

作业通关大攻略

# General Guidance

Hung-yi Lee 李宏毅

# Framework of ML

Training data:  $\{(\mathbf{x}^1, \hat{y}^1), (\mathbf{x}^2, \hat{y}^2), \dots, (\mathbf{x}^N, \hat{y}^N)\}$

本课:

$\hat{y}$ : 真实值

$y$ : 预测值

Testing data:  $\{\mathbf{x}^{N+1}, \mathbf{x}^{N+2}, \dots, \mathbf{x}^{N+M}\}$

Speech Recognition 作业2



$\mathbf{x}$ :  $\hat{y}$ : phoneme

Image Recognition 作业3



$\mathbf{x}$ :  $\hat{y}$ : soup

Speaker Recognition 作业4



$\mathbf{x}$ :  $\hat{y}$ : John  
(speaker)

Machine Translation 作业5

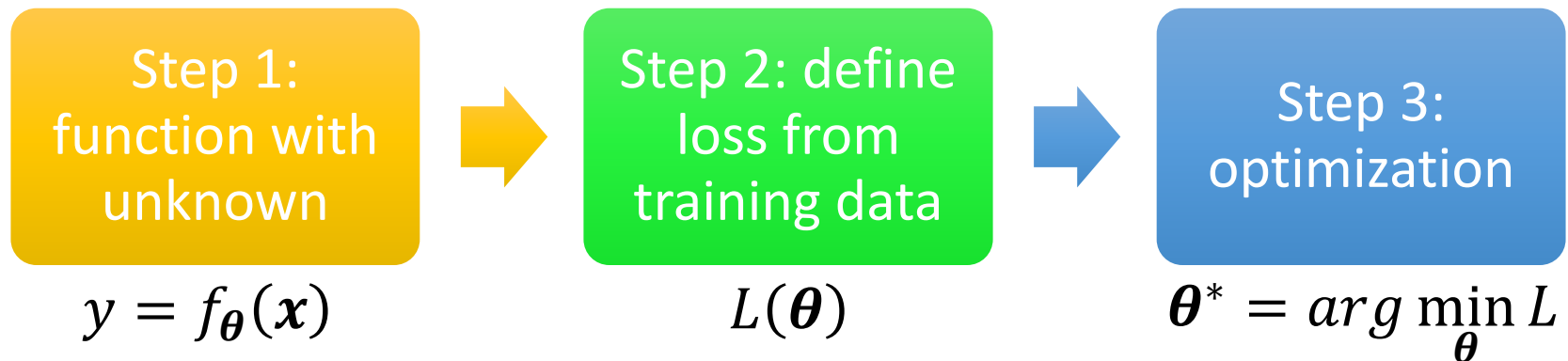
$\mathbf{x}$ : 痛みを知れ

$\hat{y}$ : 了解痛苦吧

# Framework of ML

Training data:  $\{(\mathbf{x}^1, \hat{y}^1), (\mathbf{x}^2, \hat{y}^2), \dots, (\mathbf{x}^N, \hat{y}^N)\}$

Training:

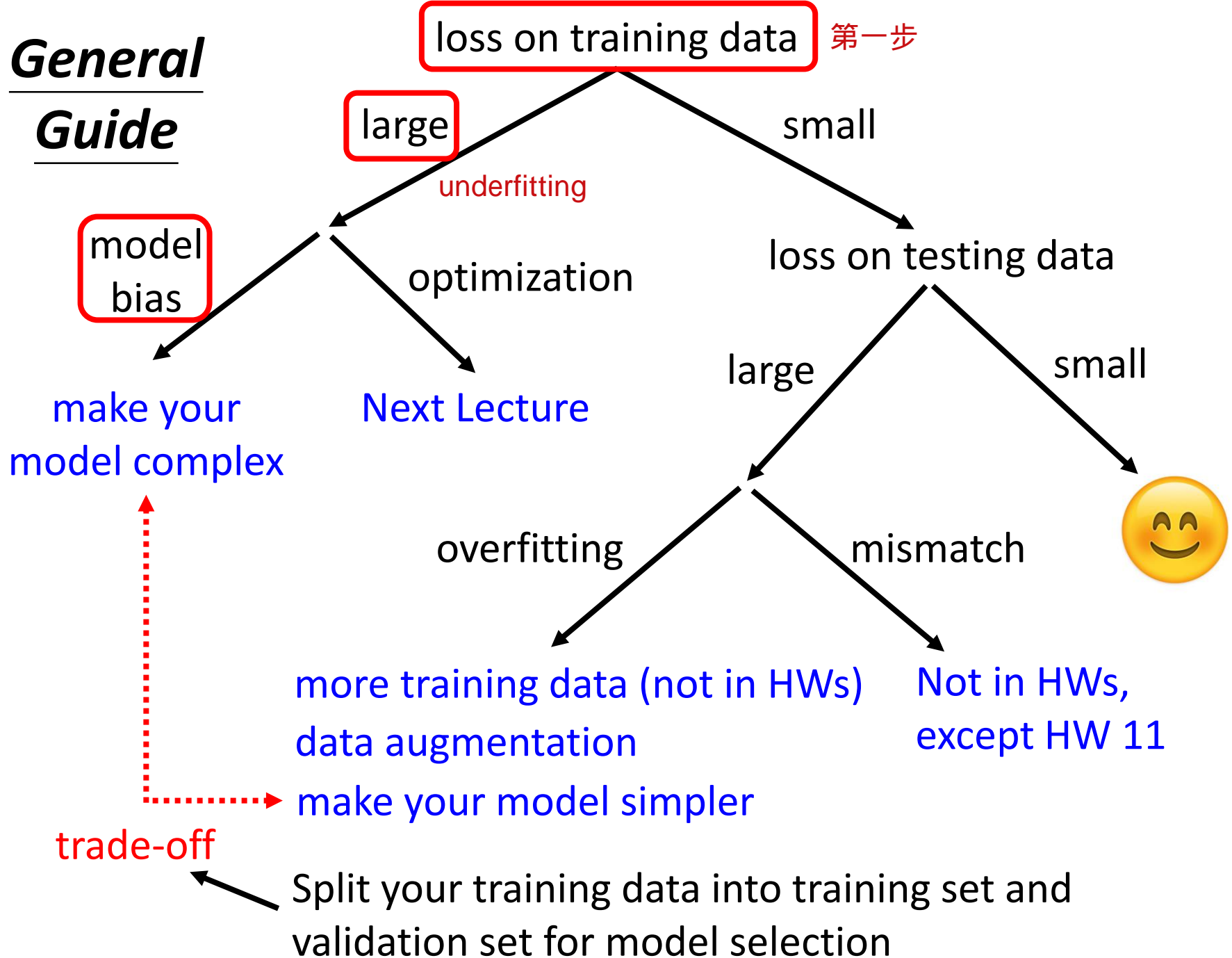


Testing data:  $\{\mathbf{x}^{N+1}, \mathbf{x}^{N+2}, \dots, \mathbf{x}^{N+M}\}$

Use  $y = f_{\theta^*}(\mathbf{x})$  to label the testing data

$\{y^{N+1}, y^{N+2}, \dots, y^{N+M}\}$  **➡** Upload to Kaggle

# General Guide

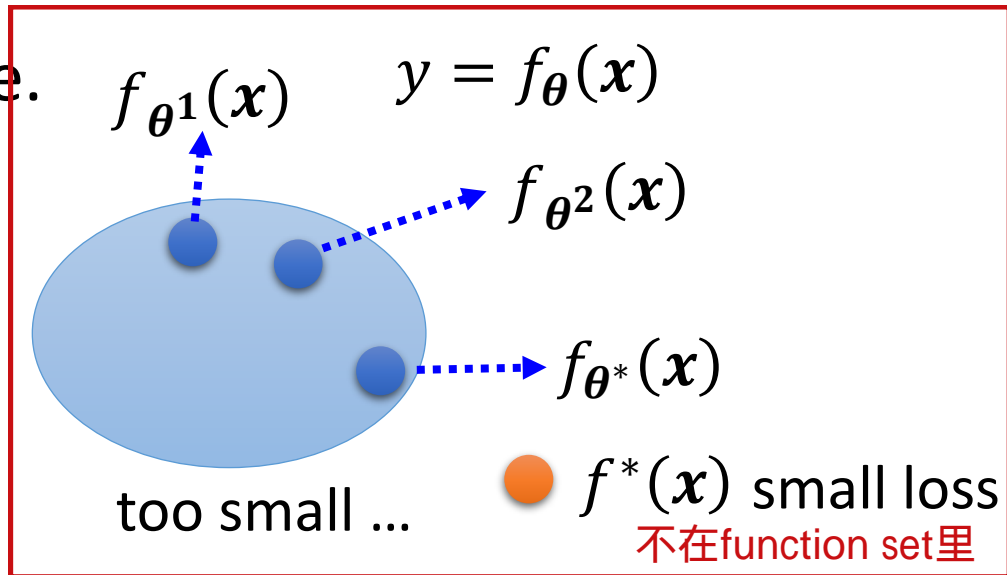


# Model Bias

underfitting的第一种可能：model bias

- The model is too simple.

