



# Regression and its Applications

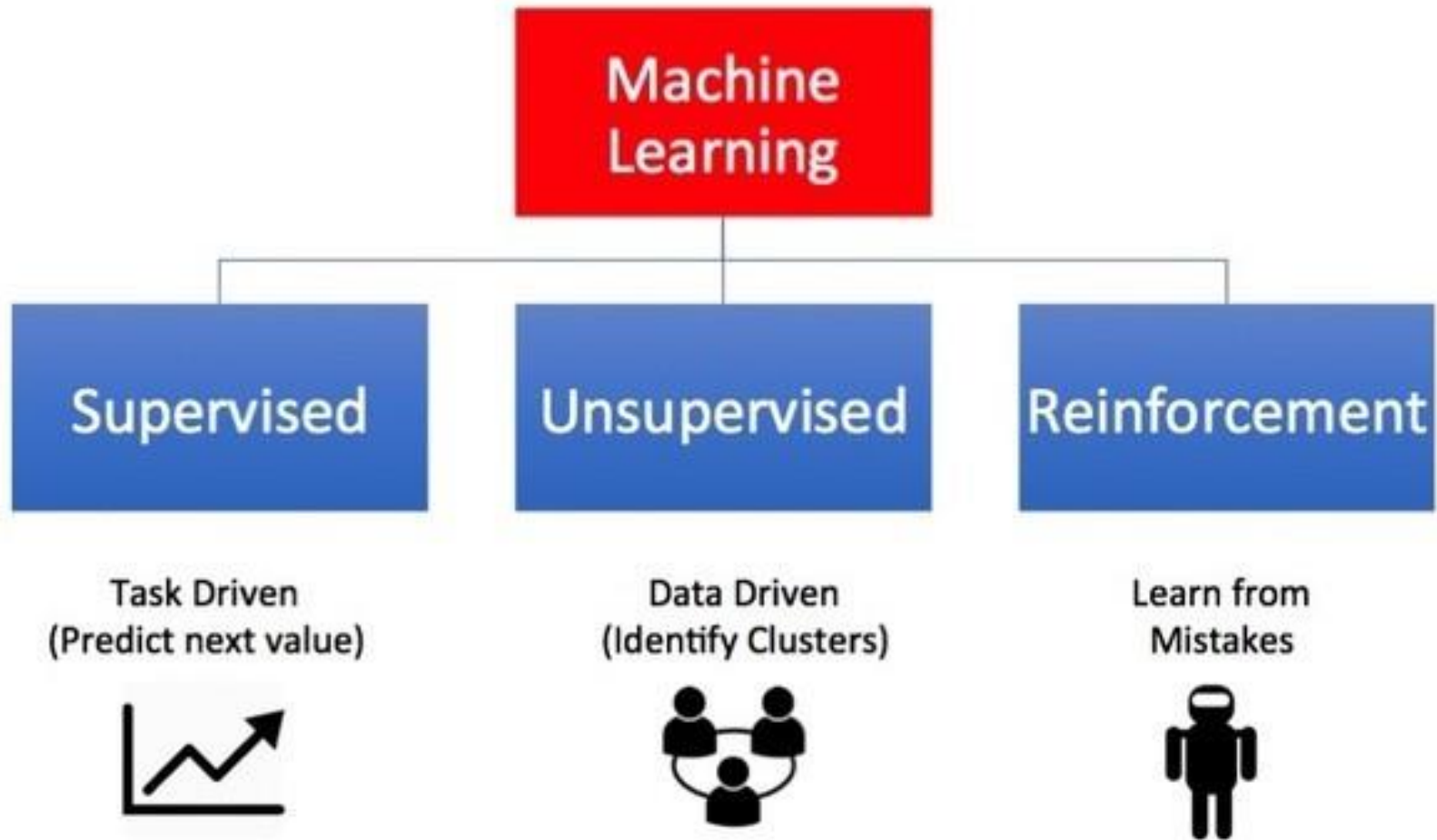
