# Overfitting and Model Tuning

## Overfitting and Model Tuning

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## Reading Materials

► Max Kuhn. Chapter 4.

#### Prediction Problem

Given data of  $X = [X_1, X_2, ..., X_d]$  and Y. Find the relation between X and Y.

## Prediction Problem - Examples

► One Input Variable X

X	Υ
13	4.0
6	3.5
14	3
10	3.9
7	2.7
12	3.8
1	1.5

How are X and Y related?

## Prediction Problem - Examples

► Multiple Input Variables

$X_1$	$X_2$	 X <sub>35</sub>	Y
1	-1	 2	Tree
2.1	0	 6	Not a Tree
3	0	 8	Tree

How are X and Y related?

#### Prediction Problem

- ▶ If *Y* is continous, we have a **regression** problem.
- ▶ If *Y* is categorical, we have a **classification** problem.
- ▶ If *Y* is binary, we have a **binary classification** problem.

## Prediction Problem - Examples

▶ This is a regression problem since *Y* is continuous.

4.0
3.5
3
3.9
2.7
3.8
1.5

#### Prediction Problem - Examples

▶ This is a binary classification Problem since *Y* is binary.

$X_1$	$X_2$	 X <sub>35</sub>	Y
1	-1	 2	Tree
2.1	0	 6	Not a Tree
3	0	 8	Tree

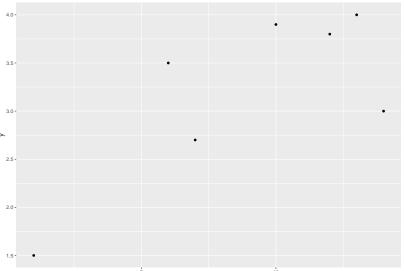
#### Overfitting

Consider the data:

Χ	Υ
13	4.0
6	3.5
14	3
10	3.9
7	2.7
12	3.8
1	1.5

- ▶ We will fit these data by polynomial model.
- ightharpoonup In polynomial model, Y is a polynomial function of X.

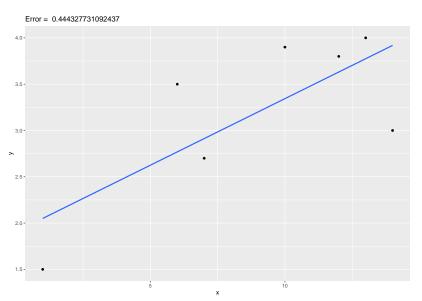
#### Overfitting



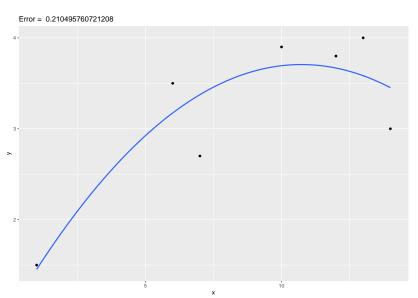
- ▶ We will fit these data by **polynomial model**.
- $\triangleright$  In polynomial model, Y is a polynomial function of X.

- ▶ In polynomial model, we need to specify the degree of the polynomial, *n*. Let try a few.
- ▶ If n = 1, we have a familiar **linear model**.
- Question: Does increasing n resuls in a better model?

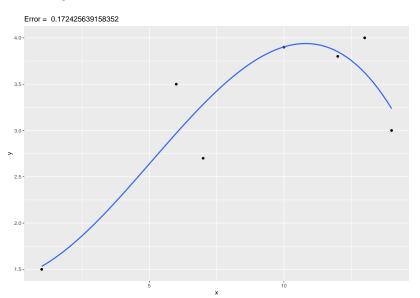
ightharpoonup n = 1.



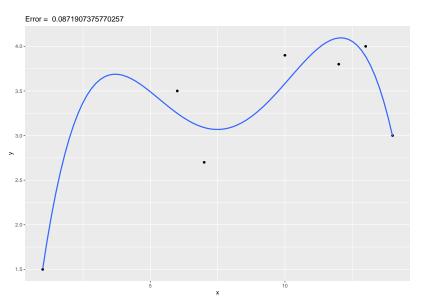
ightharpoonup n=2.



