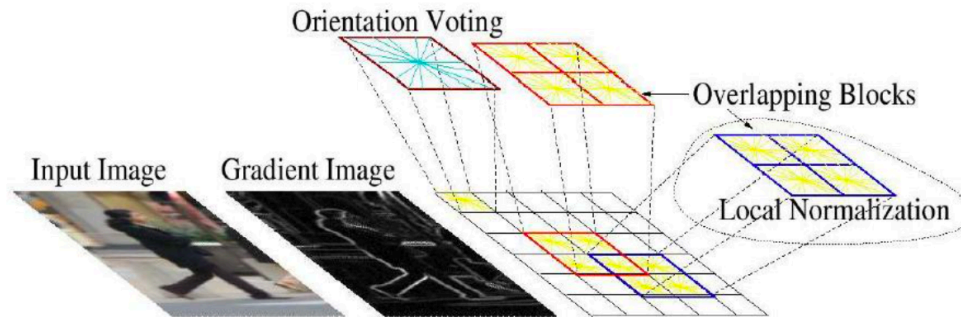
# Traditional Machine Learning Approach



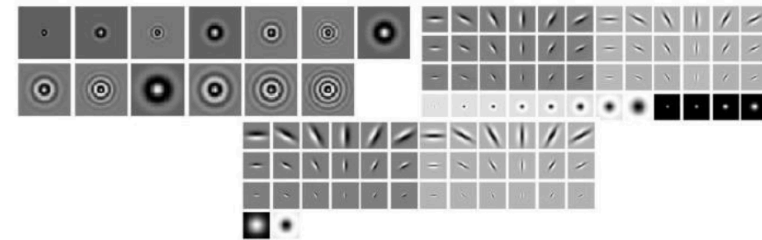
SIFT



Spin image

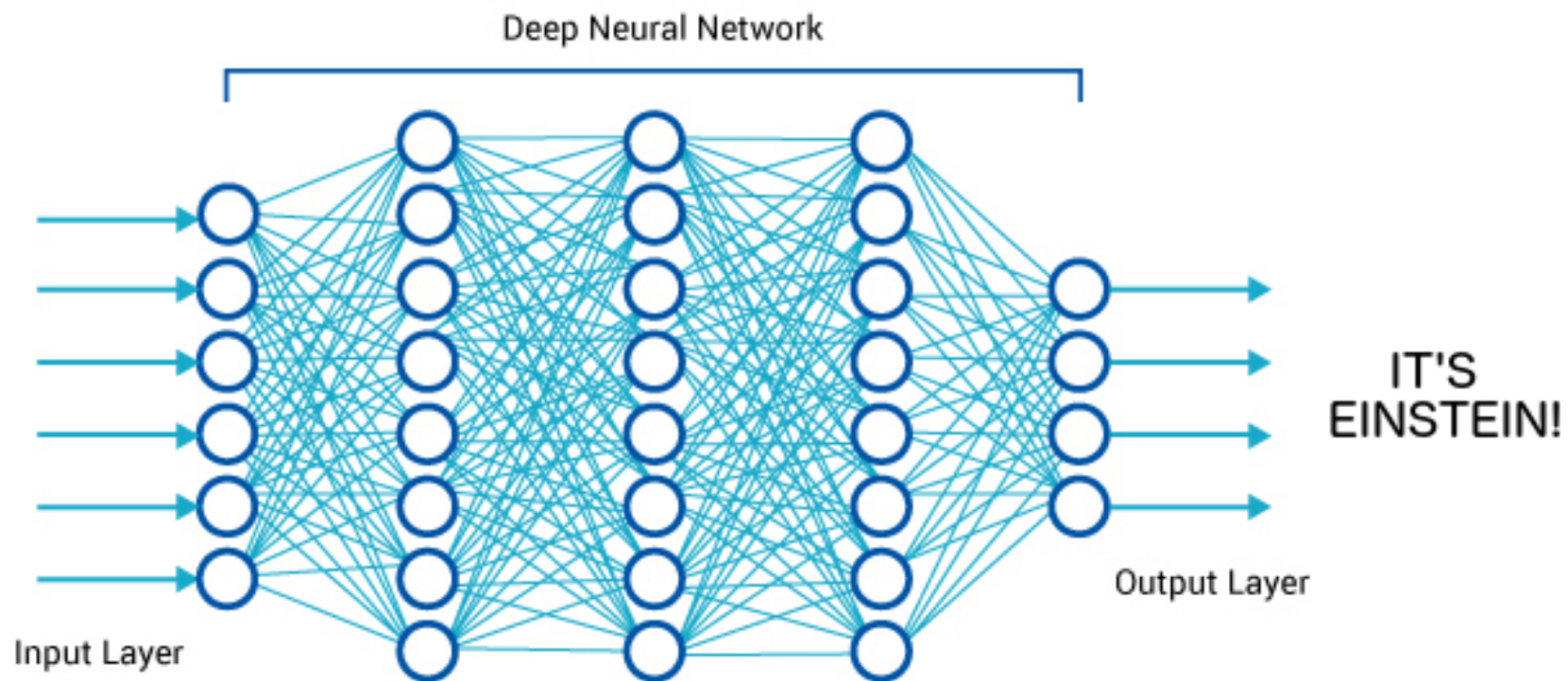
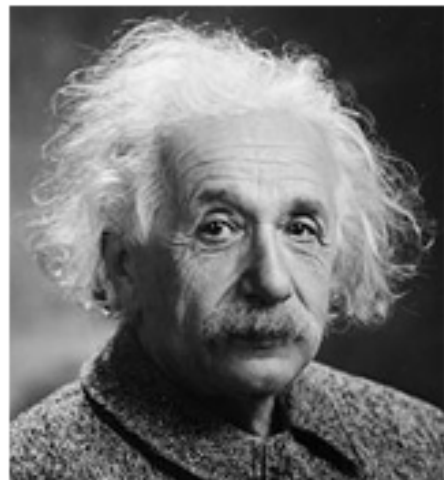


