

Overfitting and Model Tuning

Overfitting and Model Tuning

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Reading Materials

- ▶ Max Kuhn. Chapter 4.

Prediction Problem

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

Prediction Problem - Examples

► One Input Variable X

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

label | target | response variable
output

continuous label

⇒ regression problem.

How are X and Y related? Given new X , we want to predict Y

Prediction Problem - Examples

► Multiple Input Variables

X_1	X_2	...	X_{35}	Y
1	-1	...	2	Tree
2.1	0	...	6	Not a Tree
3	0	...	8	Tree

label
categorical label \Rightarrow classification problem.

How are X and Y related?

Prediction Problem

- ▶ If Y is continuous, 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.

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

Prediction Problem - Examples

- This is a binary classification Problem since Y is binary.

X_1	X_2	...	X_{35}	Y
1	-1	...	2	Tree
2.1	0	...	6	Not a Tree
3	0	...	8	Tree

Overfitting

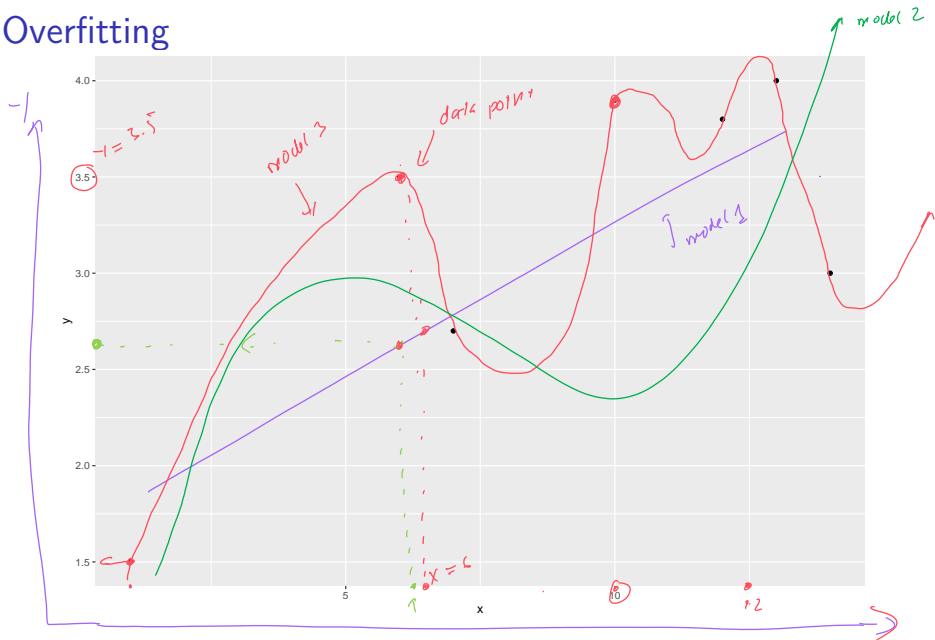
- Consider the data:

$x = 20 \rightarrow ?$

X	Y
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.
- In polynomial model, Y is a polynomial function of X .

