

NN: what and why?

Training:

- Stochastic gradient descent (SGD)
- Random initialization
- How to compute gradients

Regularization

- Early stopping
- Dropout

Hidden layer: a bunch of logistic regression classifiers

- parameters:  $w_j$  and  $b_j$  for each classifier
- equivalently: matrix  $W$  ( $h \times d$ ) and vector  $b$  (length  $h$ )
- Produces “activations” = learned feature vector
- Parameters of model are  $\theta = (W, b, v, c)$

Final layer: a linear classifier

- E.g. if logistic regression, has parameter vector  $v$  and bias  $b$

Do we need “non-linearity”?

- Without sigmoid, it becomes a linear function of  $x$ , not desired
- So we need a non-linear function between two layers
- Options
  - Sigmoid( $z$ ) =  $1/(1+e^{-z})$
  - Tanh( $z$ ) =  $(e^{2z}-1)/(e^{2z}+1)$
  - ReLU( $z$ ) =  $\max(z, 0)$

Expressiveness of NN

- 2 layer NN can solve XOR (which can't be solved by linear classifier)

Universal Approximation

- any function can be approximated by a 2 layer neural network with enough hidden units

Multi-layer perceptron

Training objectives

- loss function is the same as that of logistic regression
  - $g(x) = w^T x + b$       $g(x) = v^T \sigma(Wx + b) + c$
  - Loss =  $1/n (\sum -\log \sigma(y^i g(x^i)))$
  - More generally, loss =  $1/n (\sum L(y^i, g(x^i)))$
- SGD
  - Sample a batch of  $B$  of examples from the training dataset
  - Do the update on gradient using only the Batch (much faster than normal GD with large dataset)
  - In practice, partition training examples into batches -> use all examples
  - Batch size
    - Large -> more accurate gradient, slower
    - Smaller -> faster, less accurate updates

For NN, initialization is important because it's non-convex, may stuck in local minimum.

- initialize every entry in  $W$  to a small random number
  - depends on “fan-in”, “fan-out”

- Xavier initialization
- He initialization
- Pytorch

#### Regularization

- Weight decay AKA L2 Regularization
- Prevent overfitting by stopping training before overfit too much
  - save checkpoints, if dev set starts to increase continuously, stop training
- Dropout
  - randomly drop out some neurons by setting their values to 0