find a needle in a haystack ...  
... but there is no needle



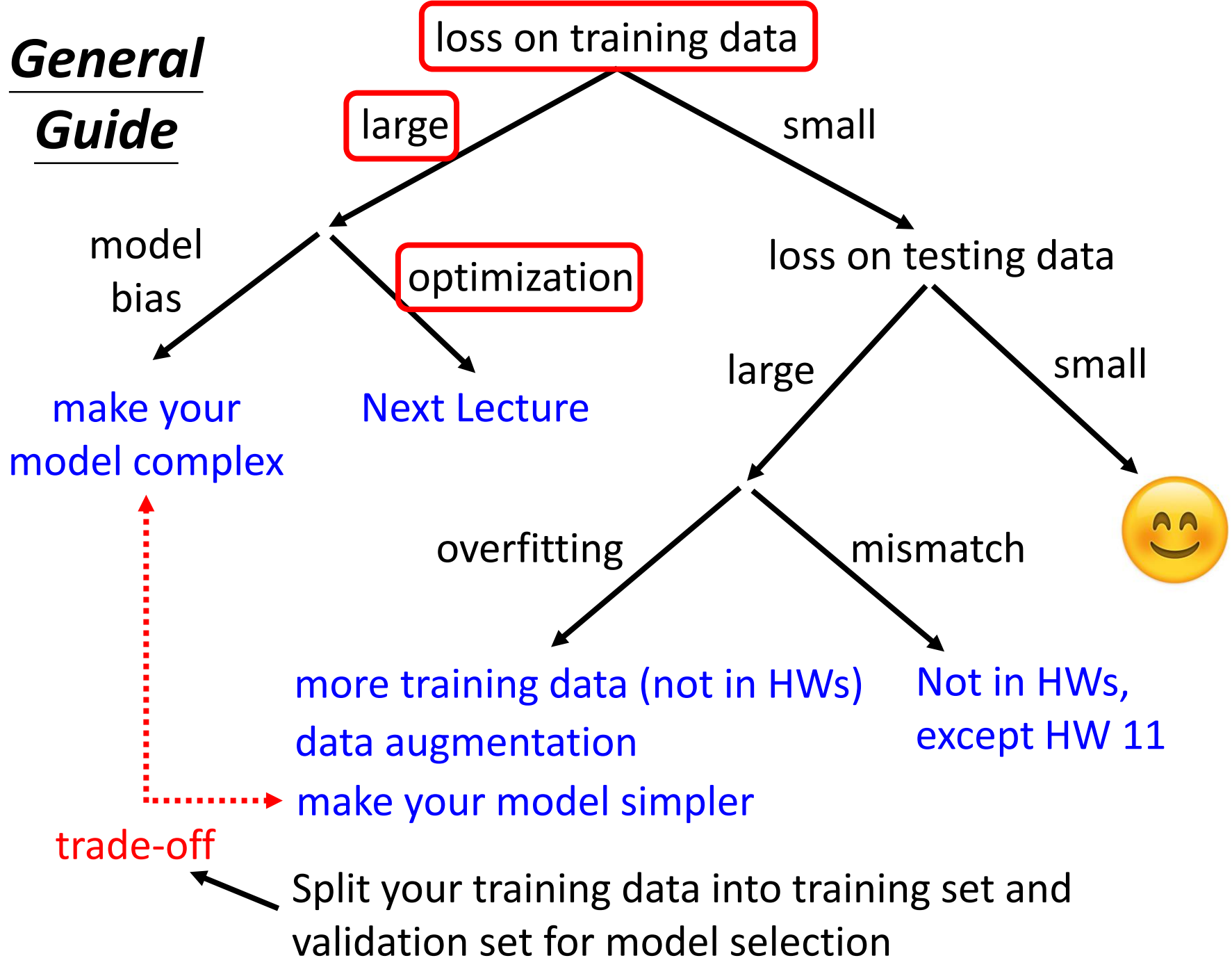
- Solution: redesign your model to make it **more flexible**

$$y = b + wx_1 \xrightarrow{\text{More features}} y = b + \sum_{j=1}^{56} w_j x_j$$

Deep Learning  
(more neurons, layers)

$$y = b + \sum_i c_i \text{sigmoid} \left( b_i + \sum_j w_{ij} x_j \right)$$

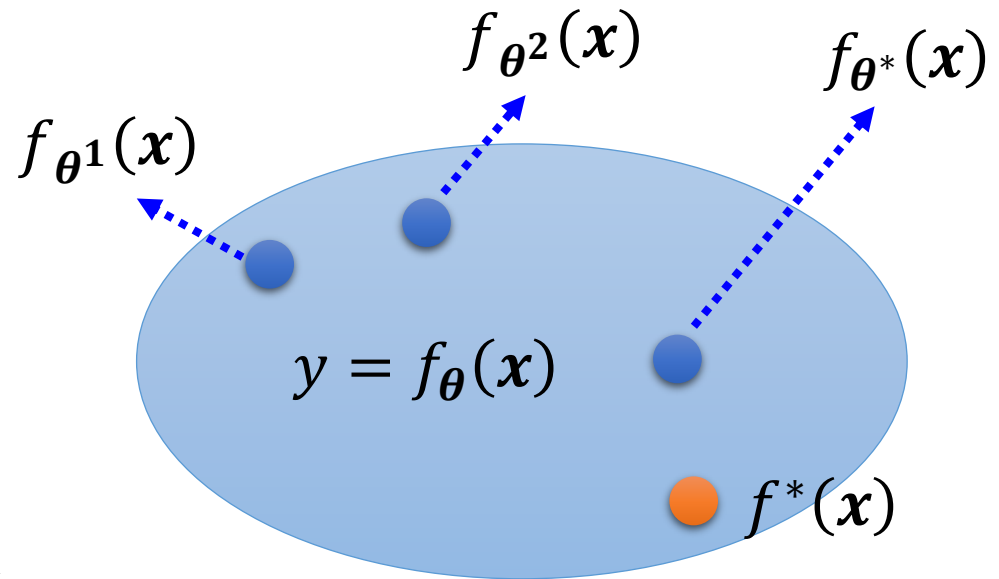
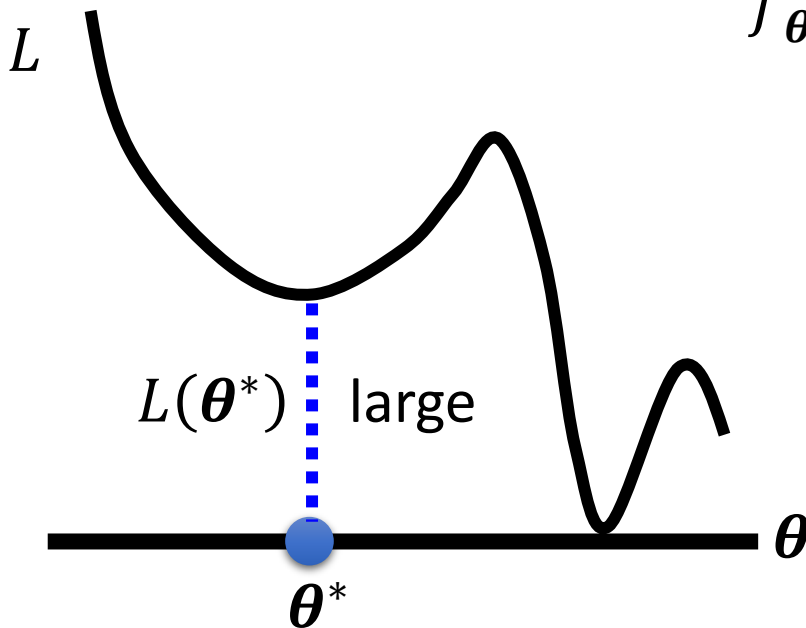
# General Guide



underfitting的第二种可能：  
optimization做得不好

# Optimization Issue

- Large loss not always imply model bias. There is another possibility ...



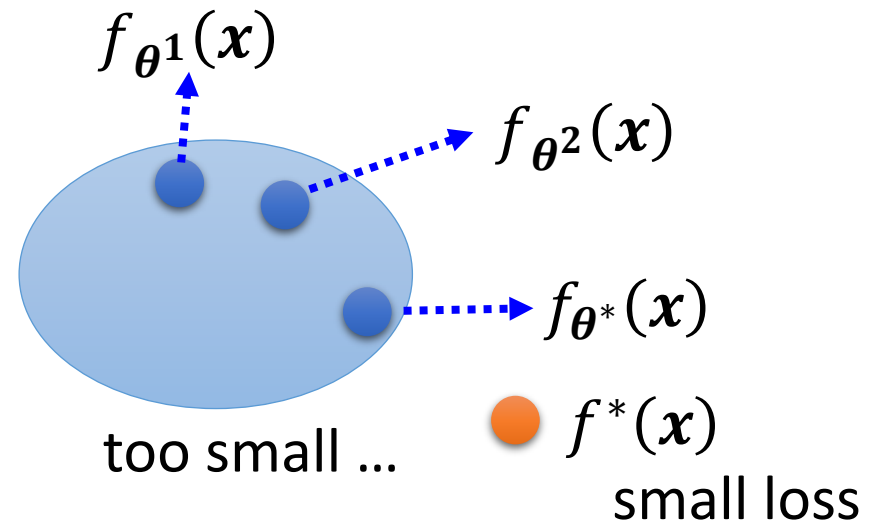
A needle is in a haystack ...

... Just cannot find it.

大海捞针，针确实在海里，但是就是找不到

## Model Bias

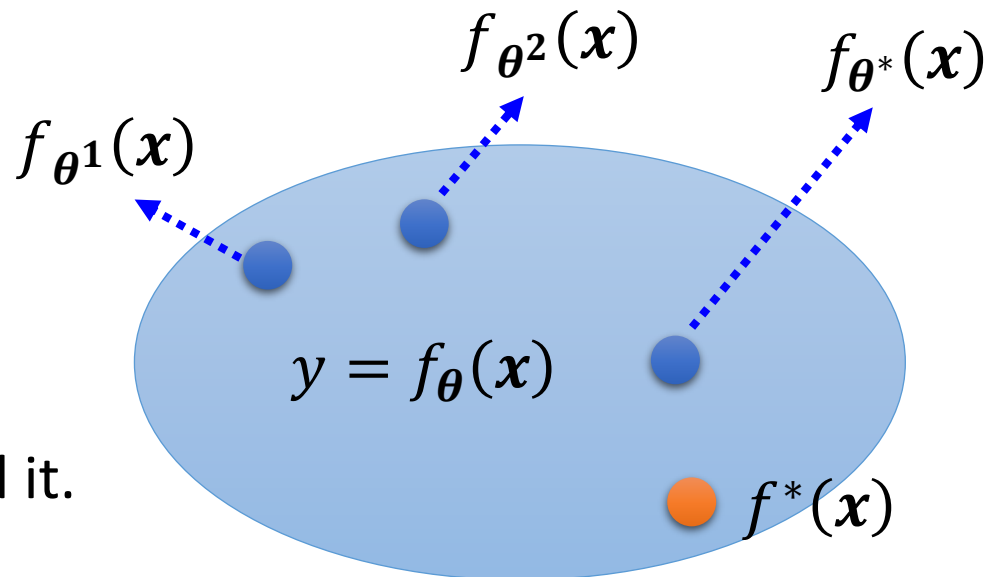
find a needle in a haystack ...  
... but there is no needle