Overfitting



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

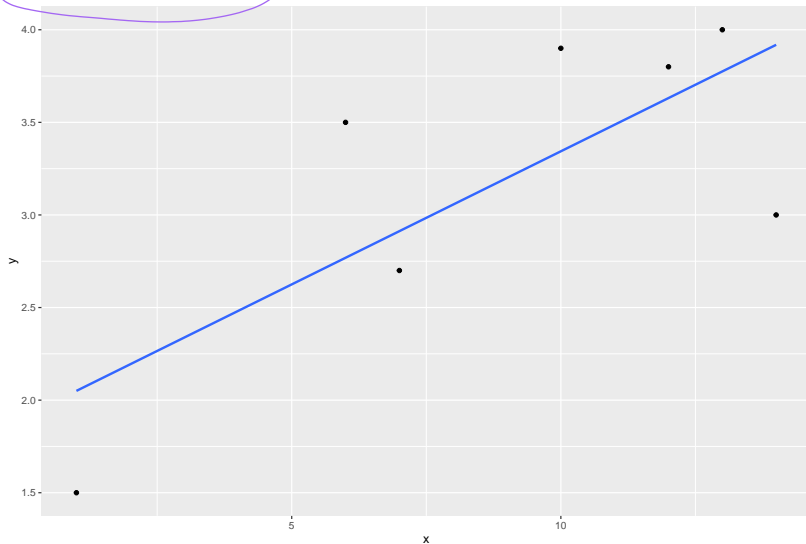
Overfitting - Polynomial Model

- ▶ 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 results in a better model?**

Overfitting - Polynomial Model

► $n = 1$.

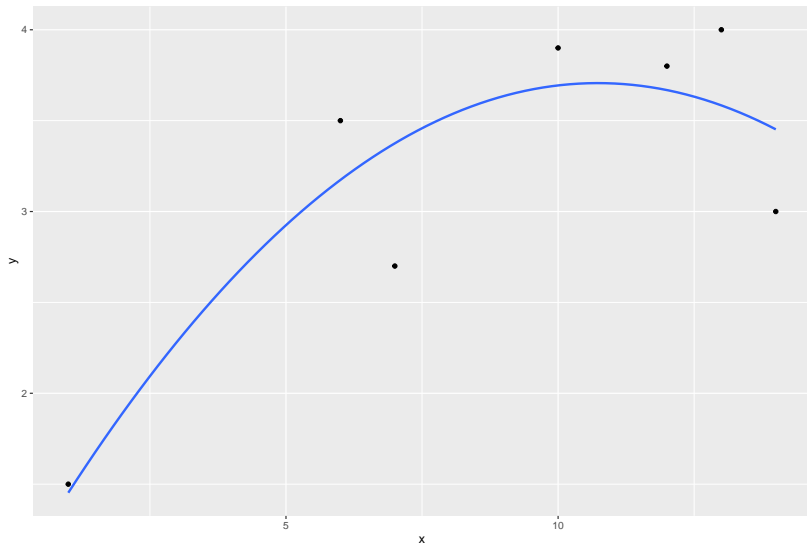
Error = 0.444327731092437



Overfitting - Polynomial Model

► $n = 2$.