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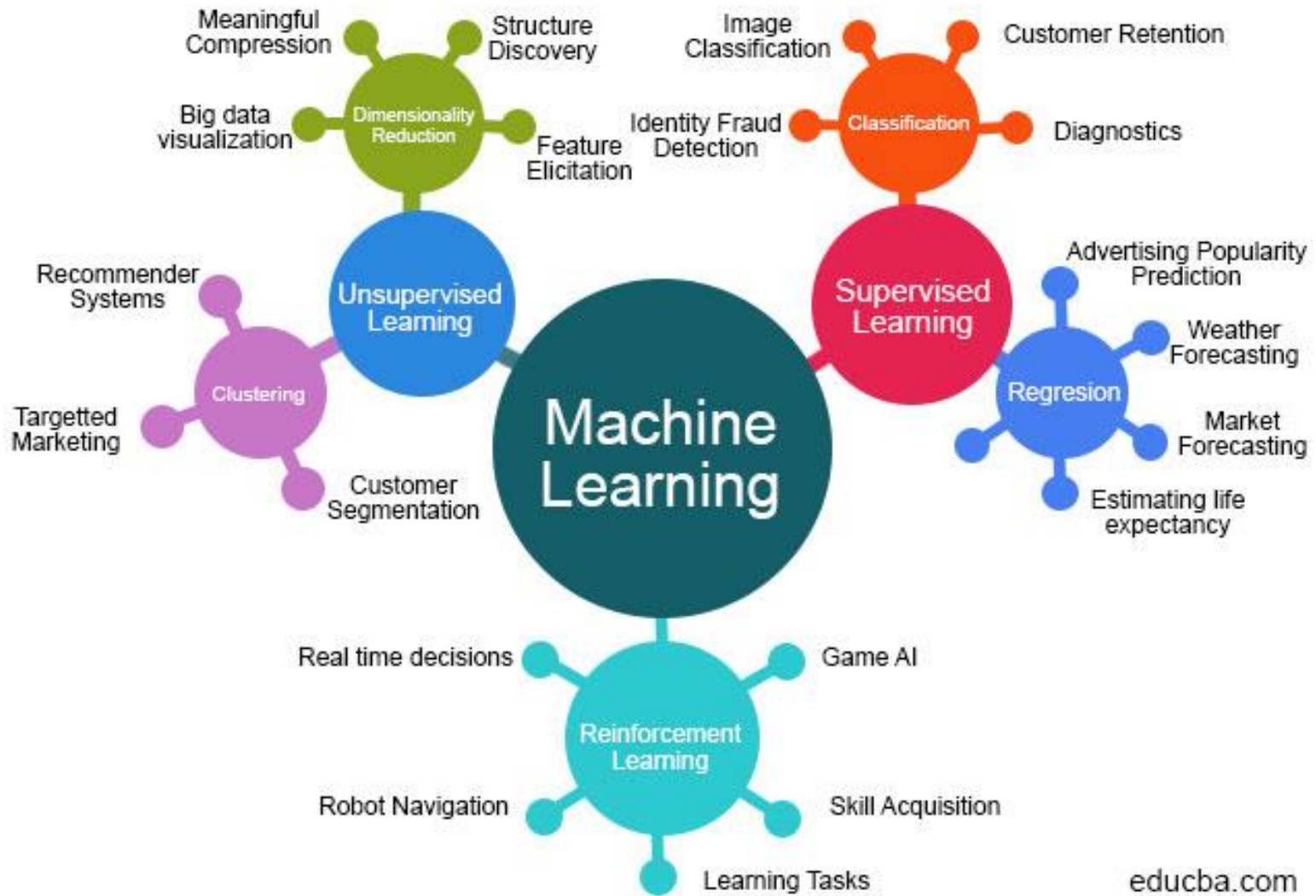
Mar 26, 2019

