#### **Advanced techniques**

#### Lecture 6

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# Hyperparameter search and cross validation

## Outline

#### 1. Hyperparameter search

# *L* of layers, #  $n^{[\ell]}$  of hidden neurons, activation learning rate, betas, batch size, # *T* of epochs, regularization factor, dropout rate, ...

2. Cross validation

## **# of layers**

Just begin with a **single hidden** layer.

# **Gradually (linearly)** ramp up # of hidden layers **until not overfitting.**

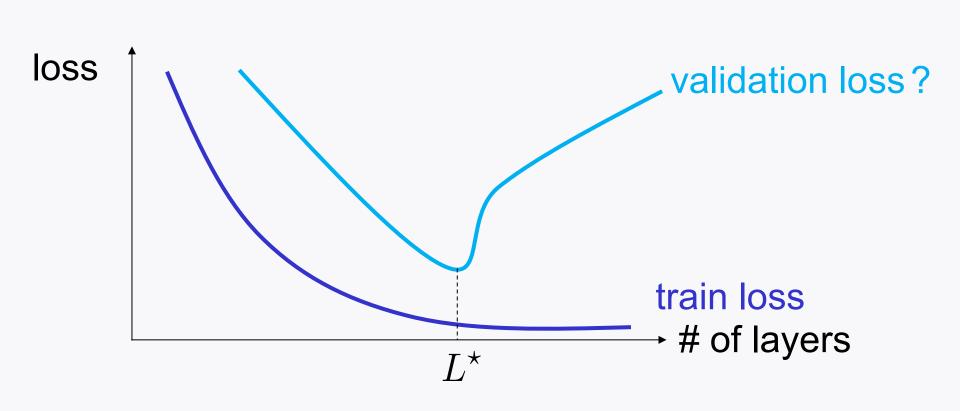
Here use the same number of hidden neurons for all hidden layers.

Set the number of hidden neurons around one half of the number of input neurons.

## # of layers vs. loss



## # of layers vs. loss



## # of hidden neurons

Two approaches:

1. Fewer neurons for deeper layers

2. Same size for all hidden layers:

Linearly increase the size until not overfitting.

## **Activation functions**

A default setup:

Hidden layers: ReLU

Output layer: Softmax for multi-class classification

## Optimizer

### A default use: Adam

Default parameters:  $(\beta_1, \beta_2) = (0.9, 0.999)$ 

Two approaches for a choice of the learning rate:

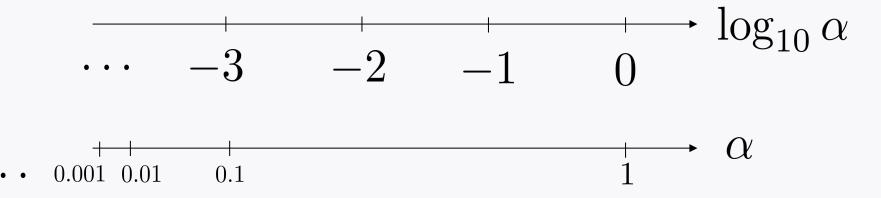
- 1. Learning rate decaying
- 2. Fixed (e.g.,  $\alpha=0.001$  )

## How to choose a fixed value of $\,\alpha\,$

Do not use a linear-scale grid search.

Try **random** values and **then do a fine search** around the good choices.