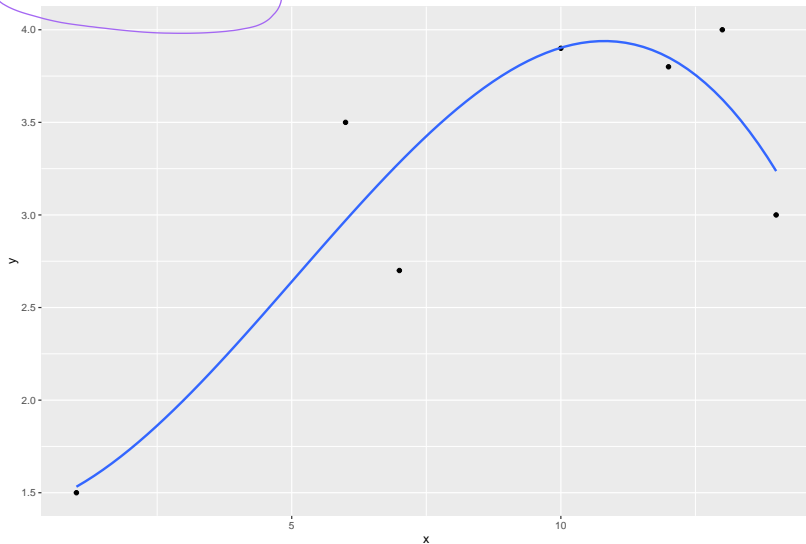
Error = 0.210495760721208



Overfitting - Polynomial Model

► $n = 3.$

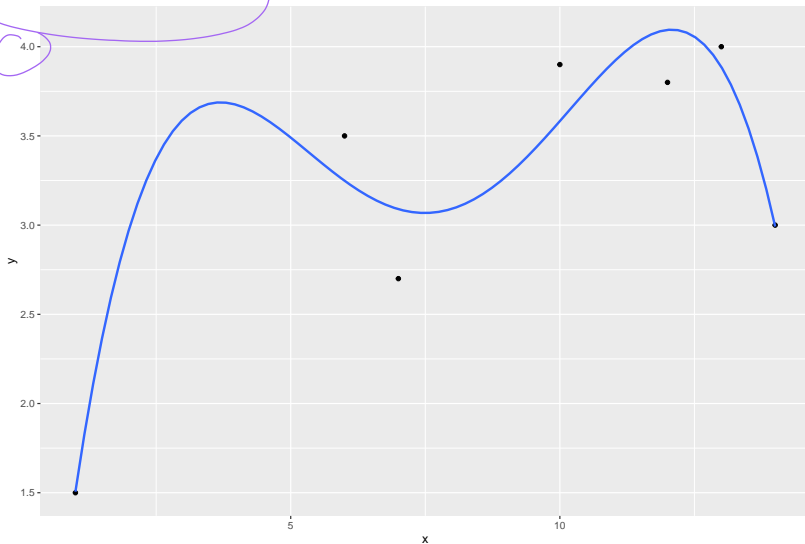
Error = 0.172425639158352



Overfitting - Polynomial Model

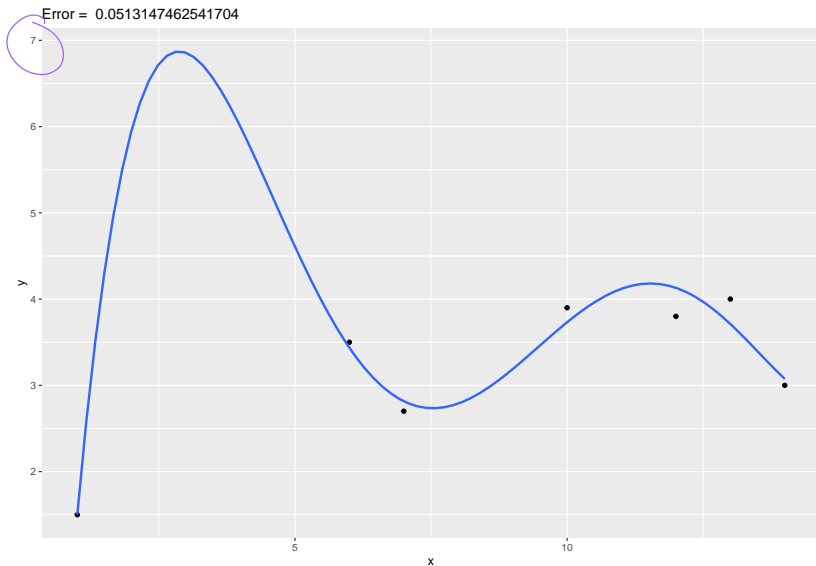
► $n = 4.$

Error = 0.0871907375770257



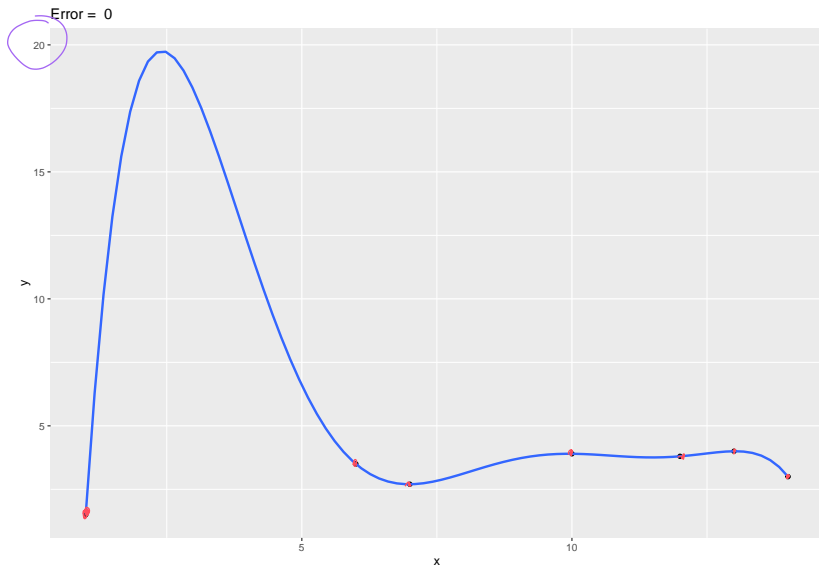
Overfitting - Polynomial Model

► $n = 5$.



Overfitting - Polynomial Model

► $n = 6$.



Overfitting - Polynomial Model

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

Overfitting - Polynomial Model

- ▶ **Question:** What are the errors when $n > 6$?
- ▶ **Answer:** The errors are all zeros. (There are actually many solutions for each degree greater than 6.)