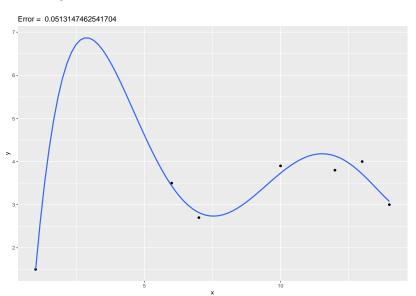
ightharpoonup n = 3.



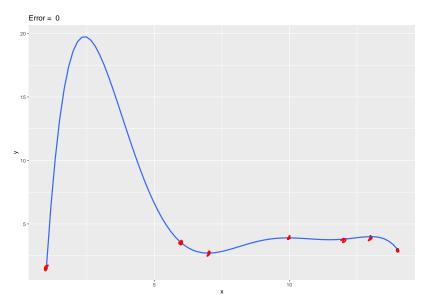
ightharpoonup n = 4.



ightharpoonup n=5.



n = 6.



**Question**: What are the errors when n > 6?

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- ► **Answer**: The errors are all zeros (There are actually many solutions for each degree greater than 6.)

▶ Question: What is the best model?

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- ► Answer We do not know. We need a validation dataset to validate the models.

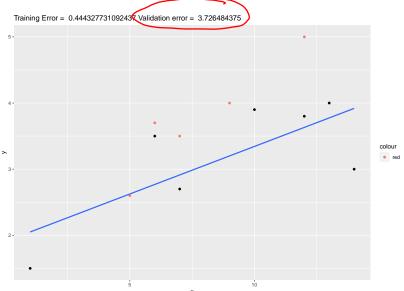
► The errors we have seen are called training errors

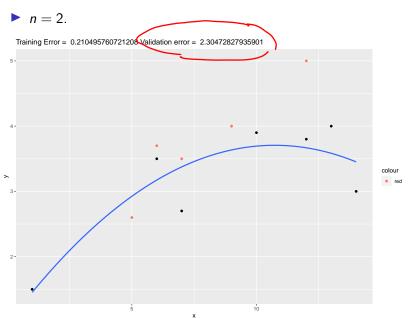
(errors on training data)

- Let's validate these models with a validation dataset
- ► Validation Data

X	Υ
5	2.6
7	3.5
9	4.0
6	3.7
12	5.0

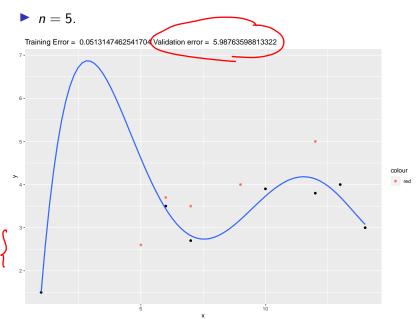
ightharpoonup n = 1.

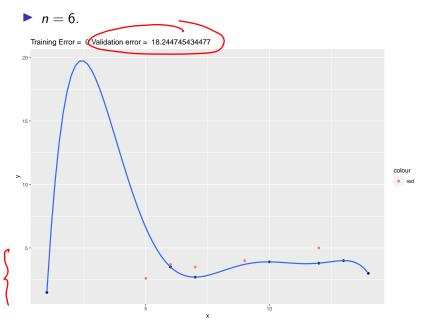




ightharpoonup n = 3.Training Error = 0.172425639158352 Validation error = 1.95519092637722 5 colour > 3 -2 -10

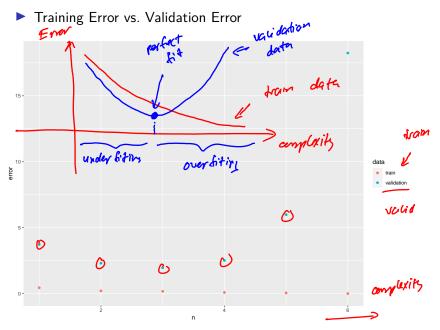
n = 4.Training Error = 0.0871907375770257 Validation error = 2.51566106871297 5 colour 3 -2 -10