HoG



Textons

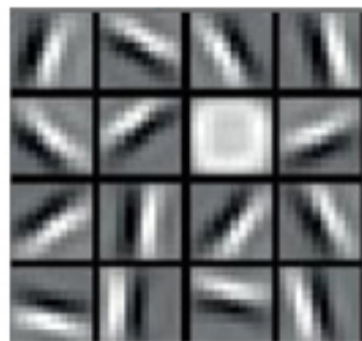




Hidden Layer 1

Hidden Layer 2

Hidden Layer 3



edges

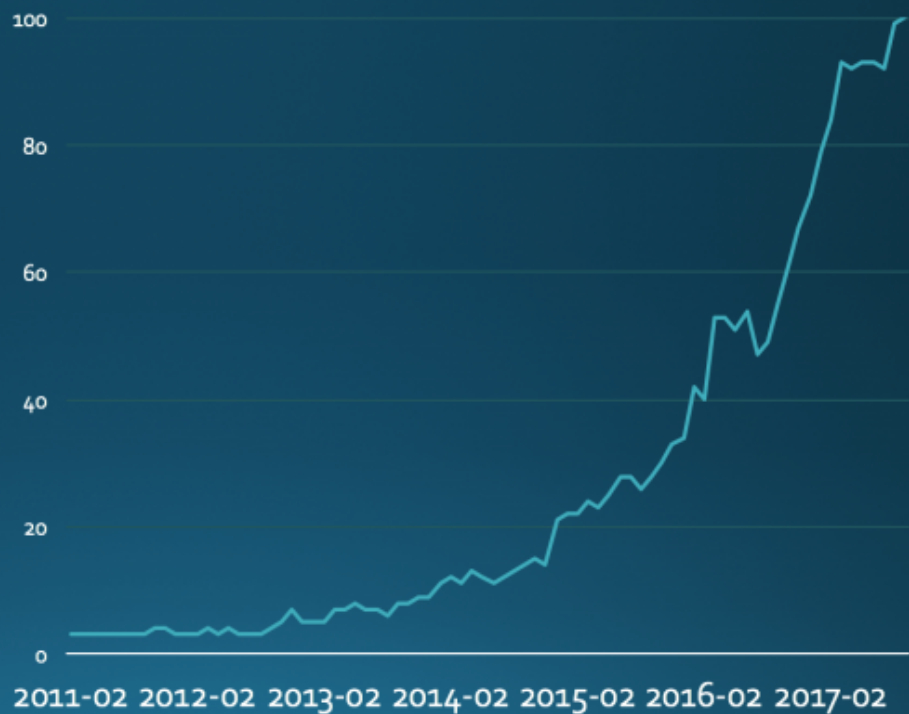