Overfitting - Polynomial Model

- ▶ **Question:** What is the best model?

Overfitting - Polynomial Model

- ▶ **Question:** What is the best model?
- ▶ **Answer:** We do not know. We need a validation dataset to validate the models.

Overfitting - Polynomial Model

- ▶ The errors we have seen are called **training errors**

Overfitting - Polynomial Model

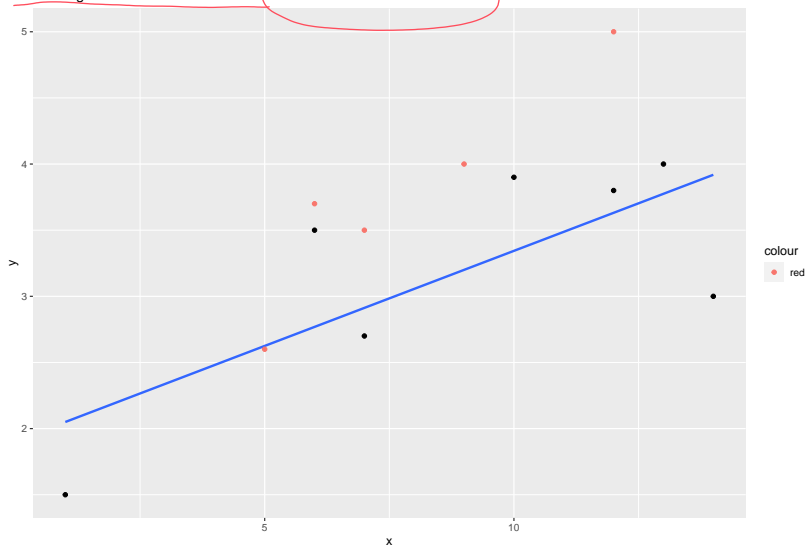
- ▶ Let's validate these models with a validation dataset
- ▶ Validation Data

X	Y
5	2.6
7	3.5
9	4.0
6	3.7
12	5.0

Overfitting - Polynomial Model

► $n = 1$.

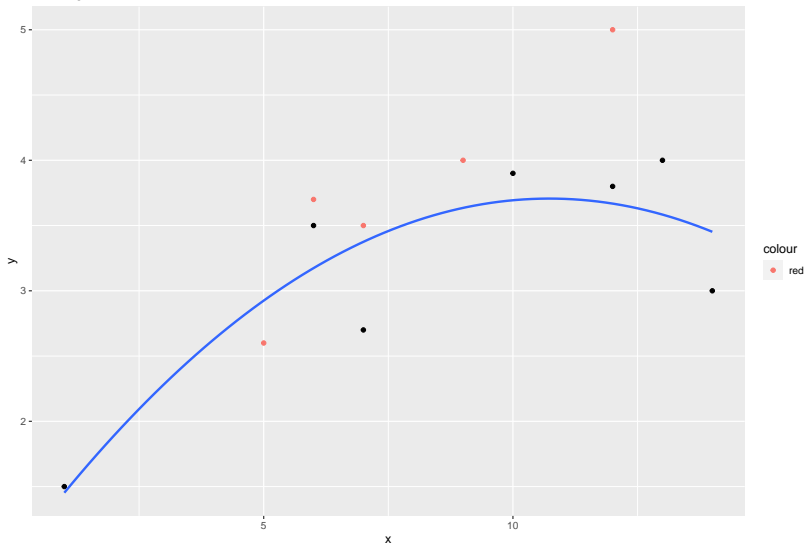
Training Error = 0.444327731092437 Validation error = 3.726484375



