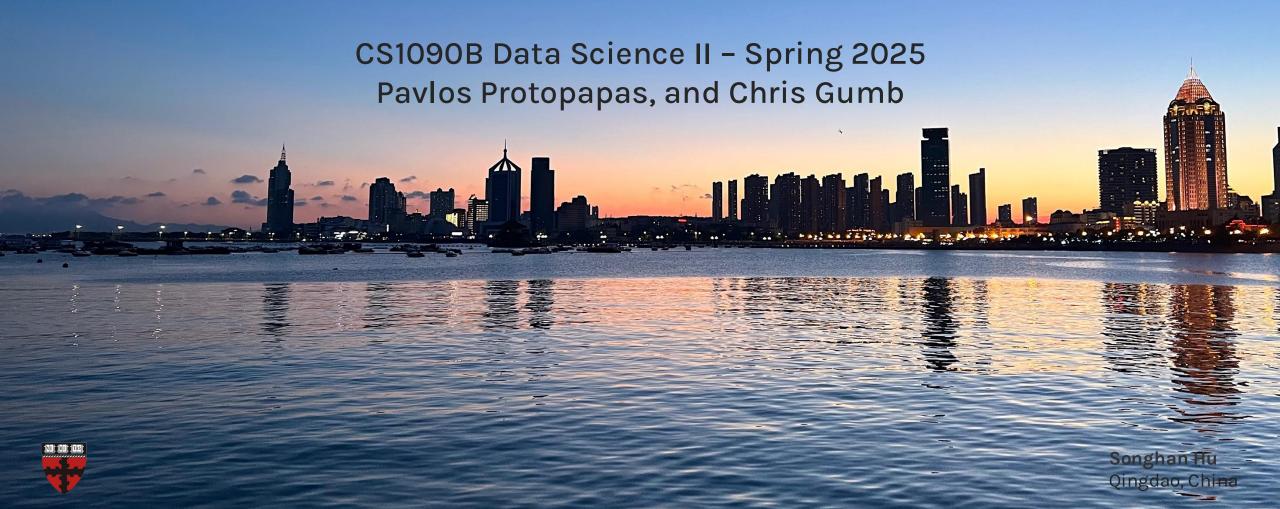
Neural Network Regularization

Part B - Dropout



Outline

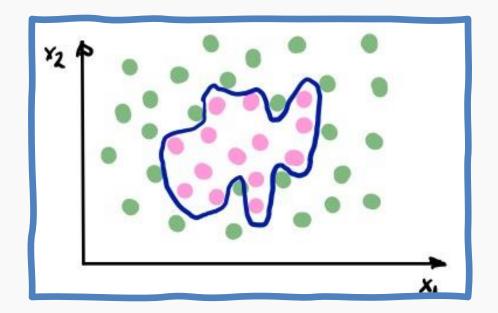
- Regularization of NN
 - Norm Penalties
 - Early Stopping
 - Data Augmentation
 - Dropout

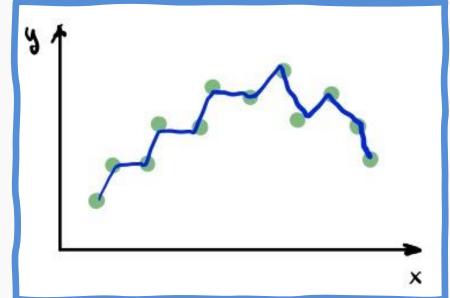
Co-adaptation

Overfitting occurs when the model is sensitive to slight variations on the input and therefore it fits the noise.

L1 and L2 regularizations 'shrink' the weights to avoid this problem.

However, in a large network many units can collaborate to respond to the input while the weights can remain relatively small. This is called co-adaptation.