combinations of edges

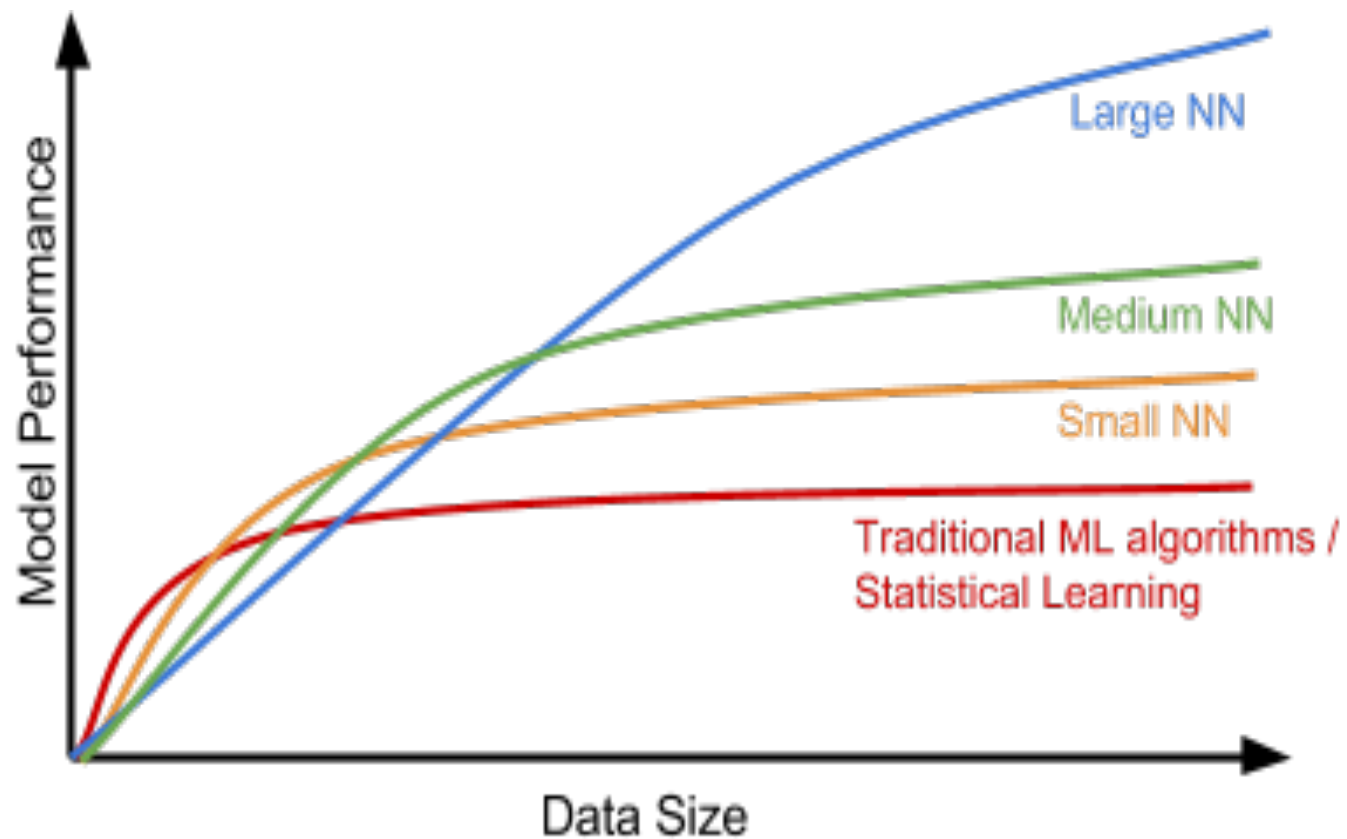


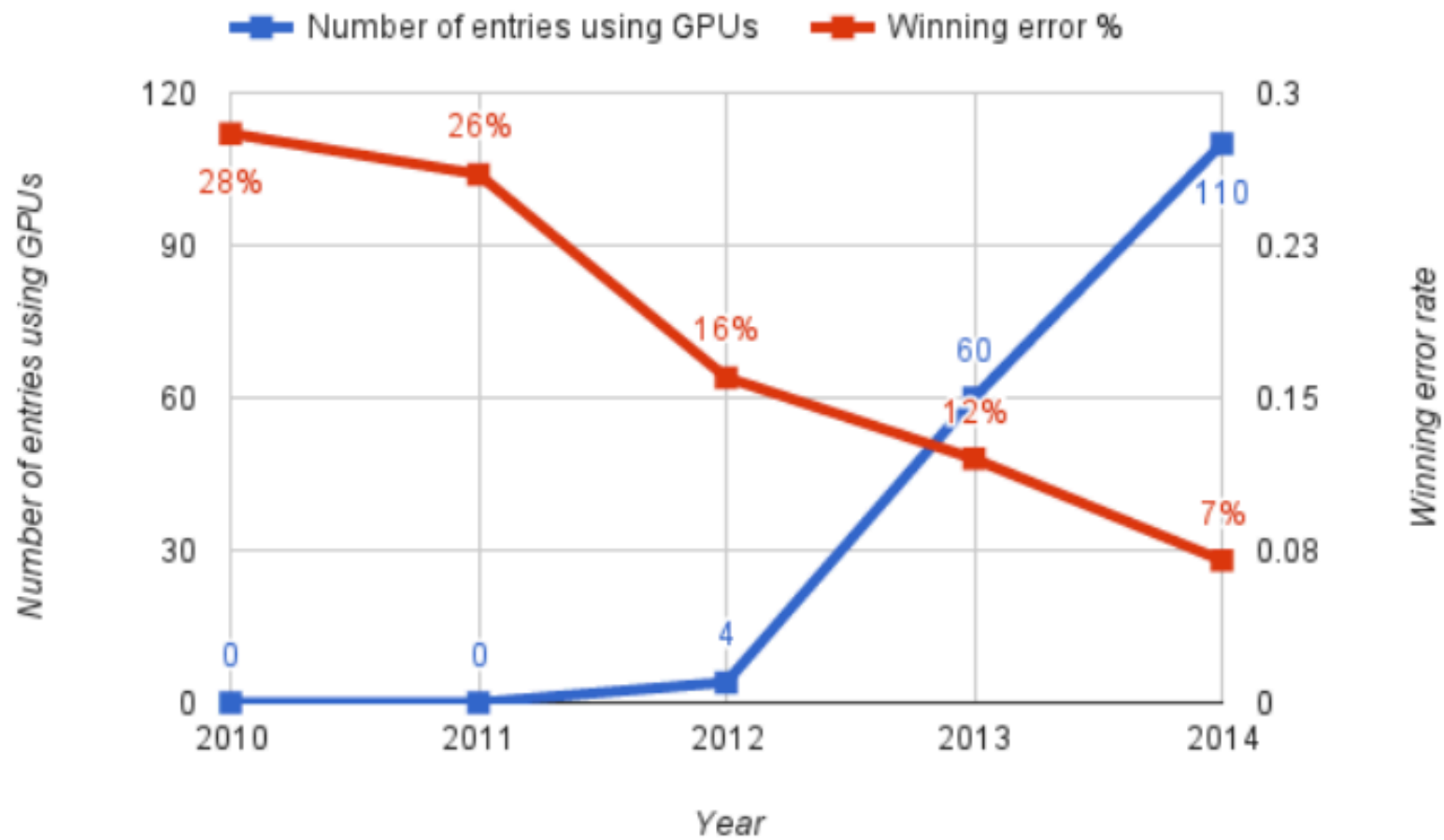
object models

