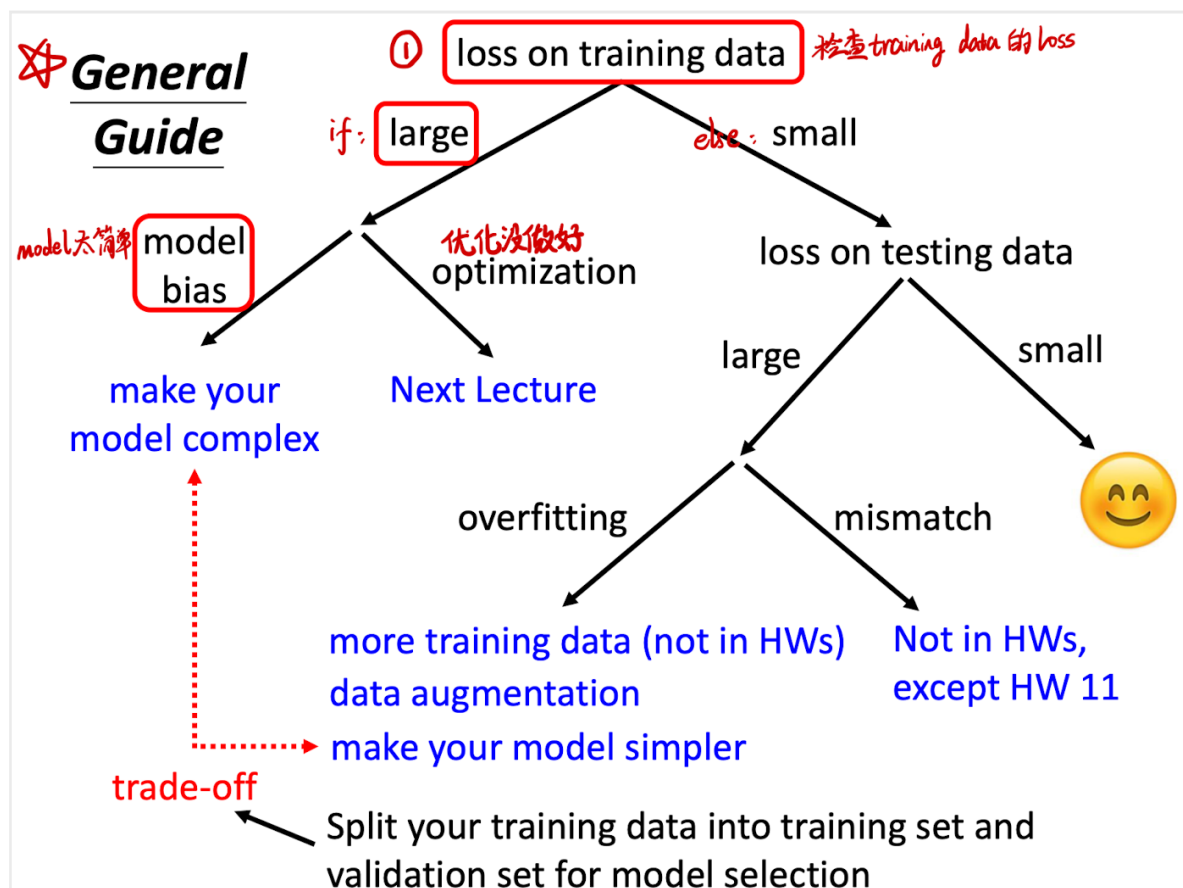


## 2.1 General Guidance

- model bias
- Model Bias v.s. Optimization Issue
- Overfitting
- Bias-Complexity Trade-off
- cross Validation



### model bias

模型过于简单

- 增加特征
- 复杂的神经网络

- The model is too simple.

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

too small ...

$f_{\theta^1}(x)$   $y = f_{\theta}(x)$   
 $f_{\theta^2}(x)$   
 $f_{\theta^*}(x)$   $f^*(x)$  small loss