**Which one???**

## Optimization Issue

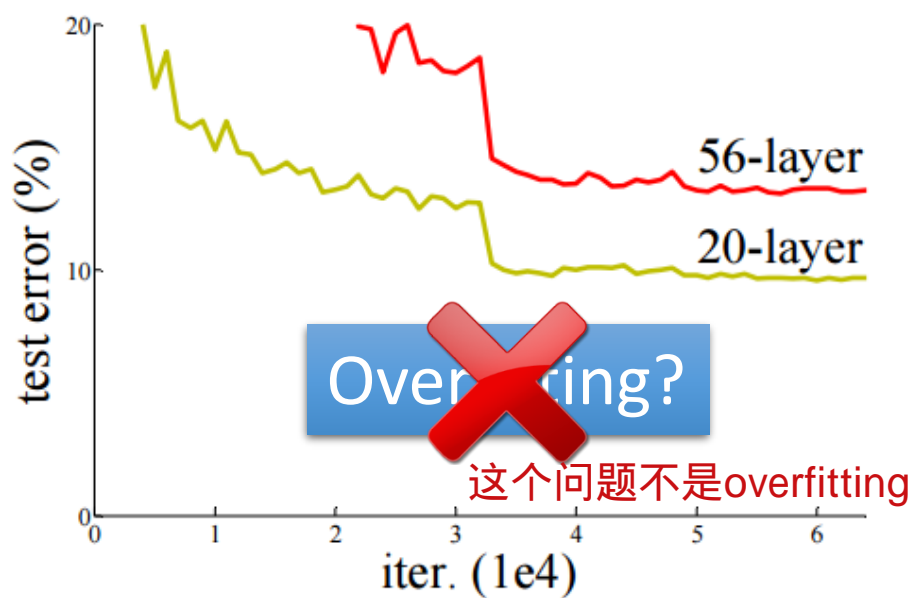
A needle is in a haystack ...  
... Just cannot find it.





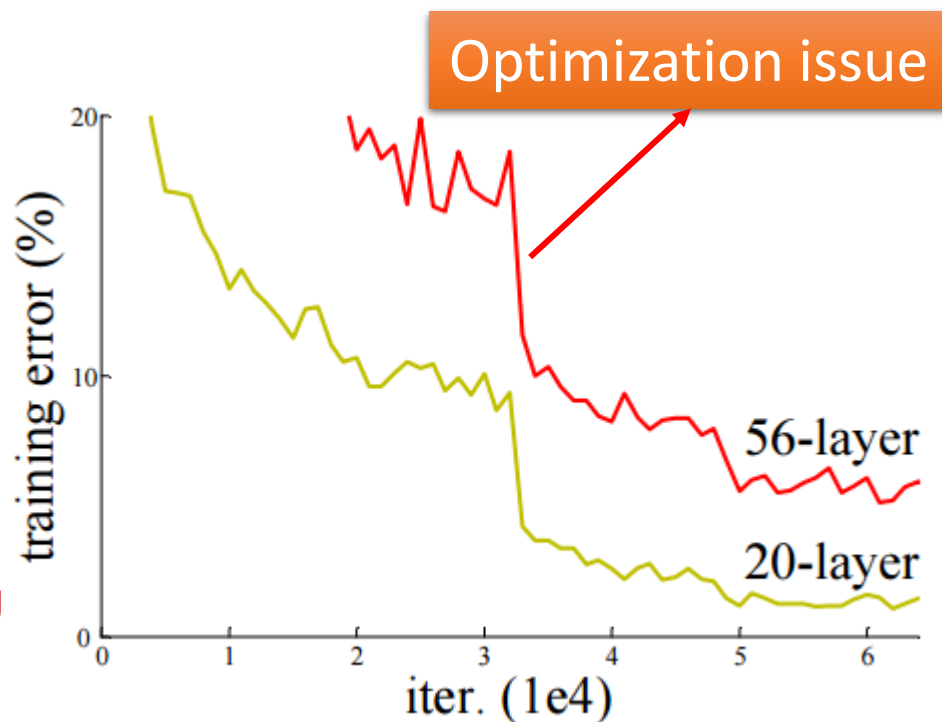
# Model Bias v.s. Optimization Issue

- Gaining the insights from comparison



Testing Data

overfitting : training上好 , testing上差



Training Data

56层的network更flexible , 但是error高 , 说明 optimization出现了问题

# Optimization Issue

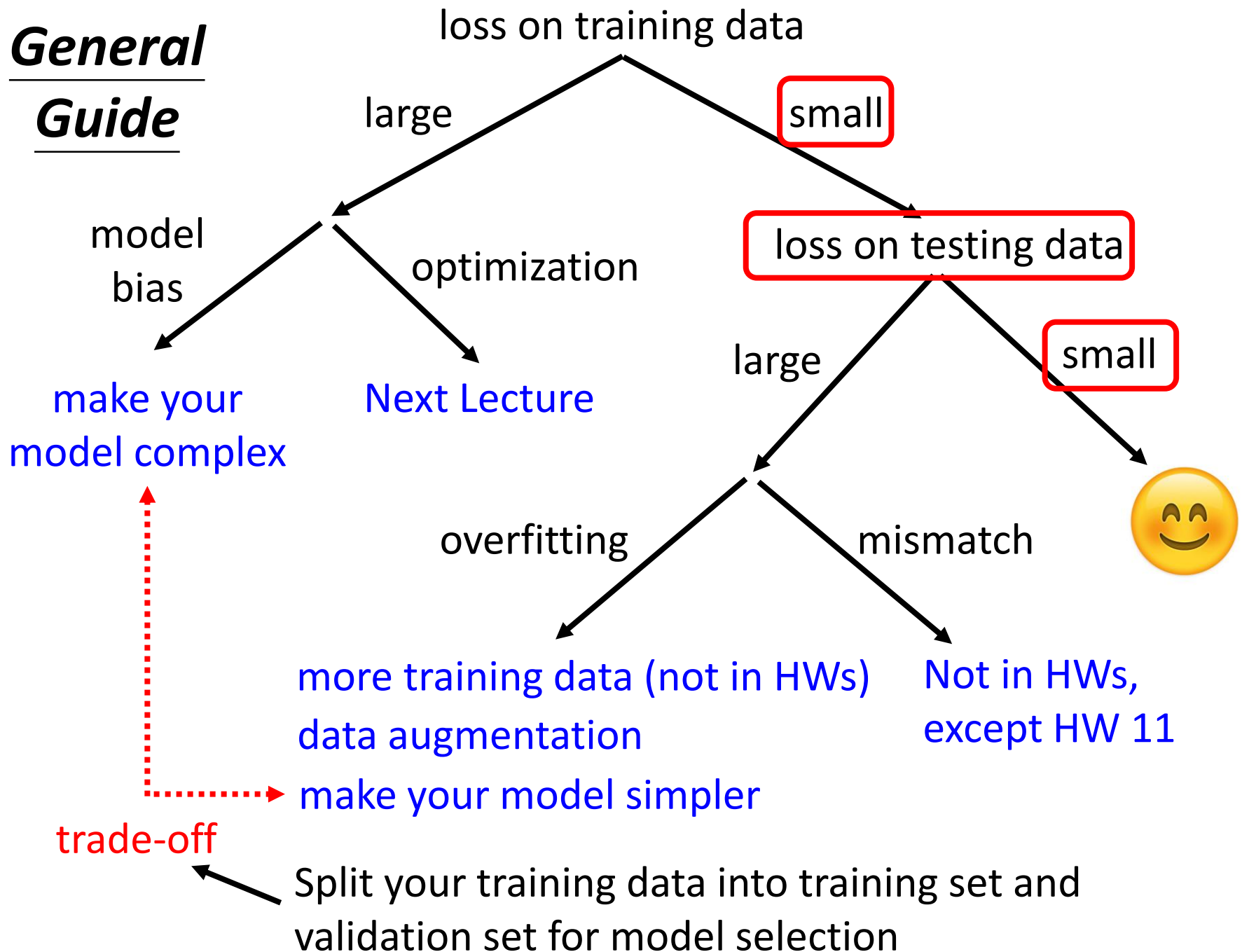
先跑比较浅的network，或者用不是deep learning的简单的模型。

- Gaining the insights from comparison
- Start from **shallower networks** (or other models), which are easier to optimize.
- If deeper networks do not obtain smaller loss on **training data**, then there is optimization issue.

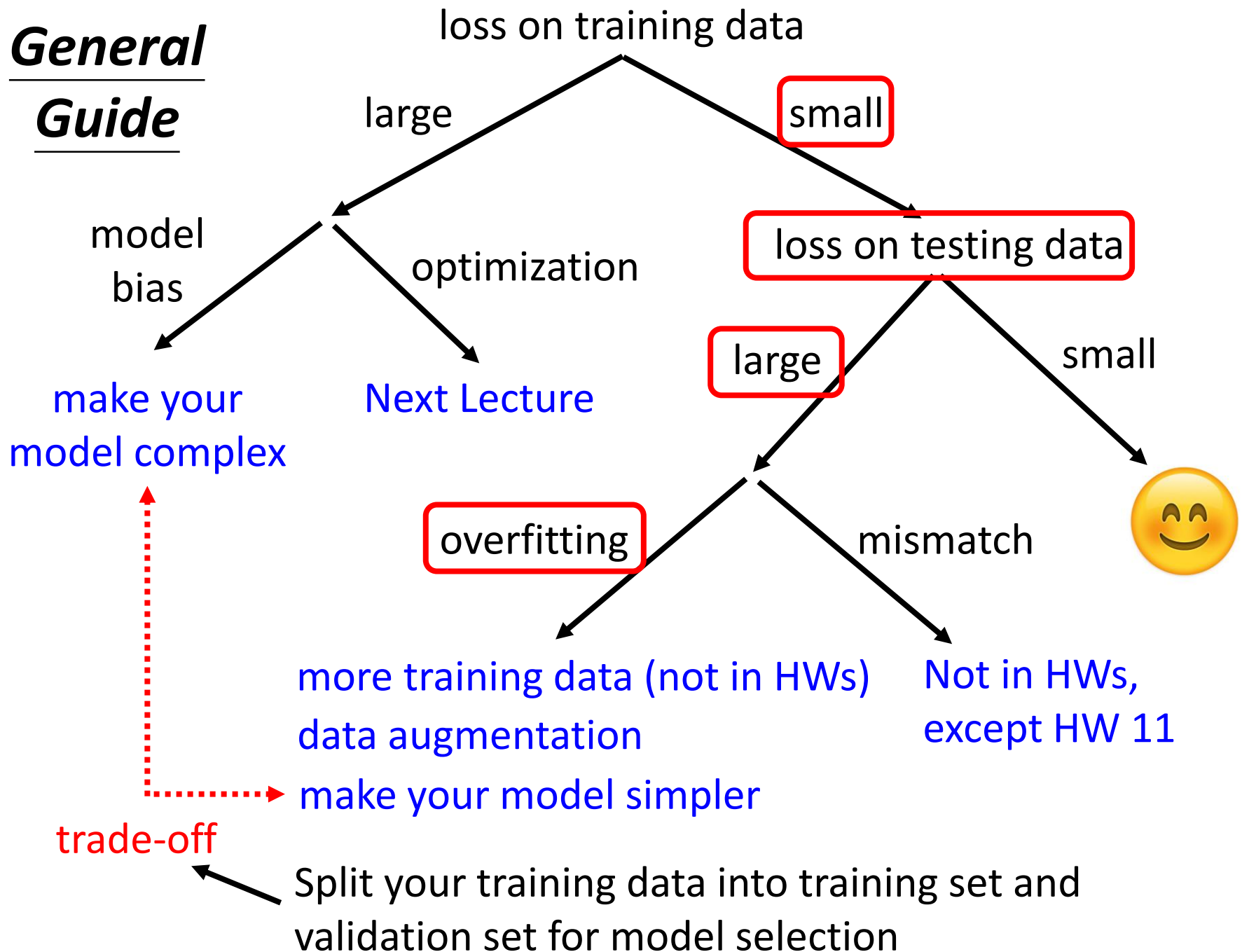
	1 layer	2 layer	3 layer	4 layer	5 layer
2017 – 2020	0.28k	0.18k	0.14k	0.10k	0.34k