Overfitting - Polynomial Model

► $n = 2$.

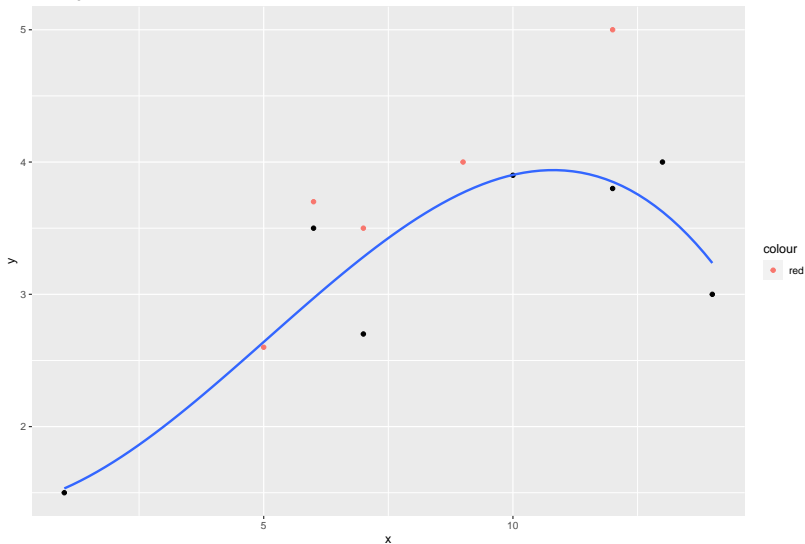
Training Error = 0.210495760721208 Validation error = 2.30472827935901



Overfitting - Polynomial Model

► $n = 3$.

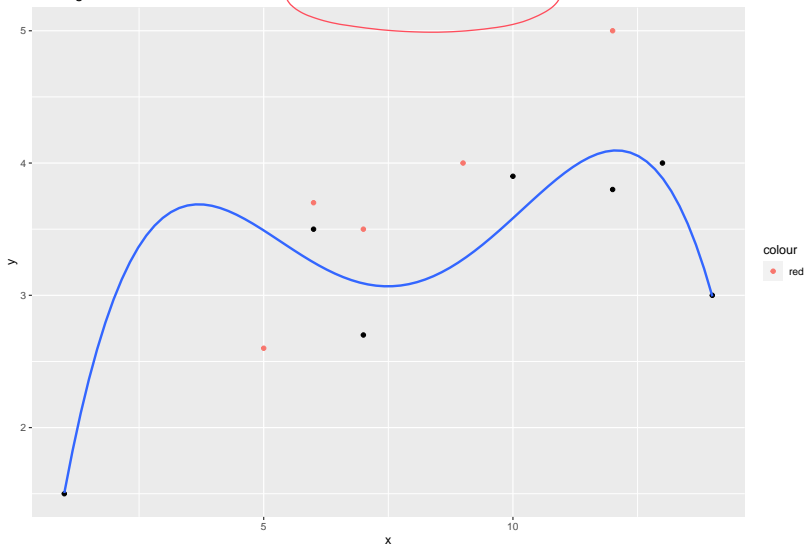
Training Error = 0.172425639158352 Validation error = 1.95519092637722



Overfitting - Polynomial Model

► $n = 4$.