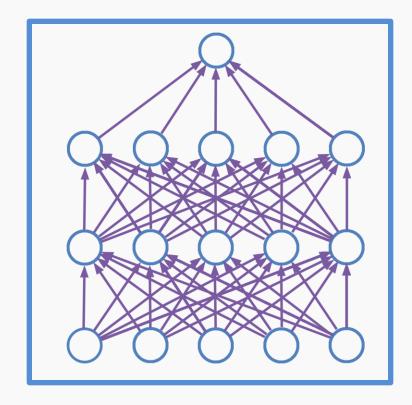
Game Time

How would you stop neuron co-adaptation?

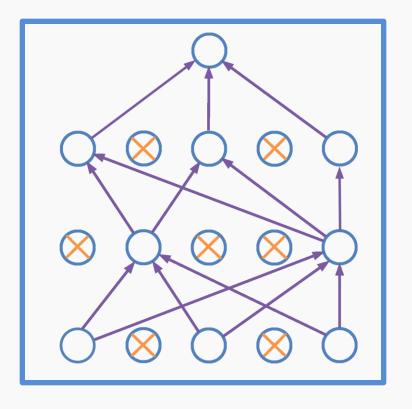
- A. Regularize even more
- B. Force some of the neurons not to participate
- C. Force some of the neurons not to participate for some of the training data
- D. Force some of the neurons not to participate randomly per batch

Dropout

- Randomly set some neurons and their connections to zero (i.e. "dropped")
- Prevent overfitting by reducing co-adaptation of neurons
- Like training many random sub-networks

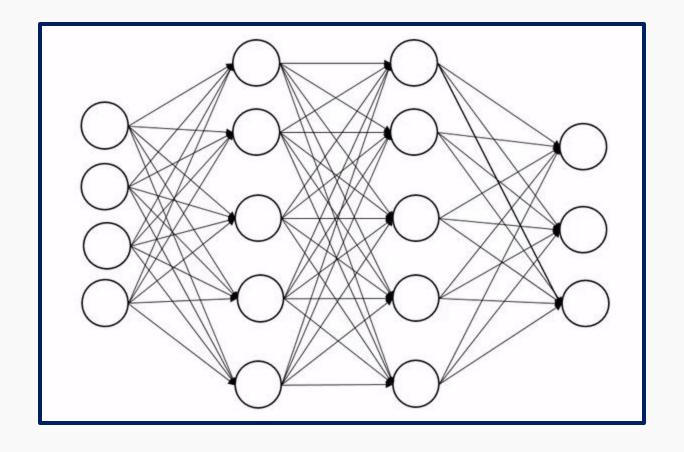


Standard Neural Network



After Applying Dropout

Dropout





Dropout | Training

For each new example in a mini-batch (could be for one mini-batch depending on the implementation):

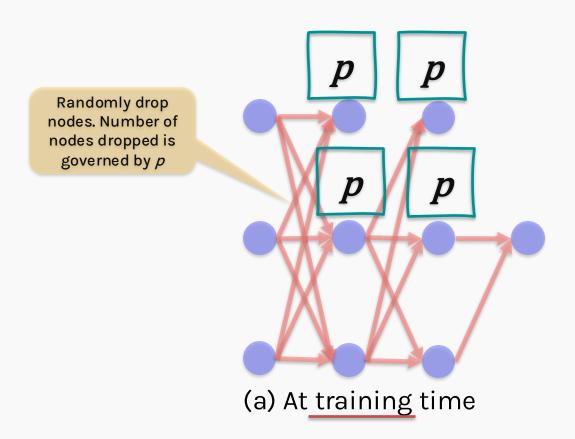
- Randomly sample a binary mask μ independently, where μ_i indicates if input/hidden node i is included
- Multiply output of node i with μ_i , and perform gradient update

Typically:

- Input nodes are included with prob=0.8 (as per original paper, but rarely used)
- Hidden nodes are included with prob=0.5