- Solution: redesign your model to make it more flexible

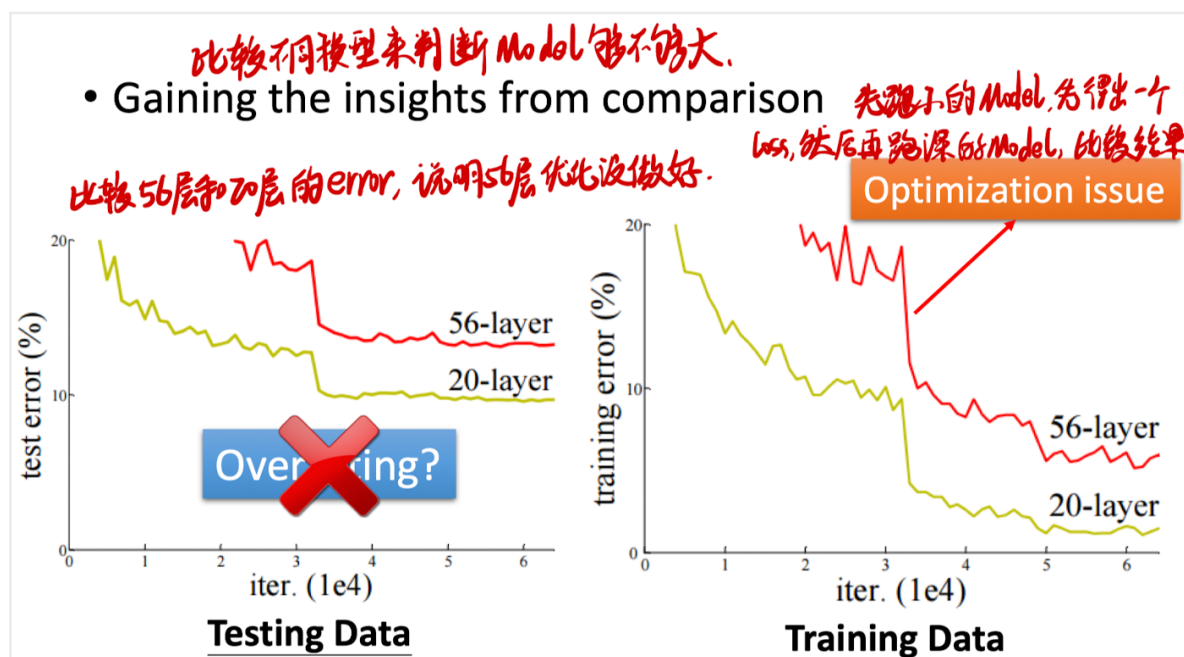
① More features  $y = b + wx_1 \rightarrow y = b + \sum_{j=1}^{56} w_j x_j$

② Deep Learning (more neurons, layers)

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

## Model Bias v.s. Optimization Issue

20层的model在training data上跑的错误率比56层的错误率低，说明56层的优化没做好



- 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.

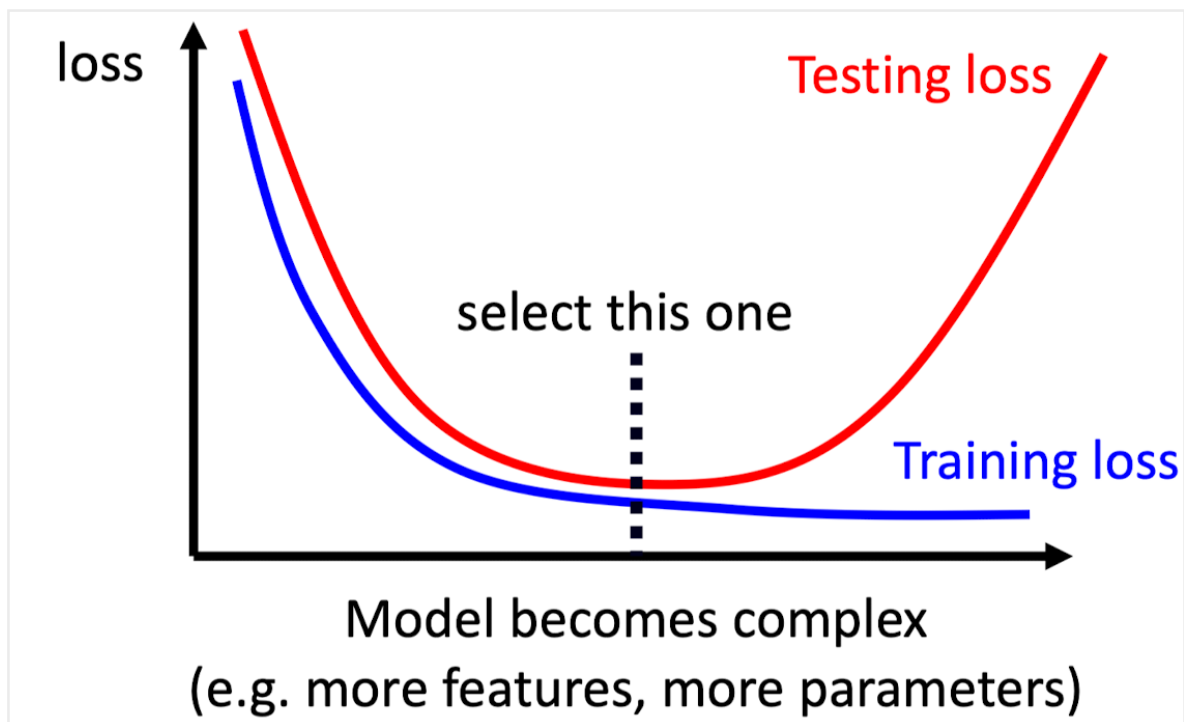
	1 layer	2 layer	3 layer	4 layer	5 layer
2017 – 2020	0.28k	0.18k	0.14k	0.10k	0.34k