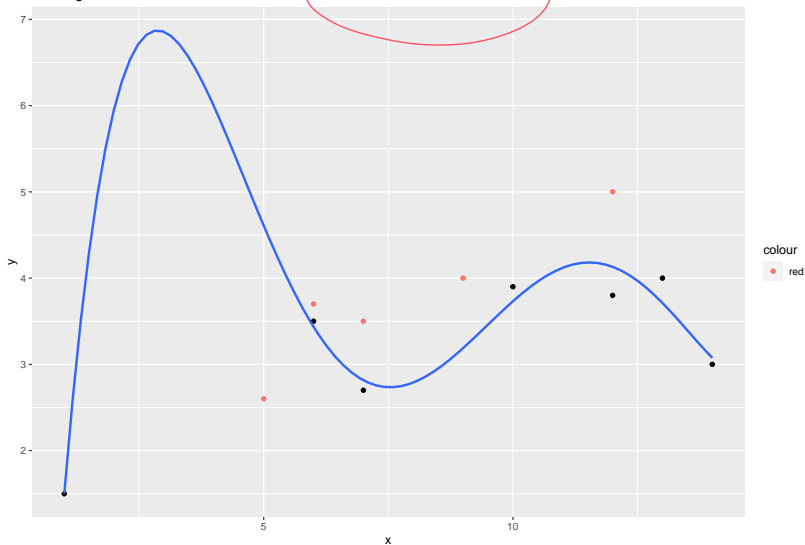
Training Error = 0.0871907375770257 Validation error = 2.51566106871297



Overfitting - Polynomial Model

► $n = 5$.

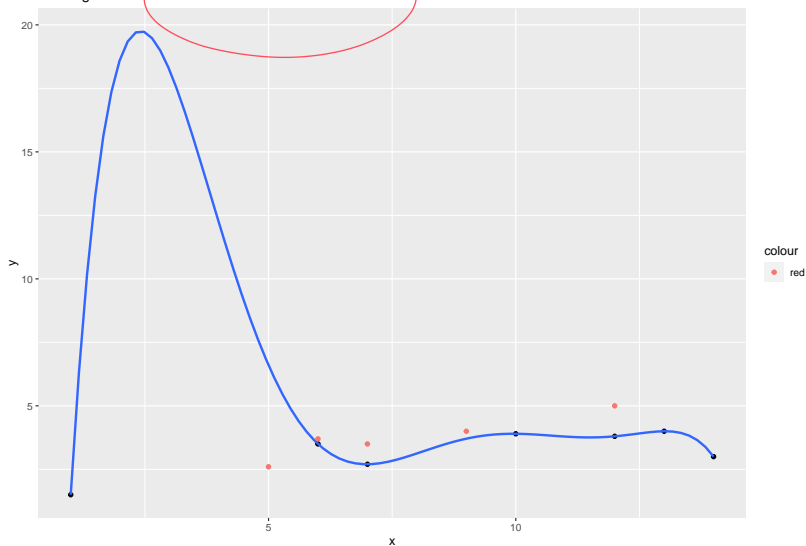
Training Error = 0.0513147462541704 Validation error = 5.98763598813322



Overfitting - Polynomial Model

► $n = 6$.

Training Error = 0 Validation error = 18.244745434477



Overfitting - Polynomial Model

► Training Error vs. Validation Error

	Training Error	Validation Error
$n = 1$	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