Grid scale for the fine search: Log scale



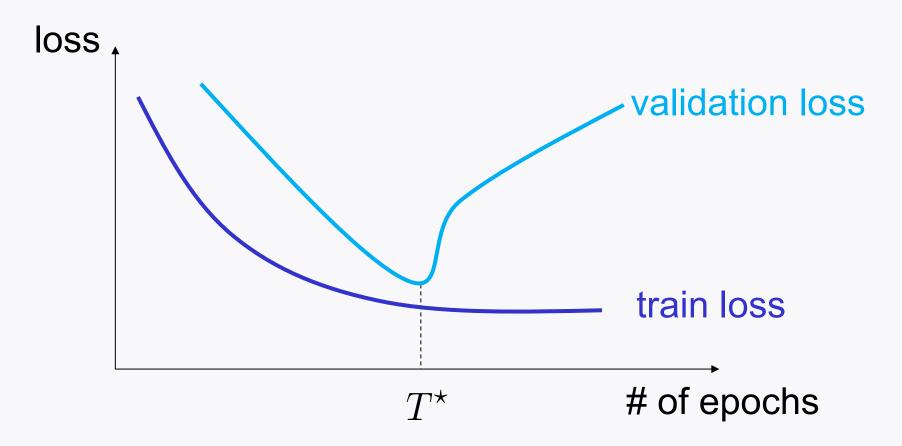
#### **Batch size**

#### A common choice: Power of two.

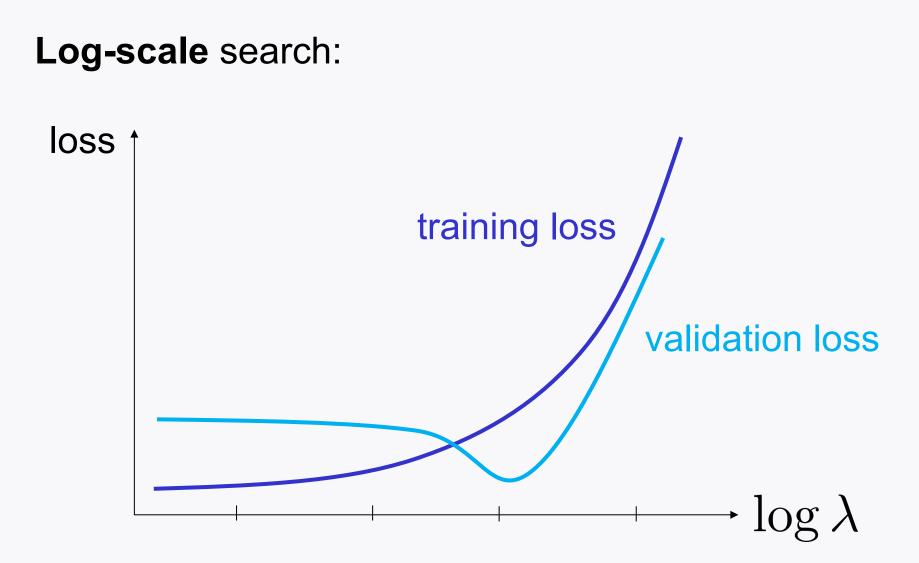
#### 4, 8, 16, 32, 64, 128, 256

## **# of epochs**

#### Choose according to early stopping:



## **Regularization factor**



## **Dropout rate**

## A typical choice: p = 0.5

# A good range: $0.2 \le p \le 0.8$

## **Cross validation**

Purpose: Obtain reliable validation loss via averaging.

Example: 4-fold cross validation

val train	train	train	test	
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 $\rightarrow$  Compute a validation loss, say val<sub>1</sub>

Take the 2<sup>nd</sup> partition for val:

train	val	train	train	test
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 $\rightarrow$  Compute a corresponding loss: val<sub>2</sub>

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## **Cross validation**

average loss.

val	train	train	train	test	$val_1$
train	val	train	train	test	$val_2$
train	train	val	train	test	$val_3$
train	train	train	val	test	$val_4$

Take the average over the 4 losses:

val loss = 
$$\frac{val_1 + val_2 + val_3 + val_4}{4}$$
  
Choose a hyperparameter that minimizes

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## A final model w.r.t. the best hyperparameter?

val	train	train	train	test	$] model_1$
train	val	train	train	test	$model_2$
train	train	val	train	test	$model_3$
train	train	train	val	test	$model_4$

Which one to take among the four models?

A final model is the one trained based on:

train	train	train	train	test
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## What is next?

One important question:

Can DNNs be specialized?

CNNs: Image data

RNNs: Text/audio data (language) and any sequential data

## **Outline of Day 3 lectures**

Focus on CNNs.

Specifically we will:

- 1. Investigate how CNNs were developed;
- Study the two key building blocks;
  Conv layer
  Pooling layer
- 3. Discuss popular CNN architectures.