- Solution: More powerful **optimization technology** (next lecture)

# General Guide



# General Guide



# Overfitting

- Small loss on training data, large loss on testing data. Why?

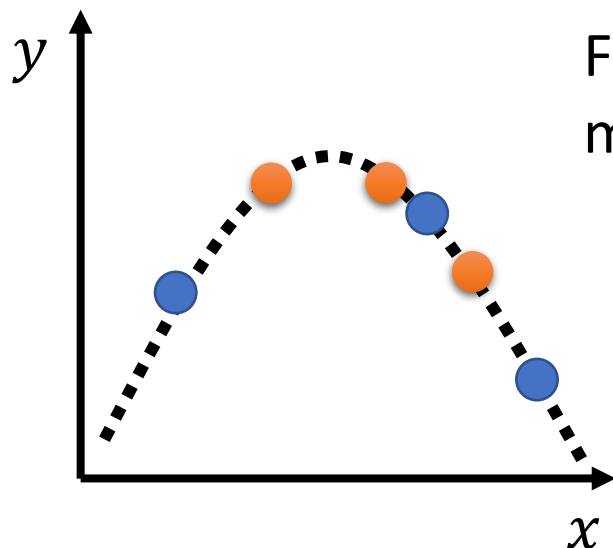
## An extreme example

Training data:  $\{(\mathbf{x}^1, \hat{y}^1), (\mathbf{x}^2, \hat{y}^2), \dots, (\mathbf{x}^N, \hat{y}^N)\}$

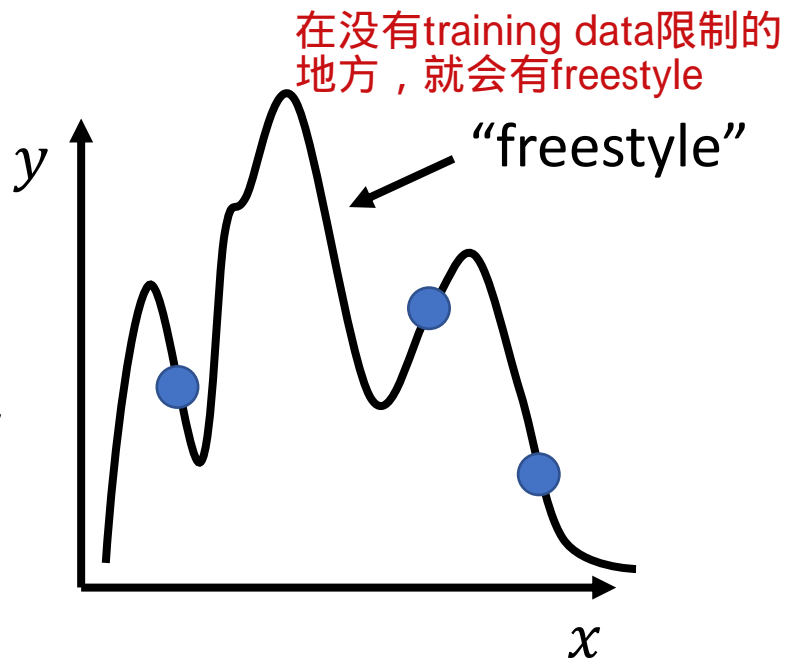
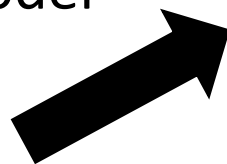
$$f(\mathbf{x}) = \begin{cases} \hat{y}^i & \exists \mathbf{x}^i = \mathbf{x} \\ random & otherwise \end{cases} \quad \text{Less than useless ...}$$

This function obtains **zero training loss**, but **large testing loss**.

# Overfitting



Flexible  
model

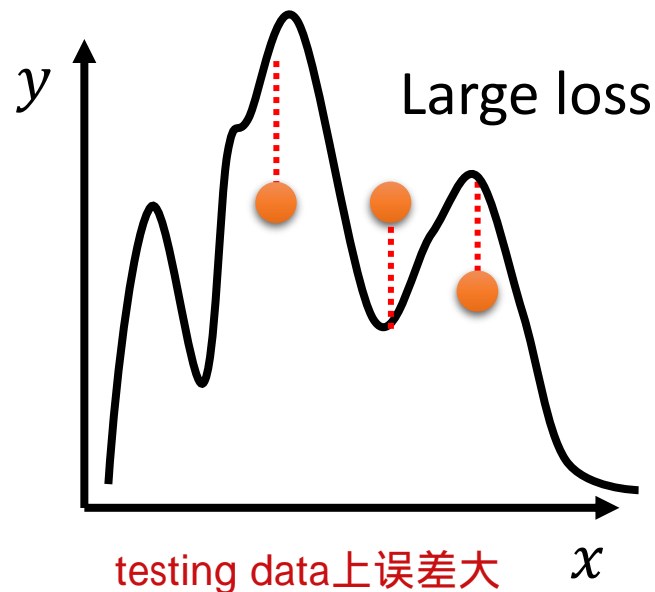


在没有training data限制的地方，就会有freestyle

---- Real data distribution  
(not observable)

● Training data

● Testing data

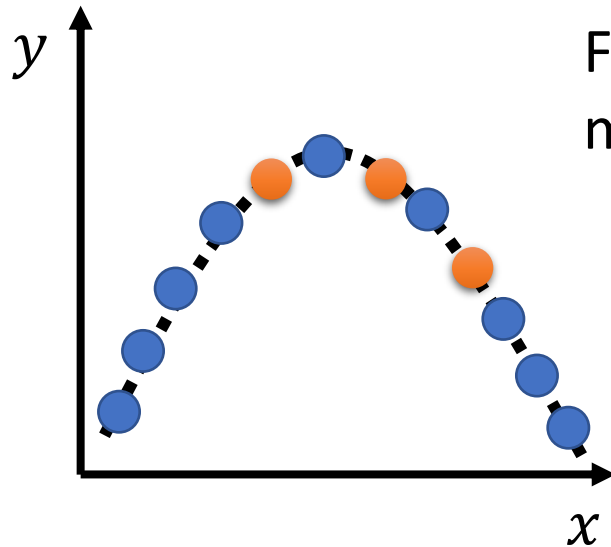


testing data上误差大

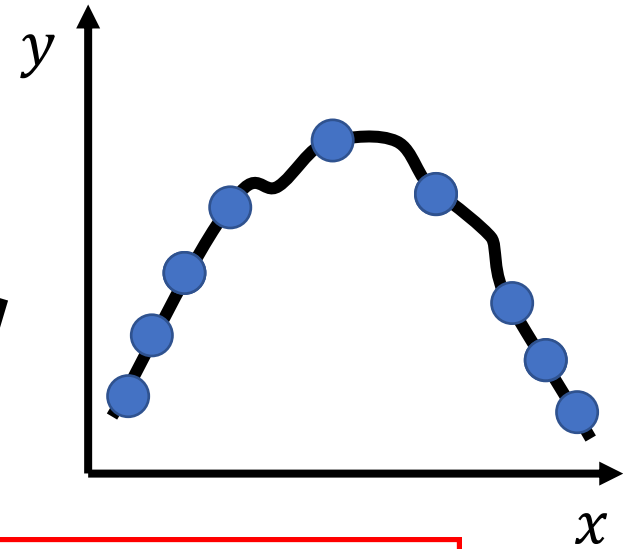
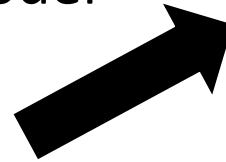
如果坚持想用flexible model的话，  
只能在数据上做文章了。

1. 更多数据
2. 数据增强

# Overfitting



Flexible  
model



1  
**More training data**

(cannot do it in HWs)