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# Types of Machine Learning



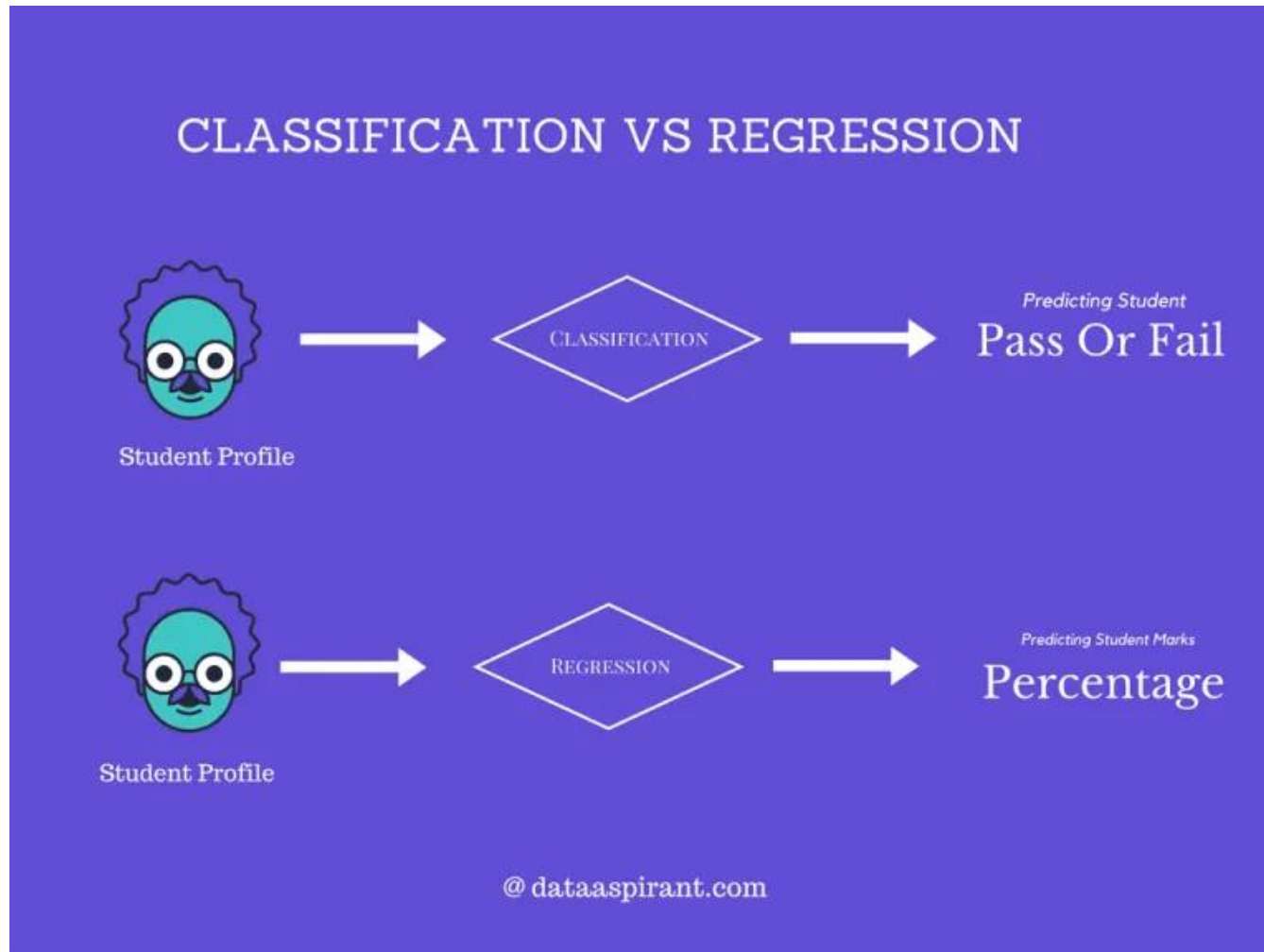
# Machine Learning Algorithms



educba.com

<https://www.educba.com/machine-learning-algorithms/>

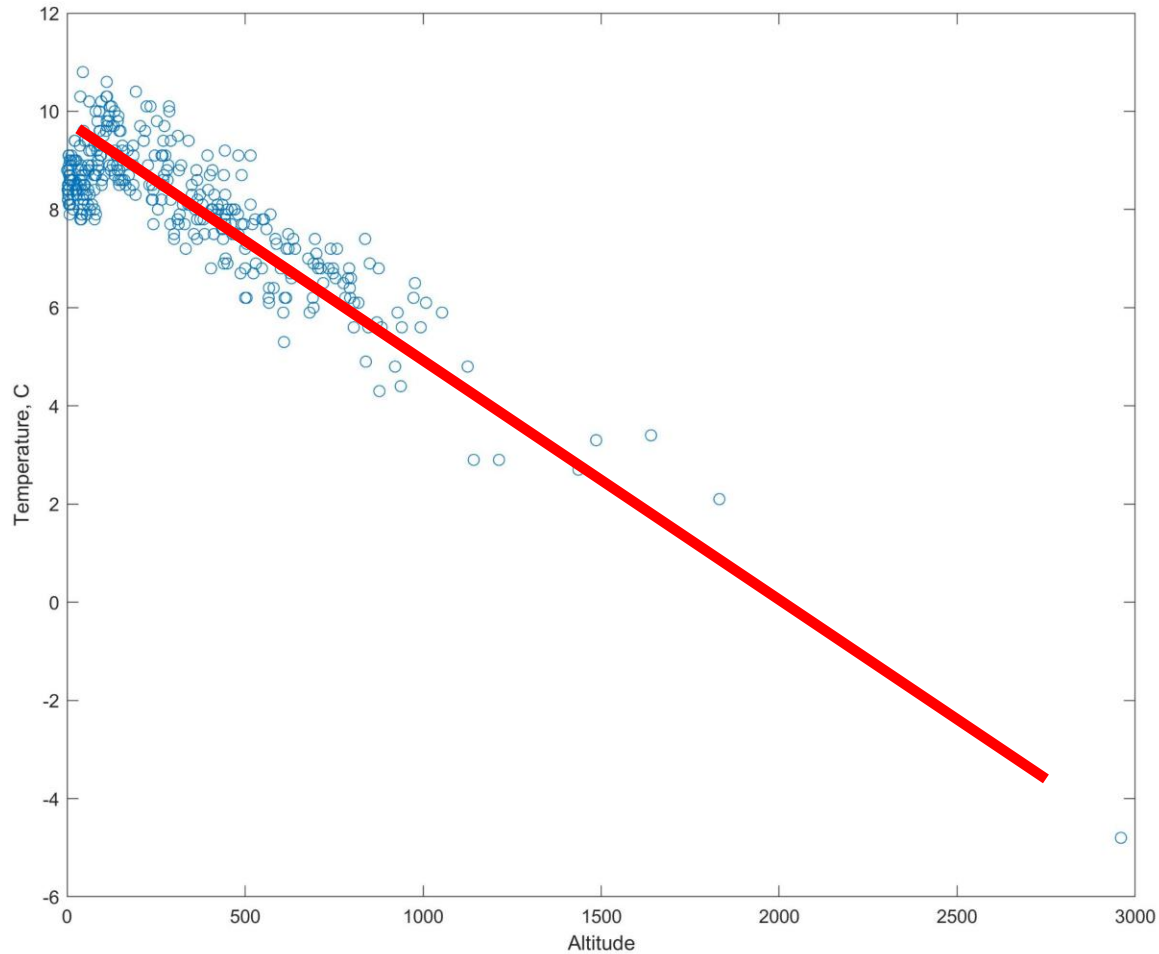
# Classification vs. Regression



# What is Linear Regression?

- **Learning**
  - A supervised algorithm that learns from a set of training samples.
  - Each training sample has one or more input values and a single output value.
  - The algorithm learns the line, plane or hyper-plane that best fits the training samples
- **Prediction**
  - Use the learned line, plane or hyper-plane to predict the output value for any
  - Input sample