# of Searches for Deep Learning from 2011 to 2017

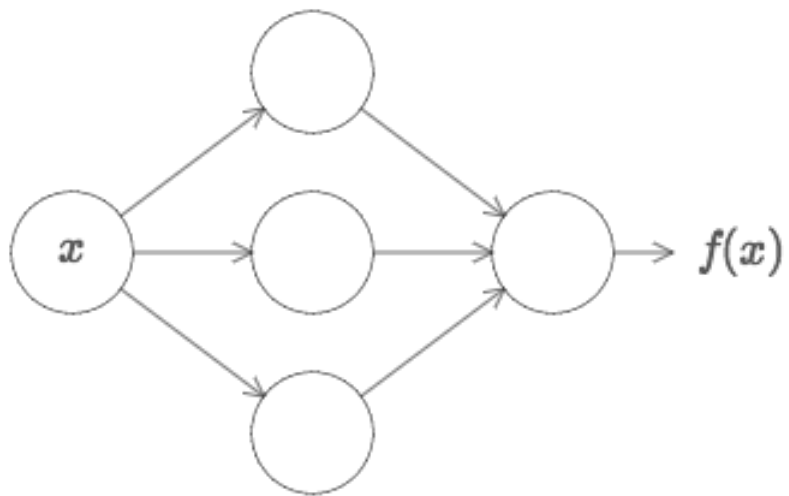
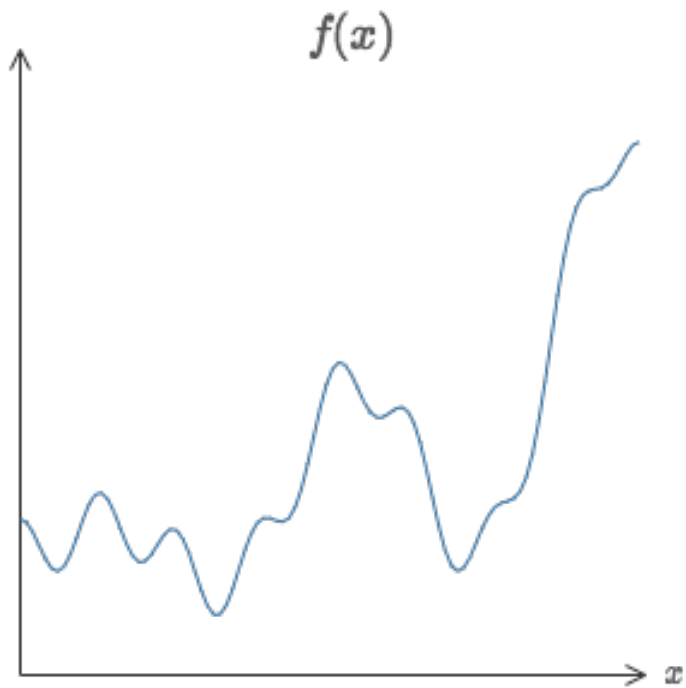