2

数据增强

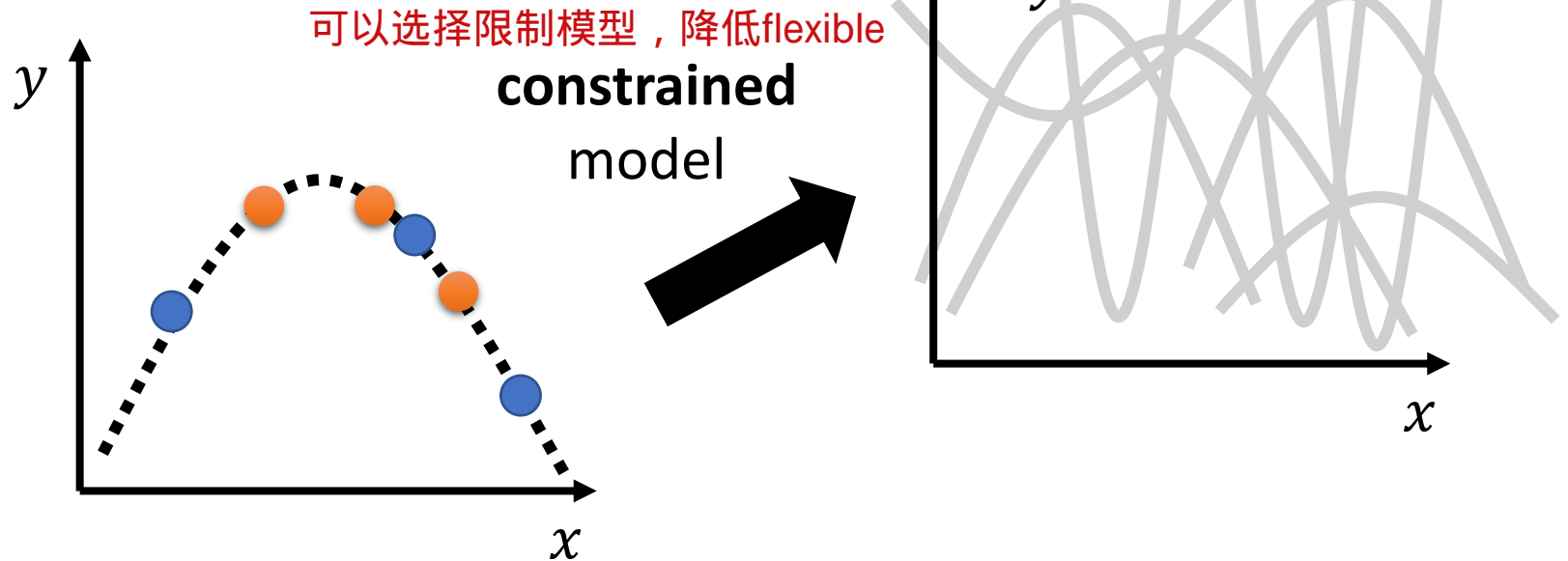
**Data augmentation**

(you can do that in HWs)



data augmentation要合适合理

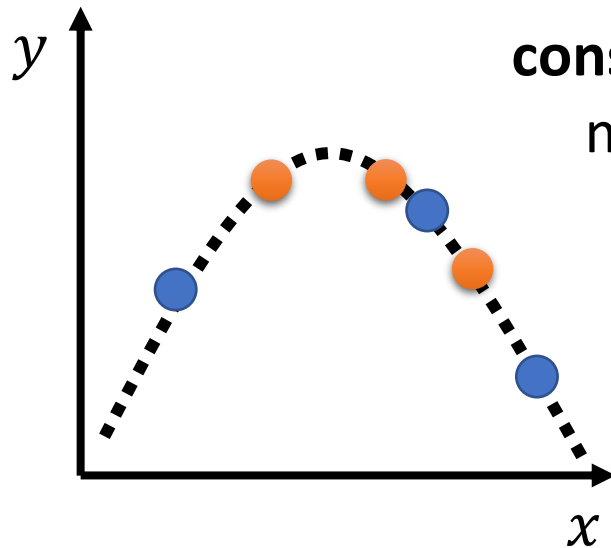
# Overfitting



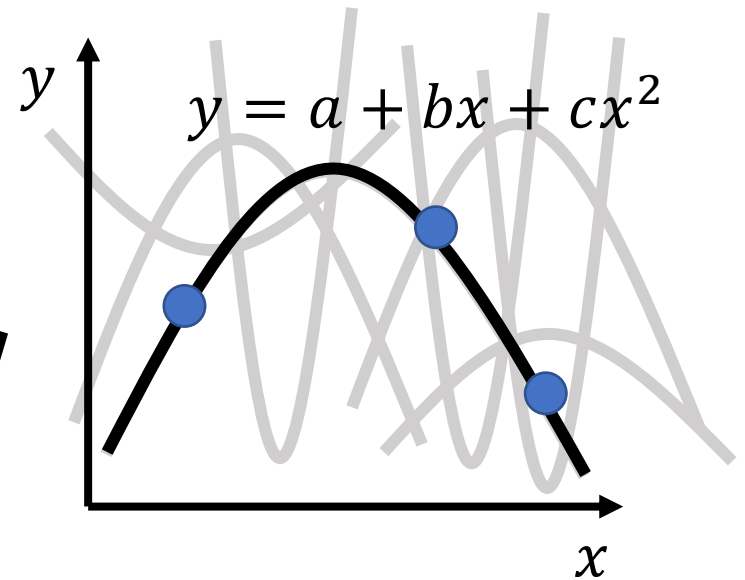
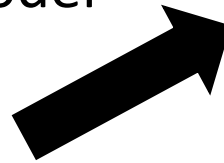
- Real data distribution (not observable)
- Training data
- Testing data



# Overfitting



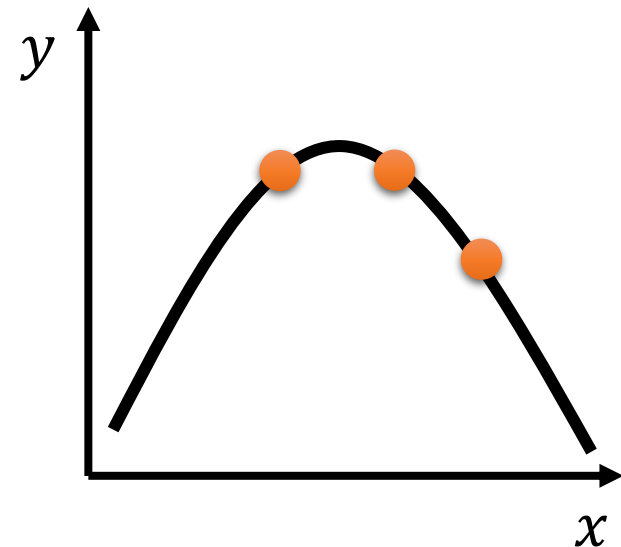
**constrained**  
model



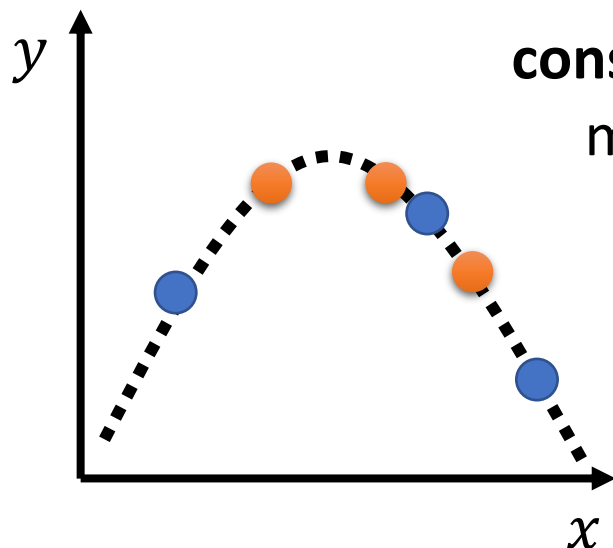
---- Real data distribution  
(not observable)

● Training data

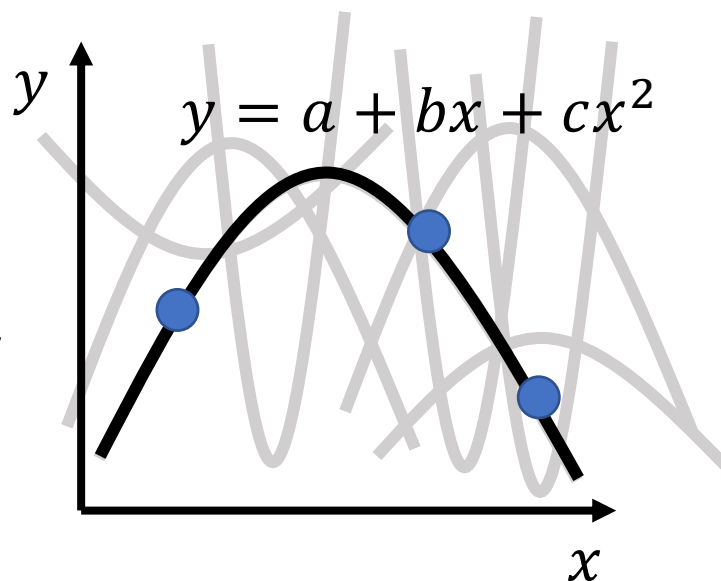
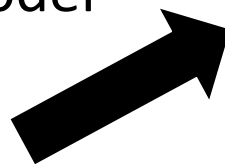
● Testing data



# Overfitting



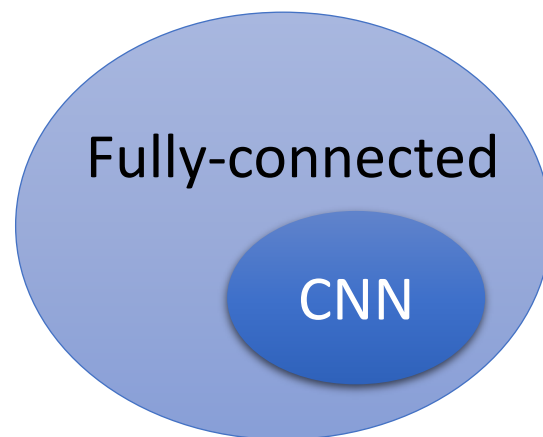
constrained  
model



限制模型的一些方法：

- Less parameters, sharing parameters
- Less features eg. 降维
- Early stopping
- Regularization
- Dropout

