

DEEP LEARNING



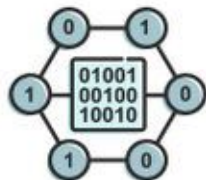
ALGORITHM



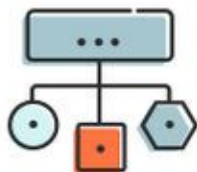
LEARNING



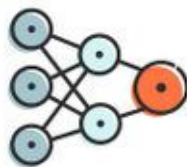
AI



DATA MINING



CLASSIFICATION



NEURAL NETWORKS



AUTONOMUS

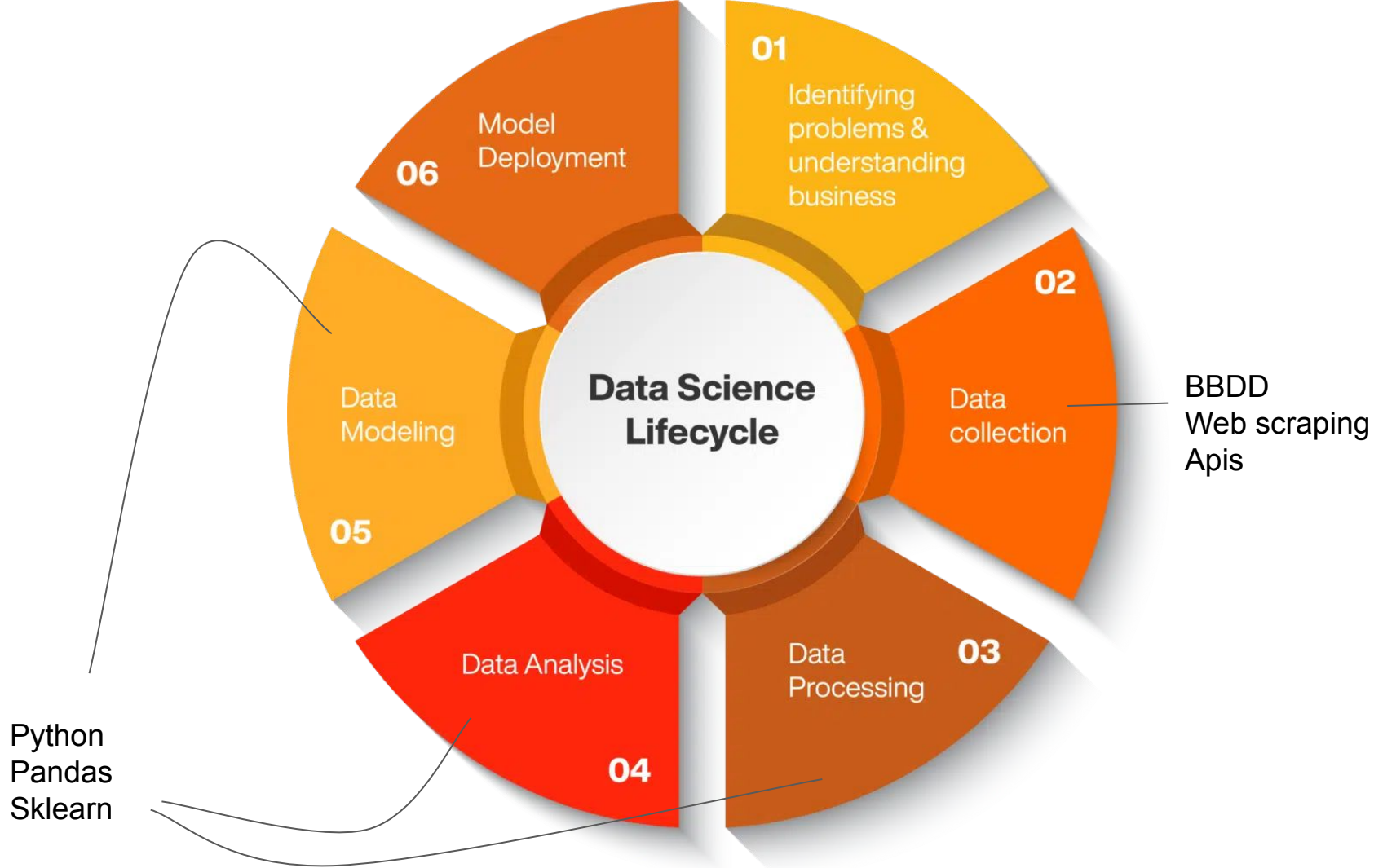


IMPROVES

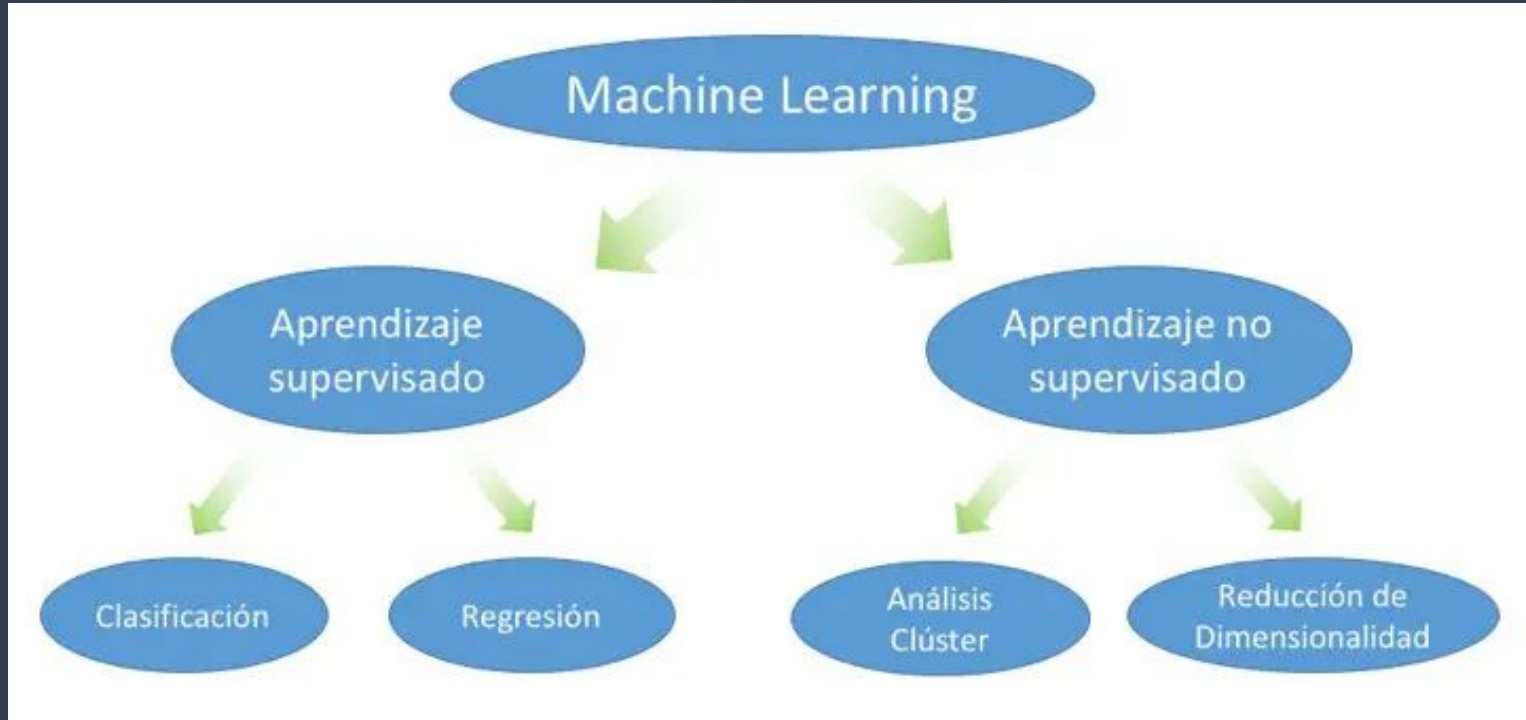


ANALYZE

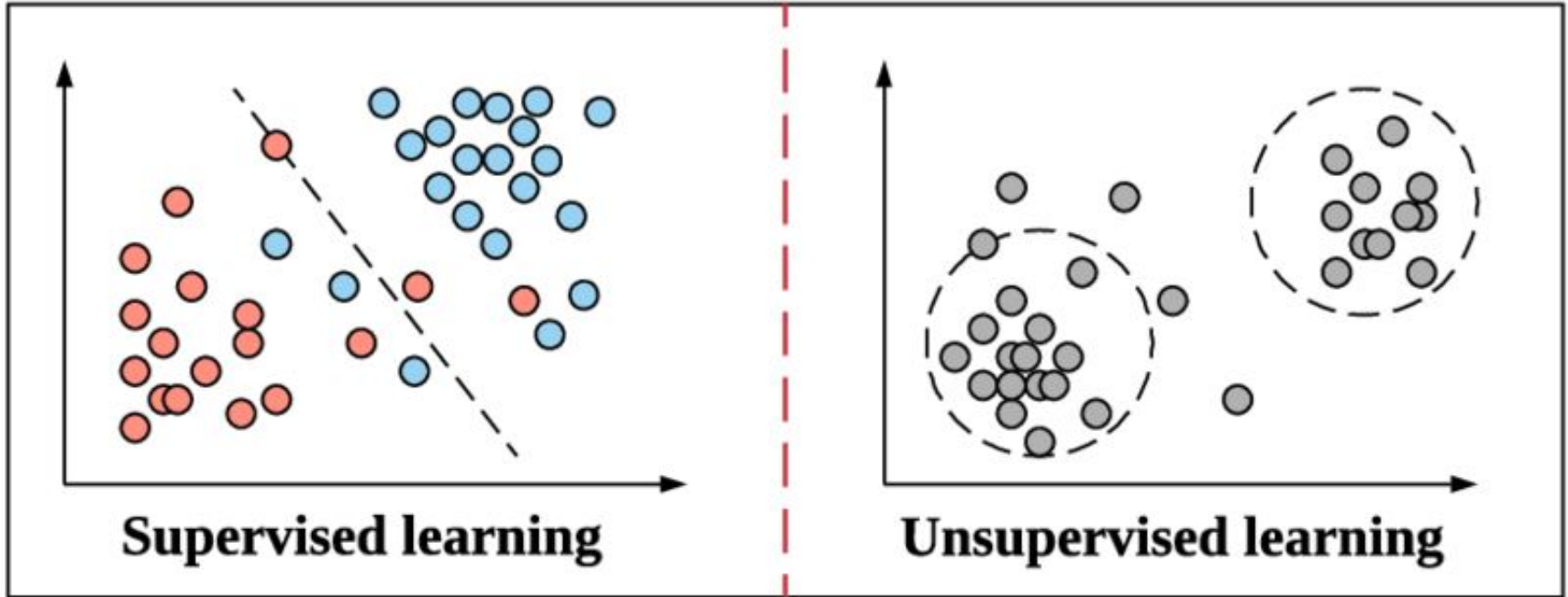
# MACHINE LEARNING



# Machine learning



# Types of algorithms



## supervised learning

Input data



Annotations

These are  
apples



Model



Prediction

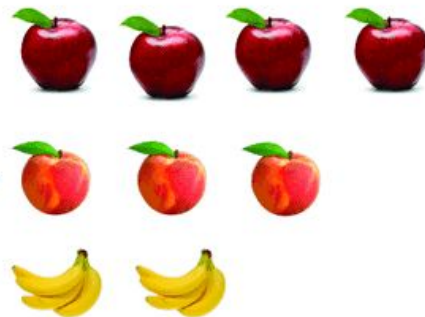
Its an  
apple!