# Altitude vs. Temperature



DWD data (Deutscher Wetterdienst)  
data was taken at 349 weather station

# Simple Linear Regression

Constant

Coefficient

$$y = b_0 + b_1 * x_1$$



Dependent variable



Independent variable

# Simple Linear Regression

Sea level  
temperature

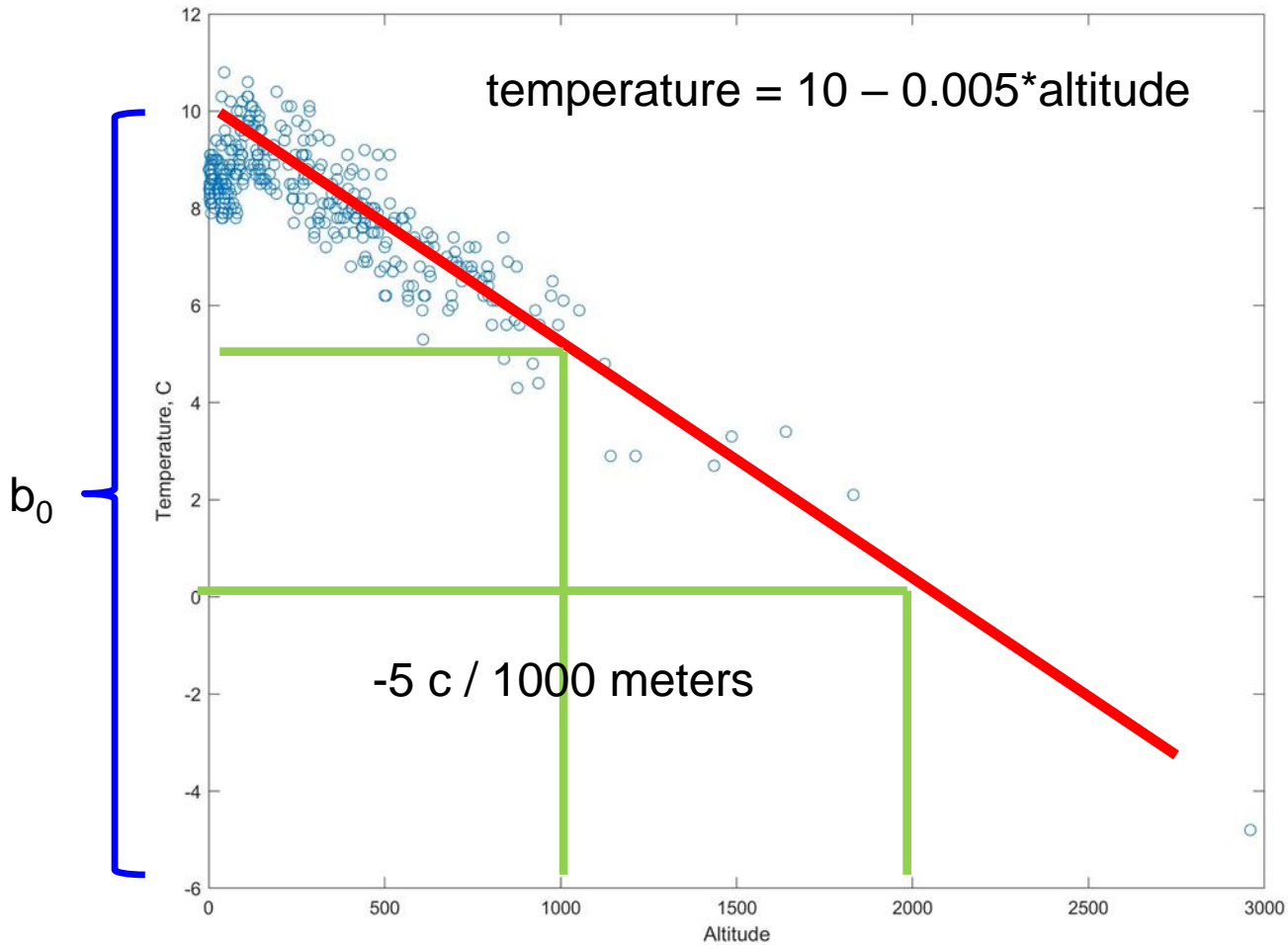
Coefficient of unit  
altitude change on  
temperature