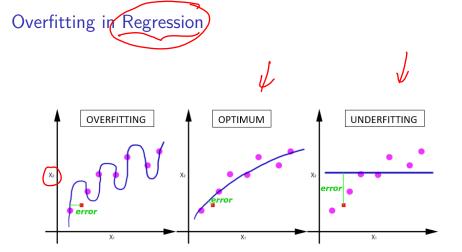
Training Error vs. Validation Error

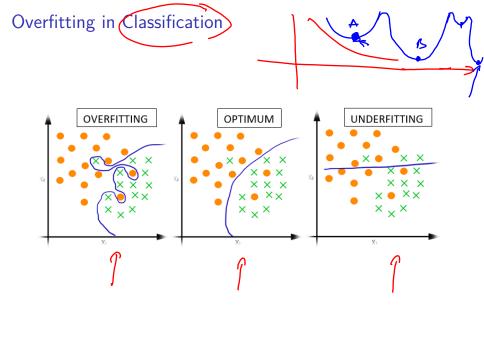
	Training Error	Validation Error
$\overline{n=1}$	0.4443277	3.726484
n = 2	0.2104958	2.304728
n = 3	0.1724256	1.955191 (Best!)
n = 4	0.08719074	2.515661
n = 5	0.05131475	5.987636
n = 6	0	18.24475



- ▶ As the degree *n* increases, the training errors decrease
- Model 6 (n = 6) is the best (perfect) in training but the worst in validation.
- The best model is the model with the best (lowest) error in validation data.

Overfitting - Polynomial Model ## - Model 4, 5 and 6 are **overfitted** - Model 1 and 2 are **underfitted** - Model 3 is the best model.





# Model Complexity Capacity

- ► In polynomial models, the larger *n*, the more complex/capable the model.
- Model complexity can be measured by the number of parameters/unknown of the model.

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Linear model:

$$y = \partial x + b$$

▶ Question: How many unknowns/parameters in linear model?

Linear model (n = 1):

$$y = ax + b$$

- Question: How many unknowns/parameters in linear model?
- **Answer**: Two unknowns/paramters:  $\hat{b}$  and  $\hat{b}$

3 UNE ROVORS

• Quadratic model (n = 2):

$$y = (b)x + (b)x$$

- ► Three unknowns: a, b, and c.
- Quadratic model has more unknowns/parameters then linear model. Thus, quadratic model is more complex than linear model



more

► The mode complex the model, the easier it becomes overfitted!



Model Tuning

- We just "tuned" the the parameter n.
- ► The parameter *n* is called **tuning parameter**, or **hyperparameter**

Turing pormets

### Model Tuning

- Model tuning is the process of finding the **best** values for the tuning parameters of the model
- This is done through **trying out** many values for the tuning parameters then select the best values.

- Model training is the process of finding the unknown/parameters of the model
- **Example**: Training linear mode y = ax + b is to find a and b that best fit the data

### Model Training vs. Model Tuning

- ► Model training finds the **parameters**
- ► Model tuning finds **hyperparameters**

## Model vs. Family of Models

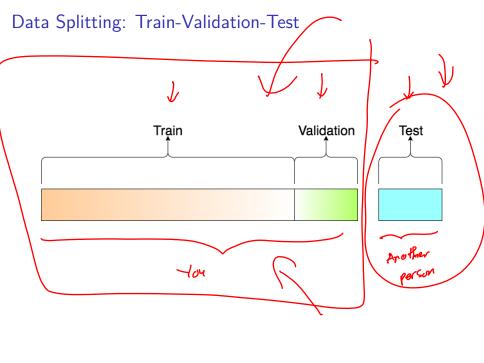
- The polynomial model is a family of models.
- Linear model is just **one** model
- A family of models has "tuning parameters".
- A single model, say, linear model, does not have tuning parameter
- Some model has multiple tuning parameters.

## Data Splitting

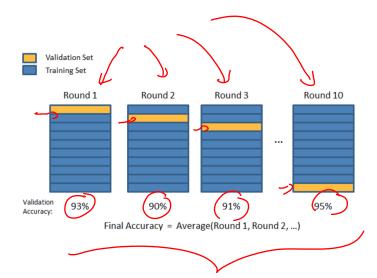
- ► We need validation data for model tuning.
  - ▶ Question: How can one of obtain validation data?

### Data Splitting

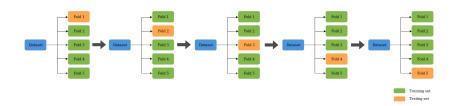
- We need validation data for model tuning.
- Question: How can one of obtain validation data?
- ► Answer: We do not use the entire data to train models. We use a portion of it for training and save data for validation and testing.



# k-folds Cross Validation



### k-folds Cross Validation



### k-folds Cross Validation and test