## unsupervised learning

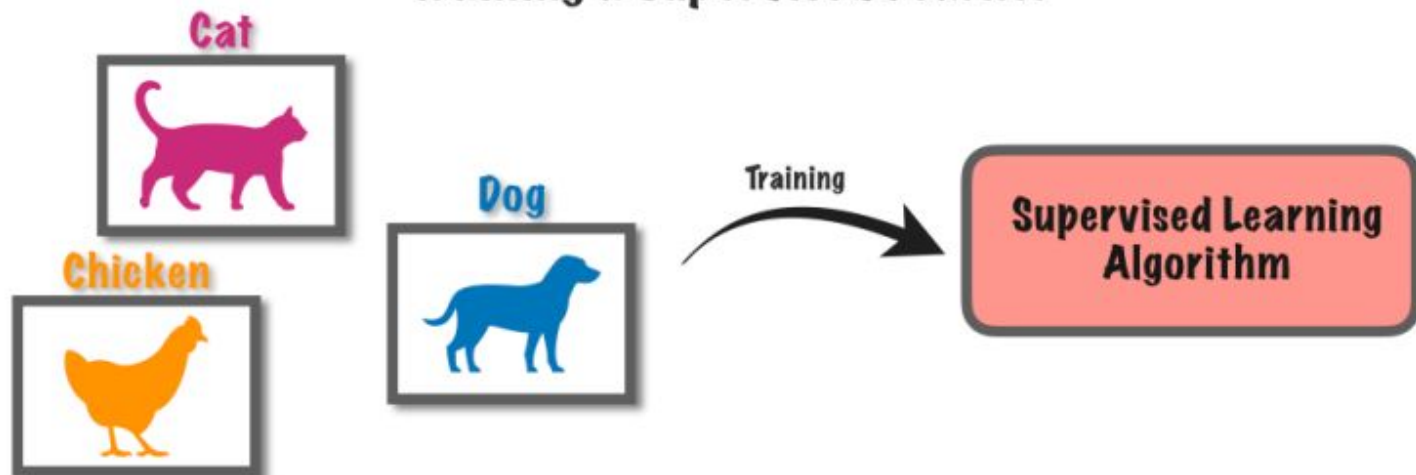
Input data



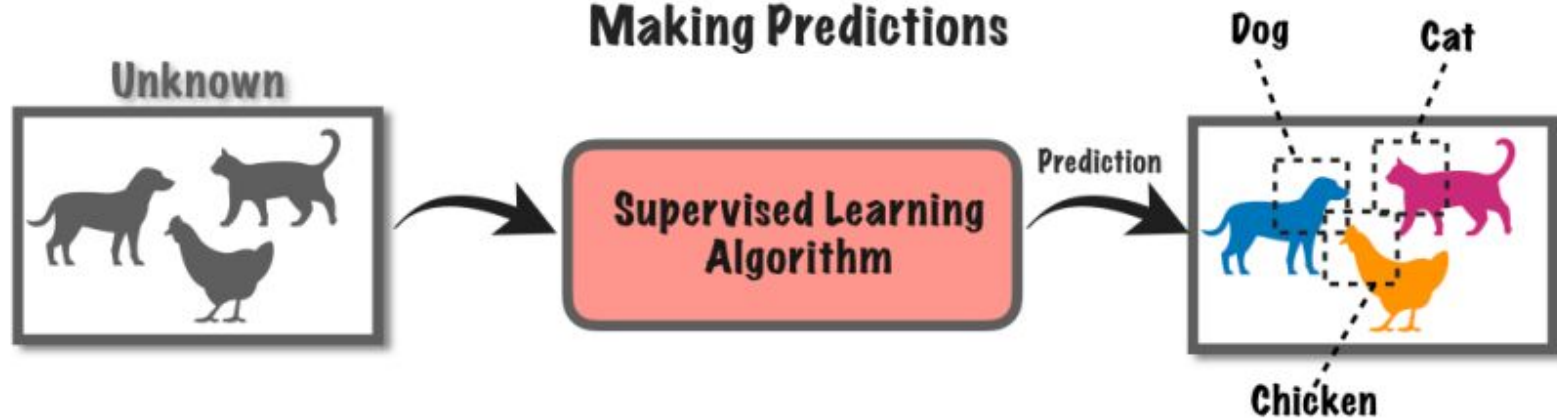
Model



## Training a Supervised Learner



## Making Predictions



Instantaneous rpm,  
torque, Fuel Cons. etc.