$$y = b_0 + b_1 * x_1$$

Temperature

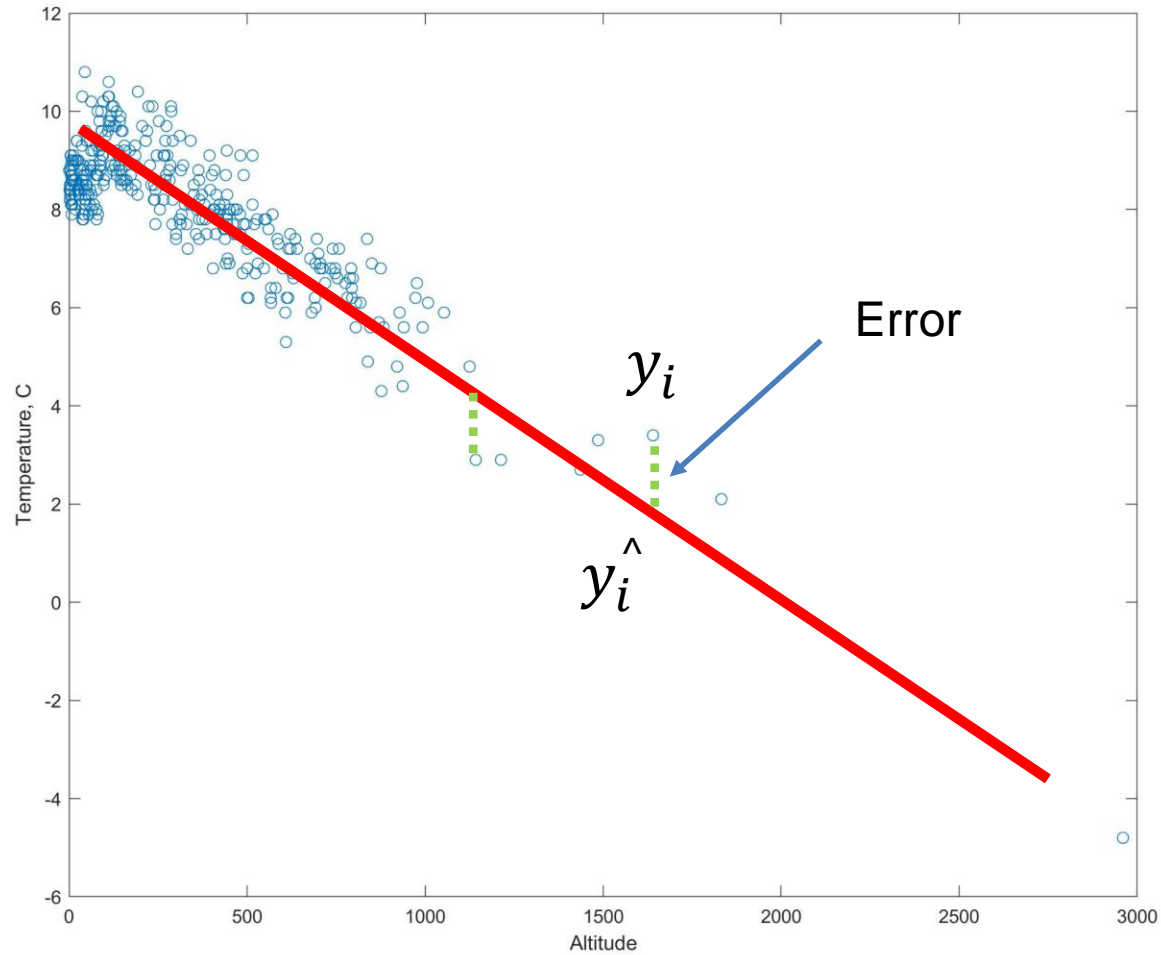
Altitude

# Altitude