Dropout | Prediction

• We can think of dropout as training many of sub-networks

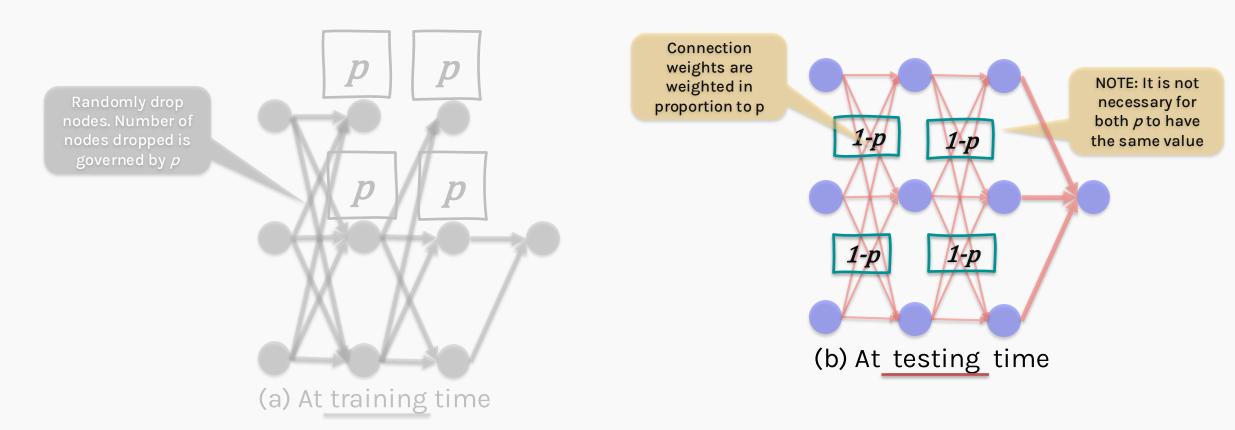


Dropout | Prediction

What do you think occurs at testing time?

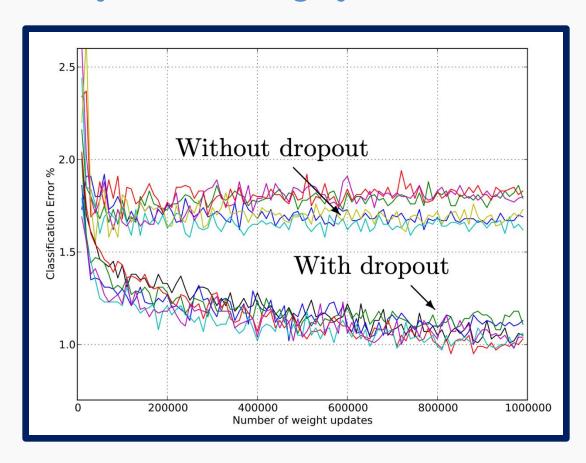
Dropout | Prediction

• At **test time**, we can "aggregate" over these sub-networks by **reducing** connection weights in proportion to dropout probability, p



Dropout

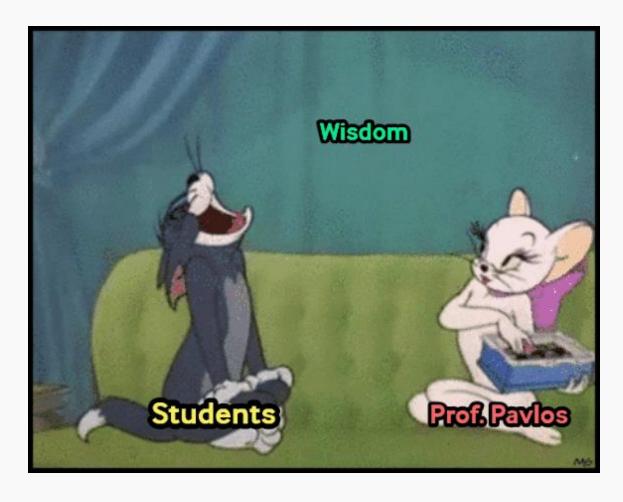
Widely used and highly effective



Test error for different architectures with and without dropout.

The networks have 2 to 4 hidden layers each with 1024 to 2048 units.

 Proposed as an alternative to ensemble methods, which is too expensive for neural nets



Thank you