Machine Learning model  
(Random Forest, ANN,...)

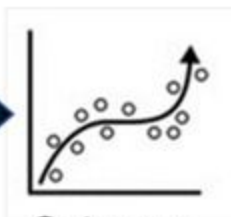


**Historical  
data**

**Model  
training**

**Predictive  
model**

Other cycle,  
calibration changes,..

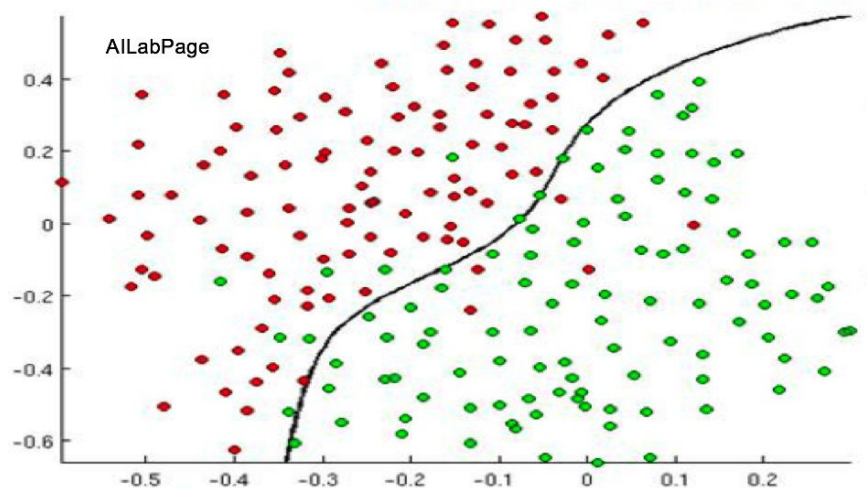
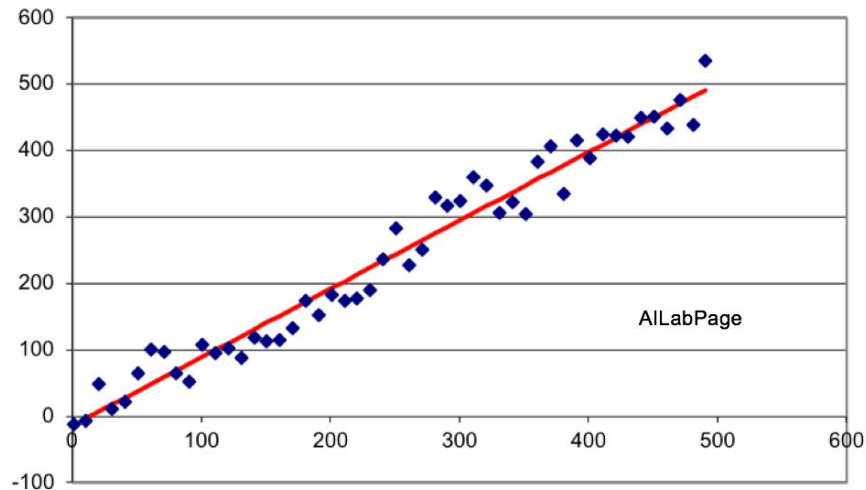


**New data**

**Predictive  
model**

**Outcome  
(predictions,  
analysis)**





## Regression

The system attempts to predict a value for an input based on past data.

Example – 1. Temperature for tomorrow



## Classification

In classification, predictions are made by classifying them into different categories.