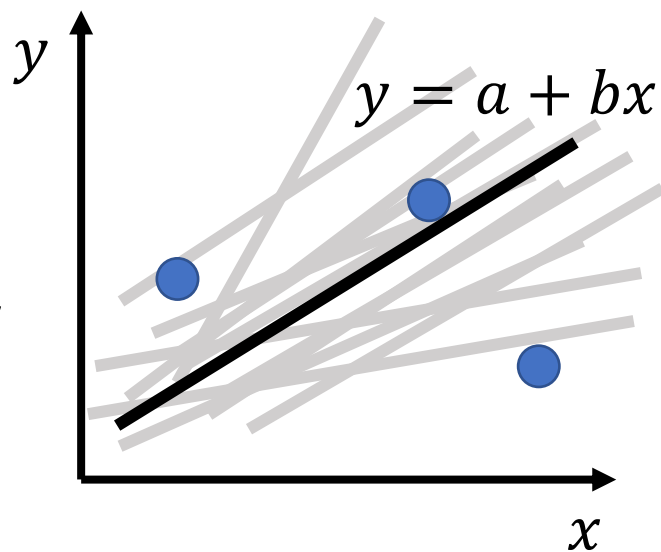
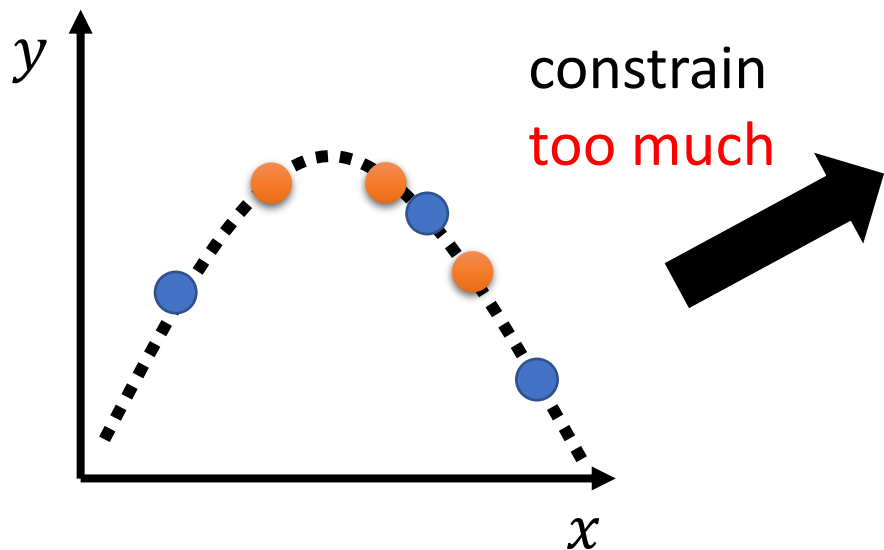
共享参数



CNN：针对图像的特性，来限制模型的弹性

如果模型的限制太大

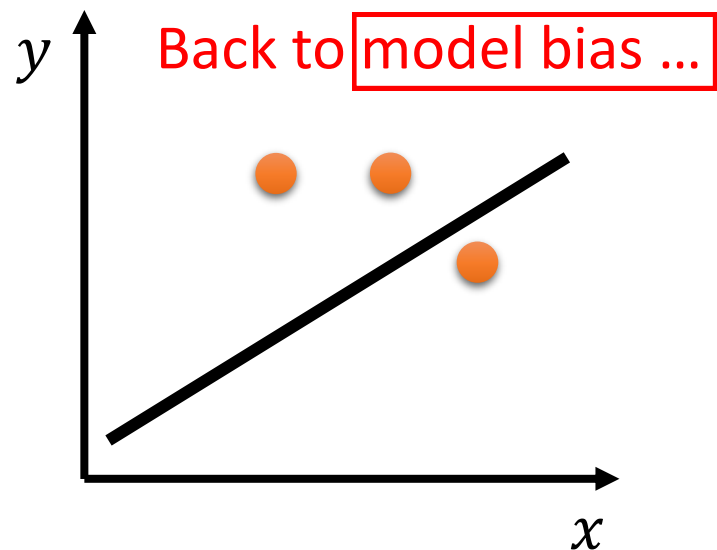
# Overfitting



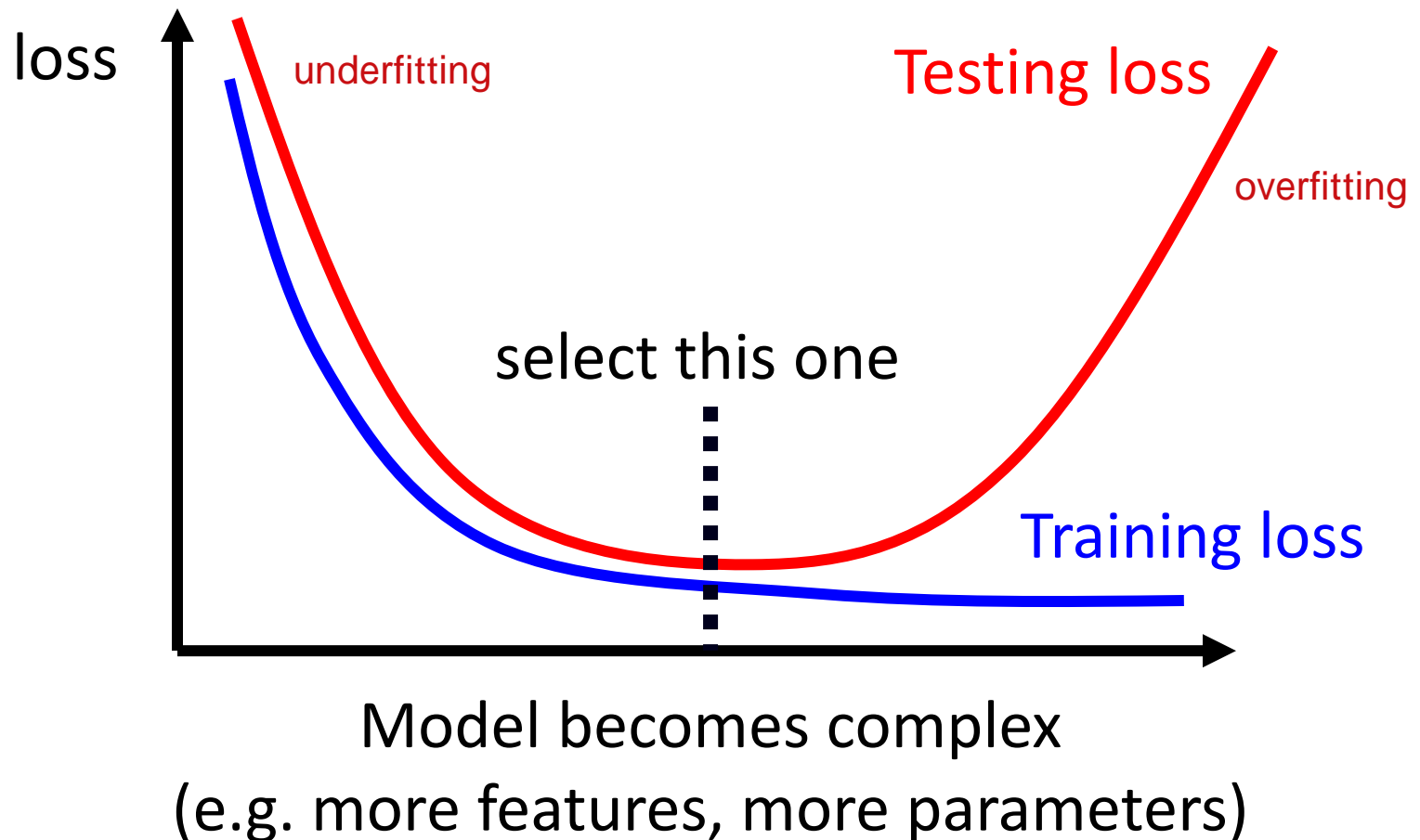
---- Real data distribution  
(not observable)

● Training data

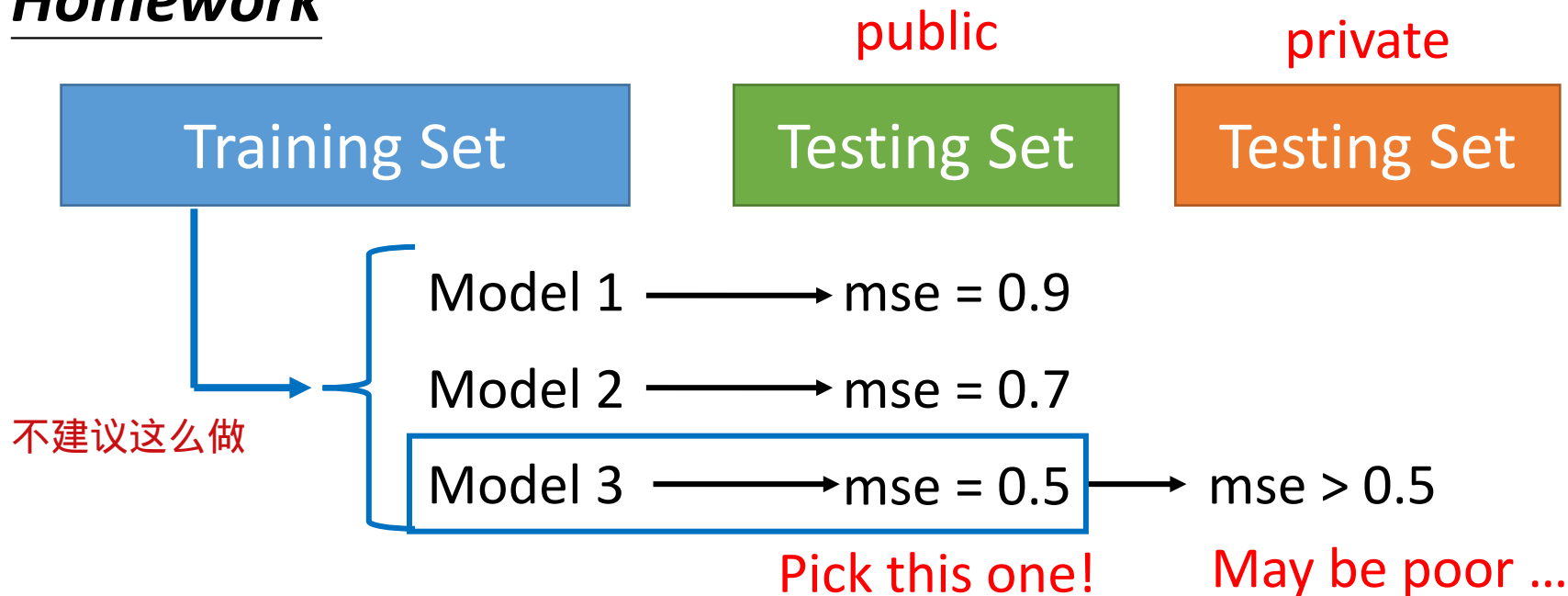
● Testing data



# Bias-Complexity Trade-off



# Homework



## The extreme example again

$$f_k(x) = \begin{cases} \hat{y}^i & \exists x^i = x \\ random & otherwise \end{cases} \quad k: 1 - 1000000000000000000000000$$

It is possible that  $f_{56789}(x)$  **happens** to get good performance on public testing set.