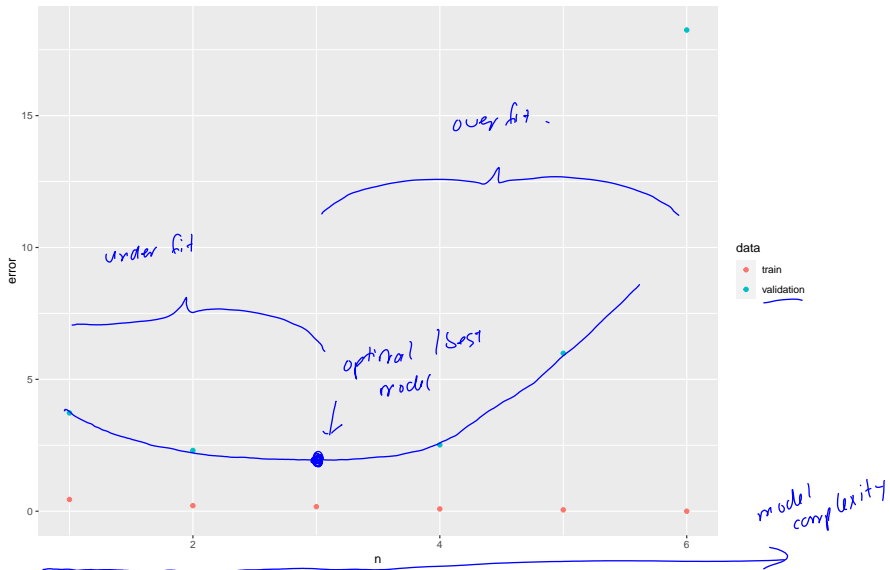
model
complexity /
flexibility



hyperparameter
tuning

Overfitting - Polynomial Model

► Training Error vs. Validation Error

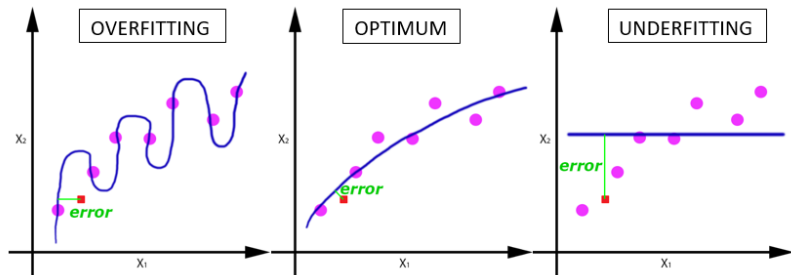


Overfitting - Polynomial Model

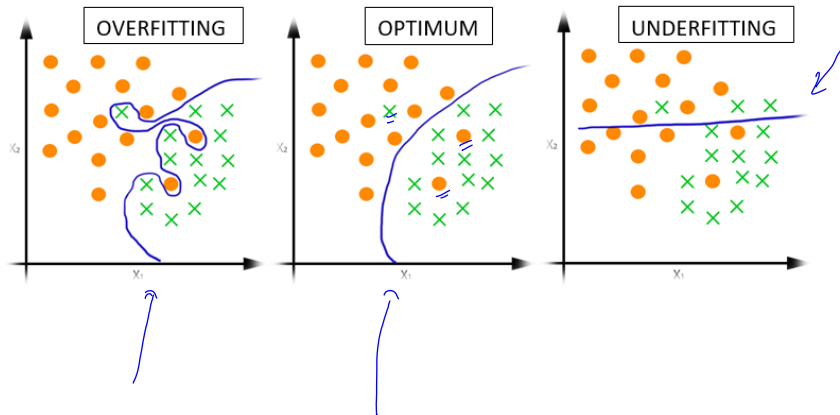
- ▶ As the degree n increases, the training errors decrease
- ▶ Model 6 ($n = 6$) is the best (perfect) in training but the worst in validation.
- ▶ The best model is the model with the best (lowest) error in **validation data**.

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

Overfitting in Regression



Overfitting in Classification



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.

Model Complexity/Capacity

- ▶ Linear model:

$$y = ax + b$$

unknowns

↓ ↓

y ax b

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

Model Complexity/Capacity

- ▶ Linear model ($n = 1$):

$$y = ax + b$$

- ▶ **Question:** How many unknowns/parameters in linear model?
- ▶ **Answer:** Two unknowns/parameters: a and b