Example – 1. Type of cancer 2. Cancer Y/N



# Machine learning

Machine learning means learning from data:

- We have a quantitative outcome (regression) or categorical outcome (classification)
- We want to predict the *outcome* based on a set of *features* (supervised)
- We have a *training set*
- We build a prediction model for new unseen objects. The objective is to predict accurately

# Vocabulary

Outcome: Usually denoted by  $Y$

Features: Usually denoted by  $X$  ( $X$  is a vector of  $k$  features)

Training set:  $(x_1, y_1), \dots, (x_n, y_n)$

Objective: get a good prediction of  $Y$  called  $\hat{Y} = f(X)$ .

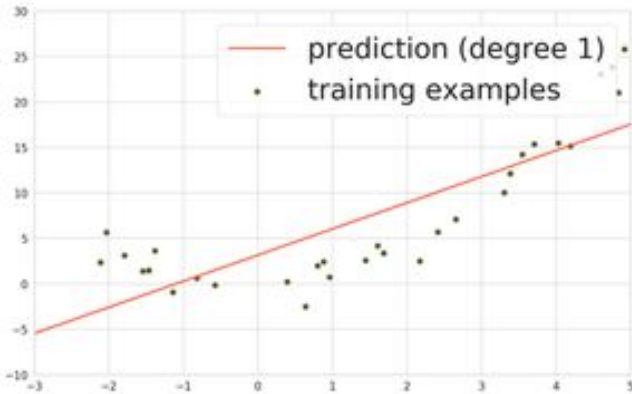
LOSS FUNCTION for penalizing errors (cost function)

Squared loss error  $(Y - f(X))^2$        $(Y - \hat{Y})^2$

# Types of Fit

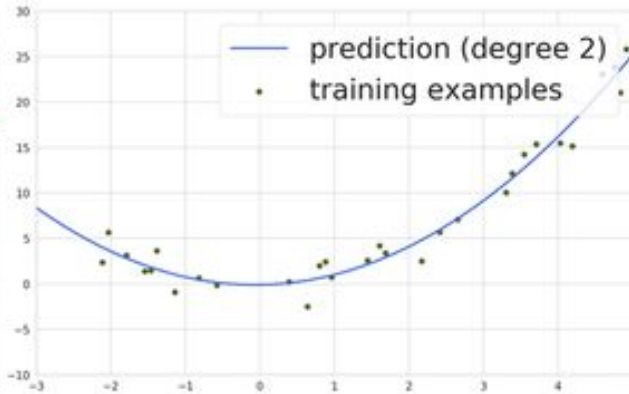
## Underfit

High bias



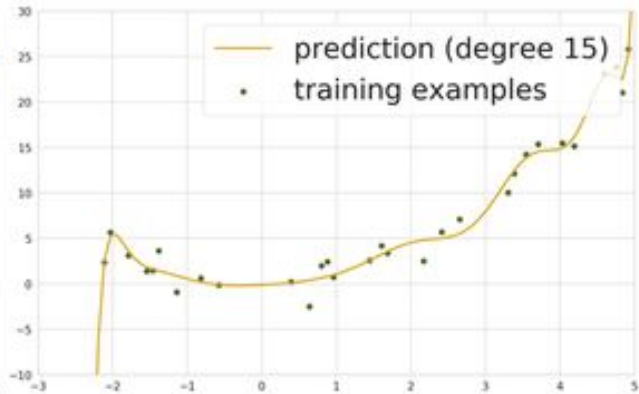
## Good Fit

Low bias, low variance



## Overfit

High variance



Types of Model Fit

# Metrics

R2:

$$R^2 = 1 - \frac{\Sigma(y - \hat{y})^2}{\Sigma(y - \bar{y})^2}$$

Adjusted R2:

$$R^2_{adj} = 1 - \left[ \frac{(1 - R^2)(n - 1)}{n - k - 1} \right]$$

RMSE:

$$\text{RMSE} = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$