So you select  $f_{56789}(x)$  ..... Random on private testing set

# Homework

public

private

Training Set

Testing Set

Testing Set

Why?

Model 1  $\longrightarrow$  mse = 0.9

Model 2  $\longrightarrow$  mse = 0.7

Model 3  $\longrightarrow$  mse = 0.5  $\longrightarrow$  mse > 0.5

Pick this one!

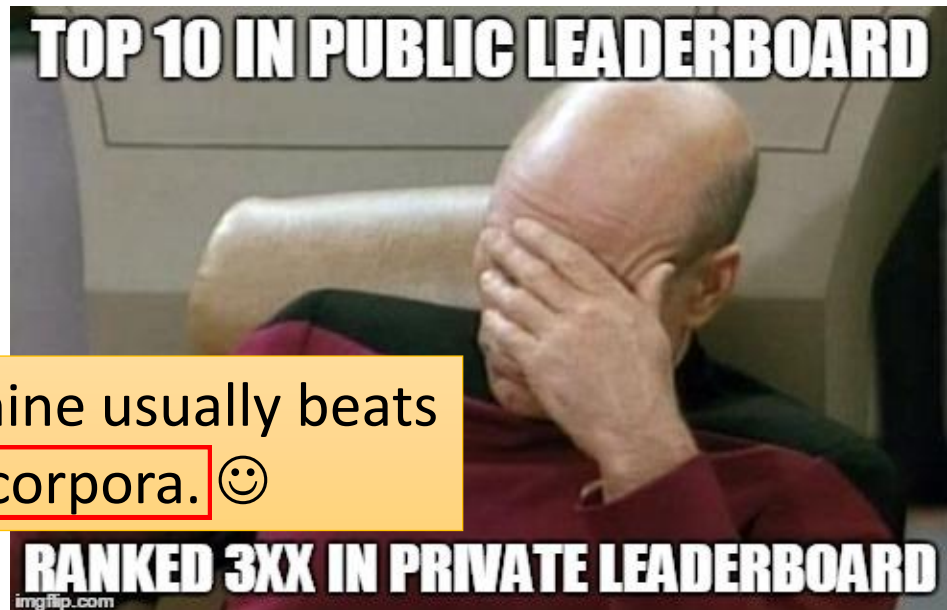
May be poor ...

What will happen?

<http://www.chioka.in/how-to-select-your-final-models-in-a-kaggle-competitio/>

This explains why machine usually beats human on benchmark corpora. 😊

因为这可以算作是public set



# Cross Validation (CV, 交叉验证)

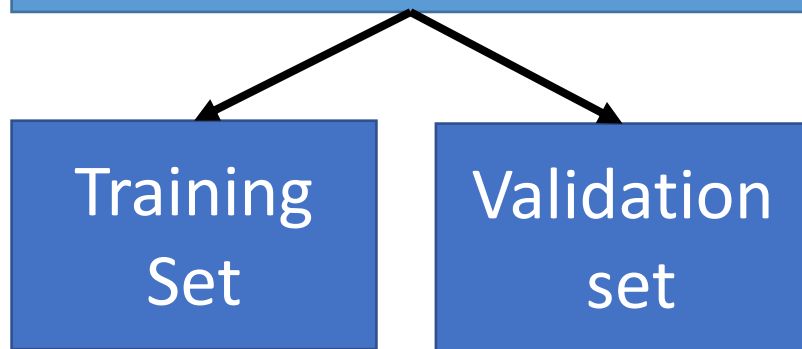
How to split?



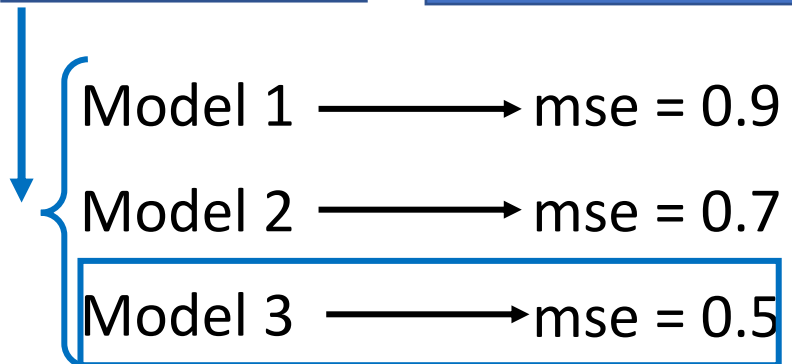
public



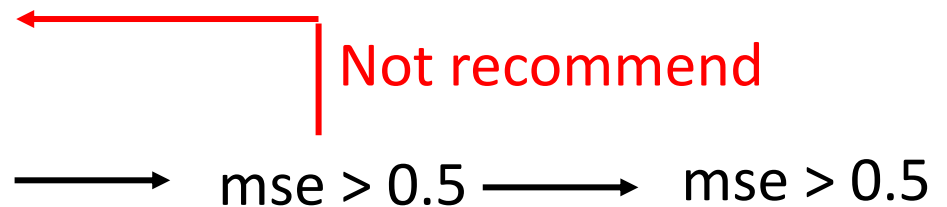
private



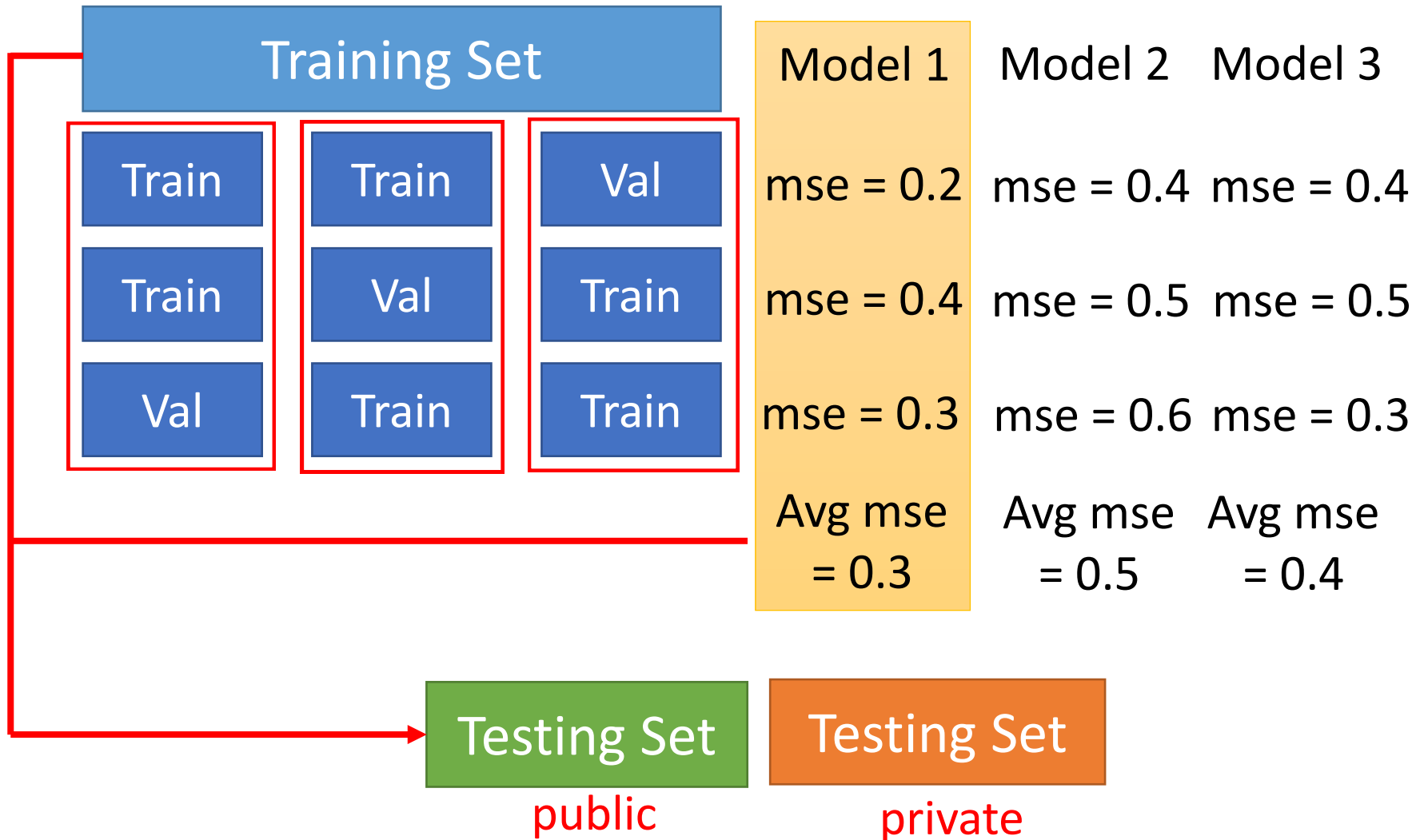
Using the results of public testing data to select your model  
You are making public set better than private set.