- Solution: More powerful optimization technology (next lecture)

## Overfitting

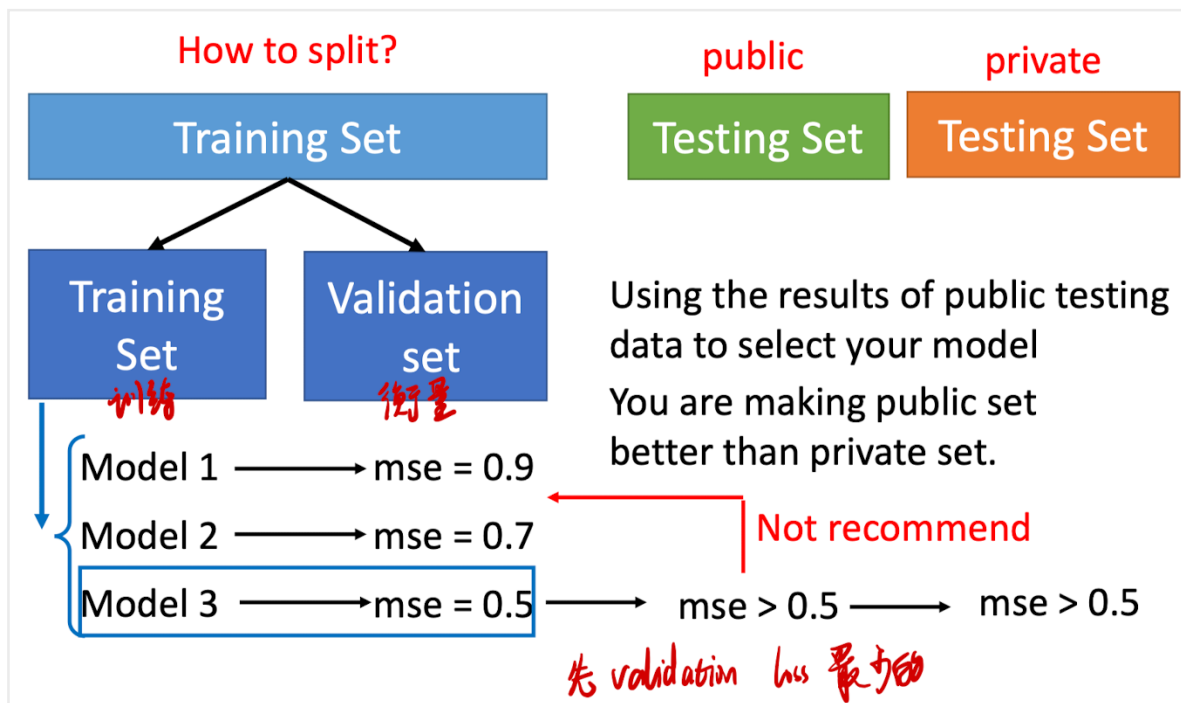
- 增加训练数据
  - data augmentation (比如左右翻转图像，具体取决于对资料的理解)
- 给模型加限制 (constrained model)
  - Less Parameters, sharing parameters
  - Less features
  - Early stopping
  - Regularization
  - Dropout

## Bias-Complexity Trade-off



### Cross Validation

合理的选择 Model 的方法：把 training 资料分成 training set 和 validation set，根据 validation set 的结果来挑选 model，然后再用 test set 测试结果。



把训练资料切成  $N$  等份，拿一份当 validation set，剩下  $n-1$  份当 training set，重复  $n$  次。把  $m$  个模型在这  $n$  个 setting 下都跑一次，然后把  $n$  次的结果平均起来，看谁的结果最好。

# N-fold Cross Validation

切训练数据为几等份, 其中一份  
为 validation SET, 另两份为 training set,  
重复3次。