Source: Google Trends. Search term "Deep Learning"

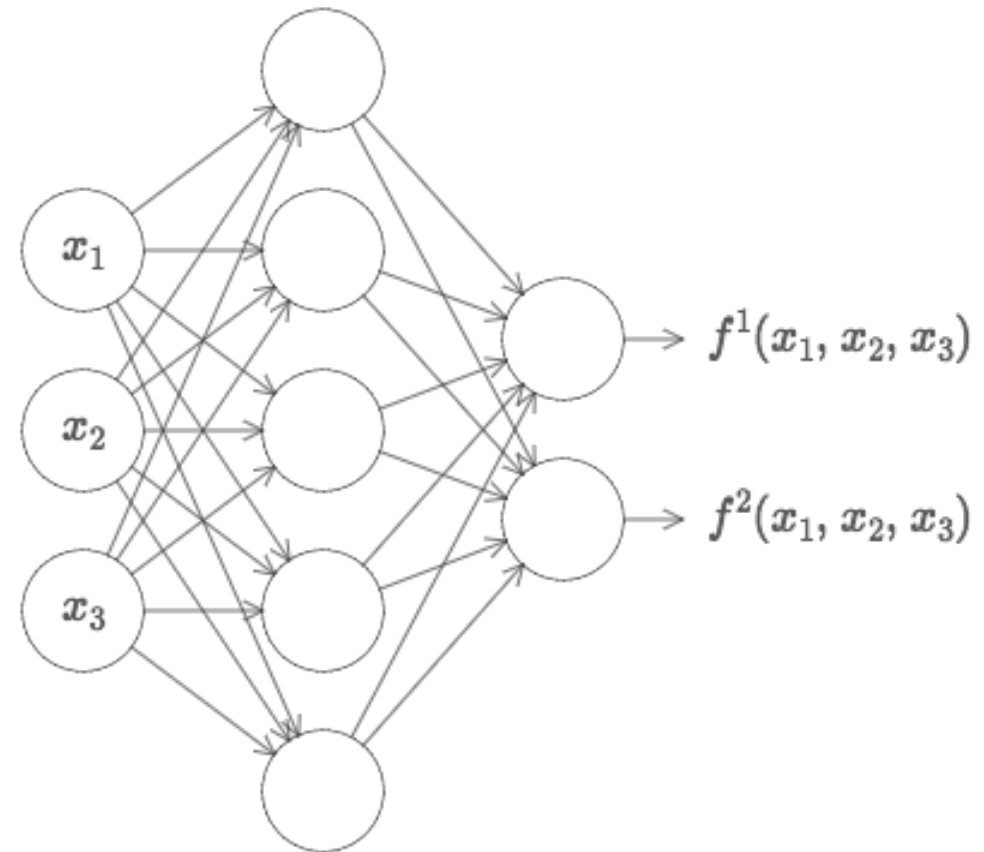




**IMAGENET**  
Krizhevsky, Sutskever, & Hinton, 2012



## Universal Approximation Theorem



Kurt Hornik (1991) *Neural Networks*, 4, 251–257.









Panda



Gibbon