Not recommend

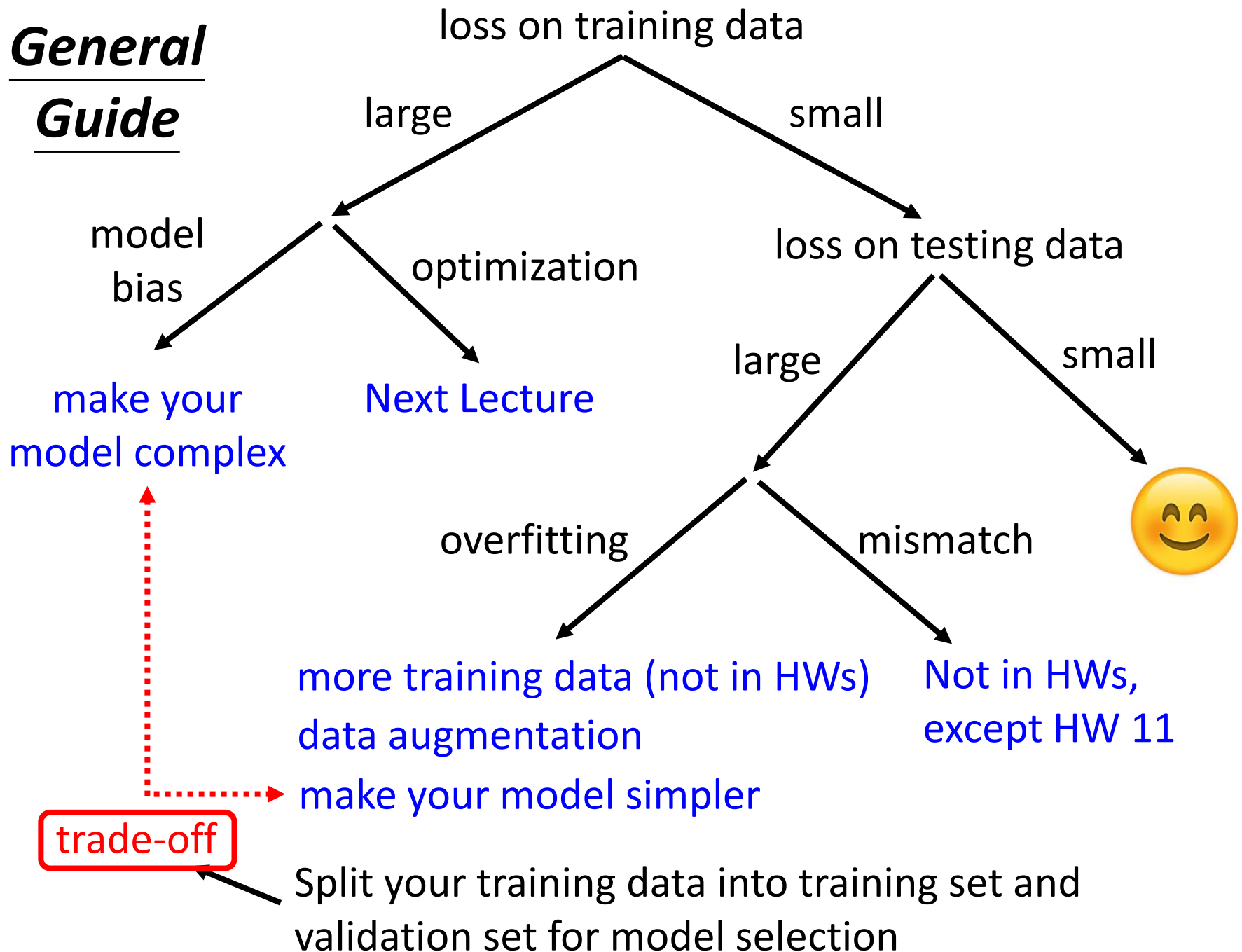


# N-fold Cross Validation K折交叉验证

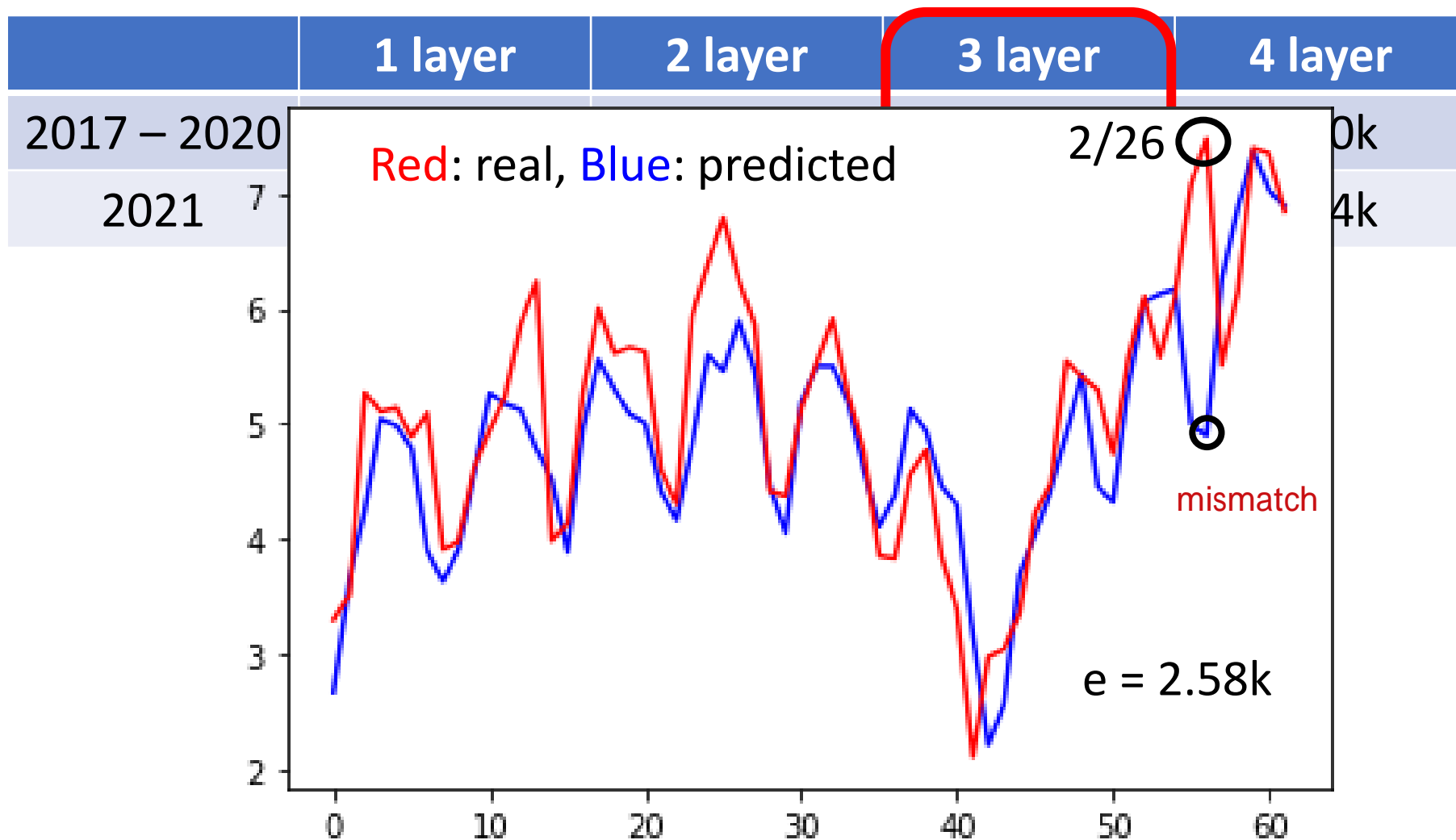




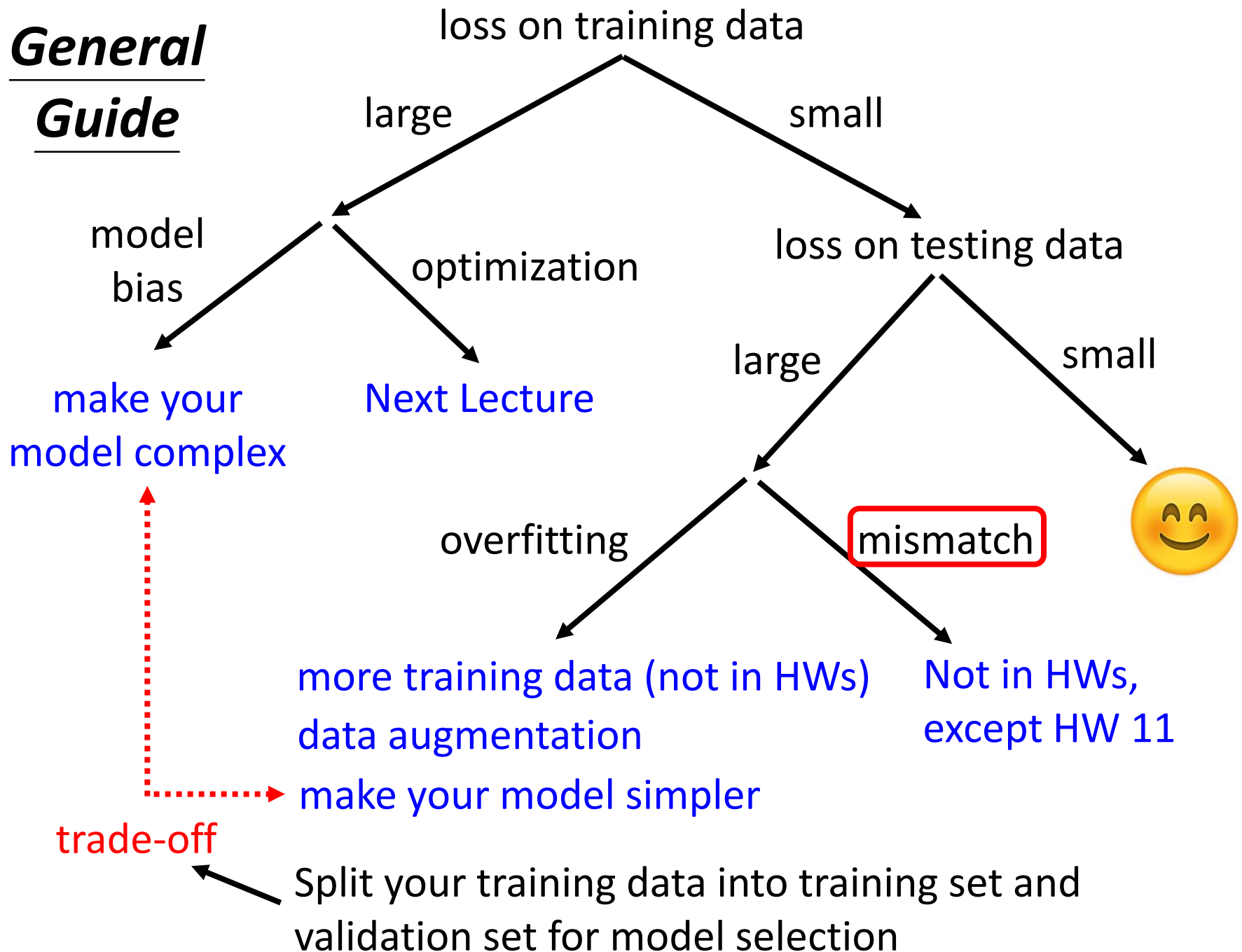
# General Guide



# Let's predict no. of views of 2/26!



# General Guide



# Mismatch

- Your training and testing data have different distributions. Be aware of how data is generated.

**Most HWs do not have this problem, except HW11**

## Training Data

horse



bed



clock



apple



cat



plane



television



dog



dolphin



spider



Simply increasing the training data will not help.

## Testing Data



# General Guide