Model Complexity/Capacity

- ▶ Quadratic model ($n = 2$):

$$y = ax^2 + bx + c$$

unknowns

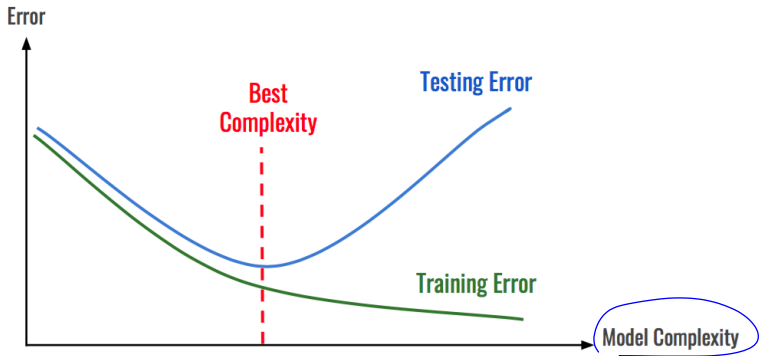
↓ ↓ ↓

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

Model Complexity/Capacity

- ▶ The more complex the model, the easier it becomes overfitted!

Model Complexity/Capacity



Model Tuning

- ▶ We just “tuned” the the parameter n .
- ▶ The parameter n is called **tuning parameter**, or **hyperparameter**

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

- ▶ Model training is the process of finding the unknown/parameters of the model
- ▶ **Example:** Training linear model $y = ax + b$ 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.

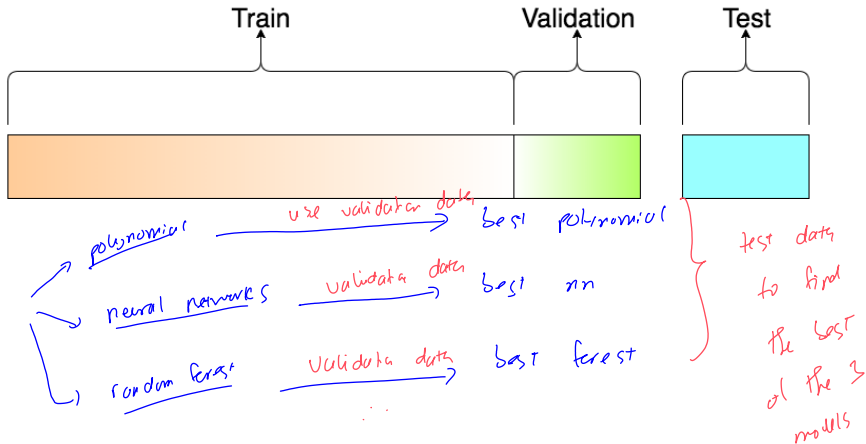
Data Splitting

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

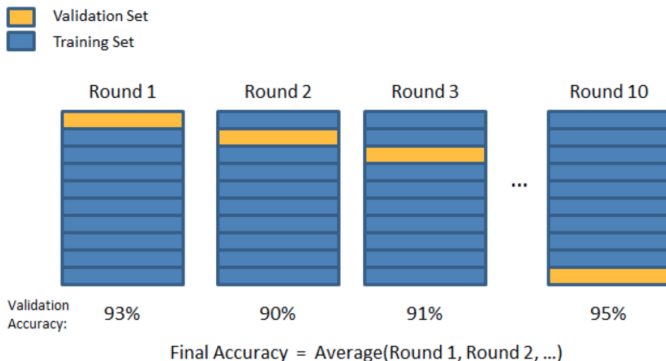
Data Splitting

- ▶ We need validation data for model tuning.
- ▶ **Question:** How can one 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.

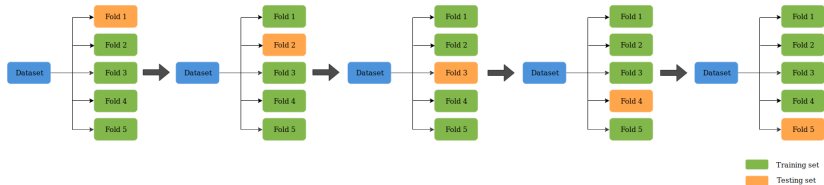
Data Splitting: Train-Validation-Test



k-folds Cross Validation



k-folds Cross Validation



k-folds Cross Validation and test

