

# What We Did Not Cover

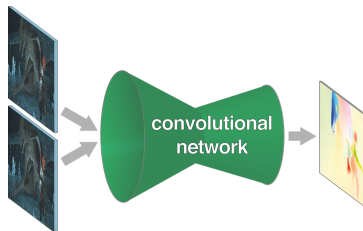
COMPSCI 371D — Machine Learning

# What We Did Not Cover

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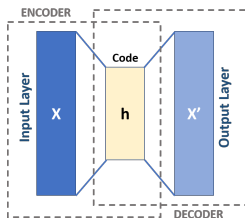
# Much More Detail

- Computationally efficient training algorithms:  
Optimization techniques
- Deep learning architectures for special problems:  
Image motion analysis, video analysis, ...



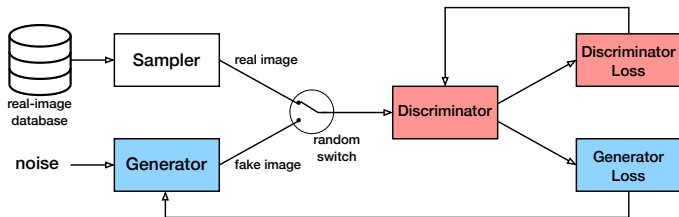
# Beyond Discriminative Neural Networks

- Abstraction for its own sake: Auto-encoders
- A game-theoretical technique to draw from a distribution: Generative Adversarial Networks
- See also recent diffusion methods



Which image is fake?

# Generative Adversarial Networks



- Discriminator guesses if input is real or fake
- Discriminator loss penalizes wrong predictions
- Generator loss penalizes correct predictions
- After training keep only the generator

# Statistical Machine Learning

- How to measure the size of  $\mathcal{H}$ : Vapnik-Chervonenkis dimension, Rademacher complexity
- How large must  $T$  be to get an  $h$  that is within  $\epsilon$  of a performance target with probability greater than  $1 - \delta$ : Probably Approximately Correct (PAC) learning
- $\mathcal{H}$  is *learnable* if there exists a size of  $T$  that is large enough for this goal to be achieved
- Which  $\mathcal{H}$ s are learnable?
- How large must  $S$  be to get a performance measure accurate within  $\epsilon$ : Concentration bounds, statistical estimation theory, PAC-like techniques

# Other Supervised Techniques

- Boosting: How to use many bad predictors to make one good one

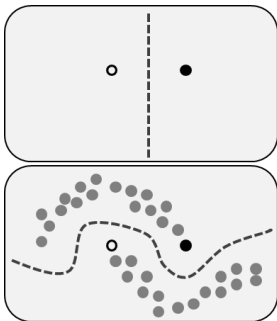
Similar in principle to ensemble predictors, different assumptions and techniques

- Learning to rank

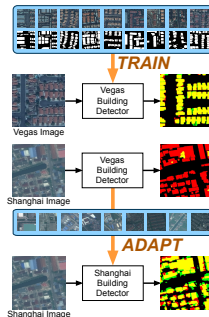
Example: Learning a better Google

# Reducing the Burden of Labeling

- Semi-supervised methods: Build models of the data  $\mathbf{x}$  to leverage sparse labels  $y$
- Domain adaptation: Train a classifier on source-domain labeled data ( $\mathbf{x}, y$ ) and target-domain unlabeled data  $\mathbf{x}$  so that it works well in the target domain



<https://en.wikipedia.org>





# Unsupervised Methods

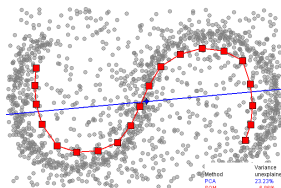
- Dimensionality reduction:

Compressing  $X \subseteq \mathbb{R}^d$  to  $X' \subseteq \mathbb{R}^{d'}$  with  $d' \ll d$

- Principal or Independent Component Analysis (PCA, ICA)
- Manifold learning, GANs

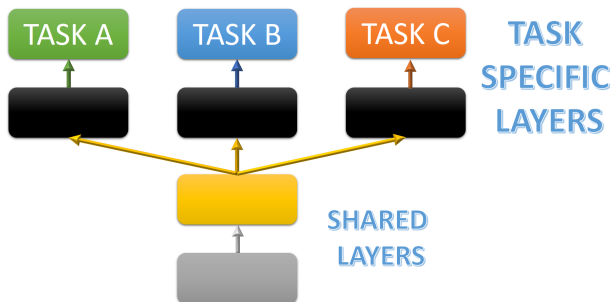
- Clustering:

- $K$ -means
- Expectation-Maximization
- Agglomerative methods
- Splitting methods



# Addressing Multiple Learning Tasks Together

- Multi-task learning: How to learn representations that are common to different but related prediction tasks



# Prediction over Time

- State-space methods
    - Time series analysis
    - Stochastic state estimation
    - System identification
  - Recurrent neural networks
  - Reinforcement learning: Actions over time
- Learning policies underlying observed sequences