vs. Temperature

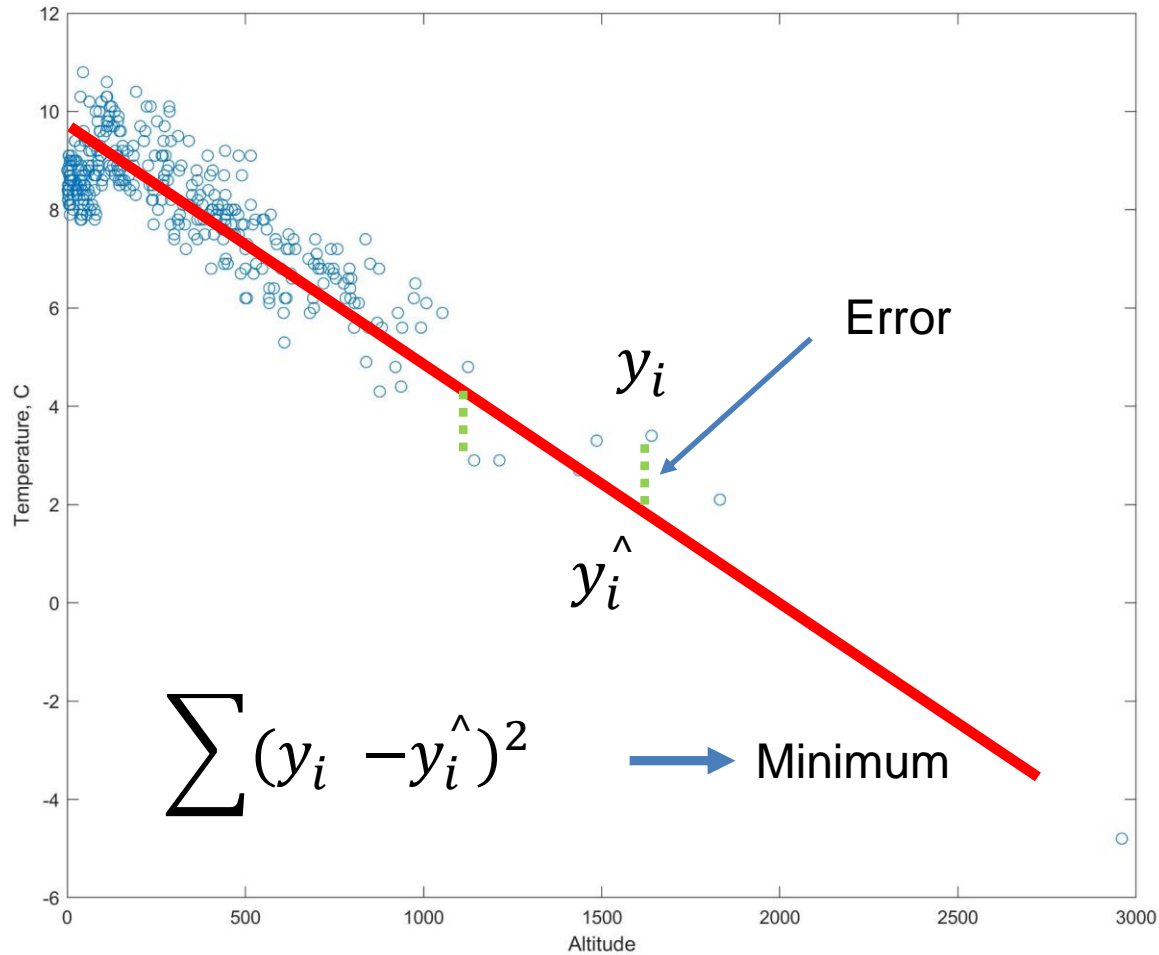


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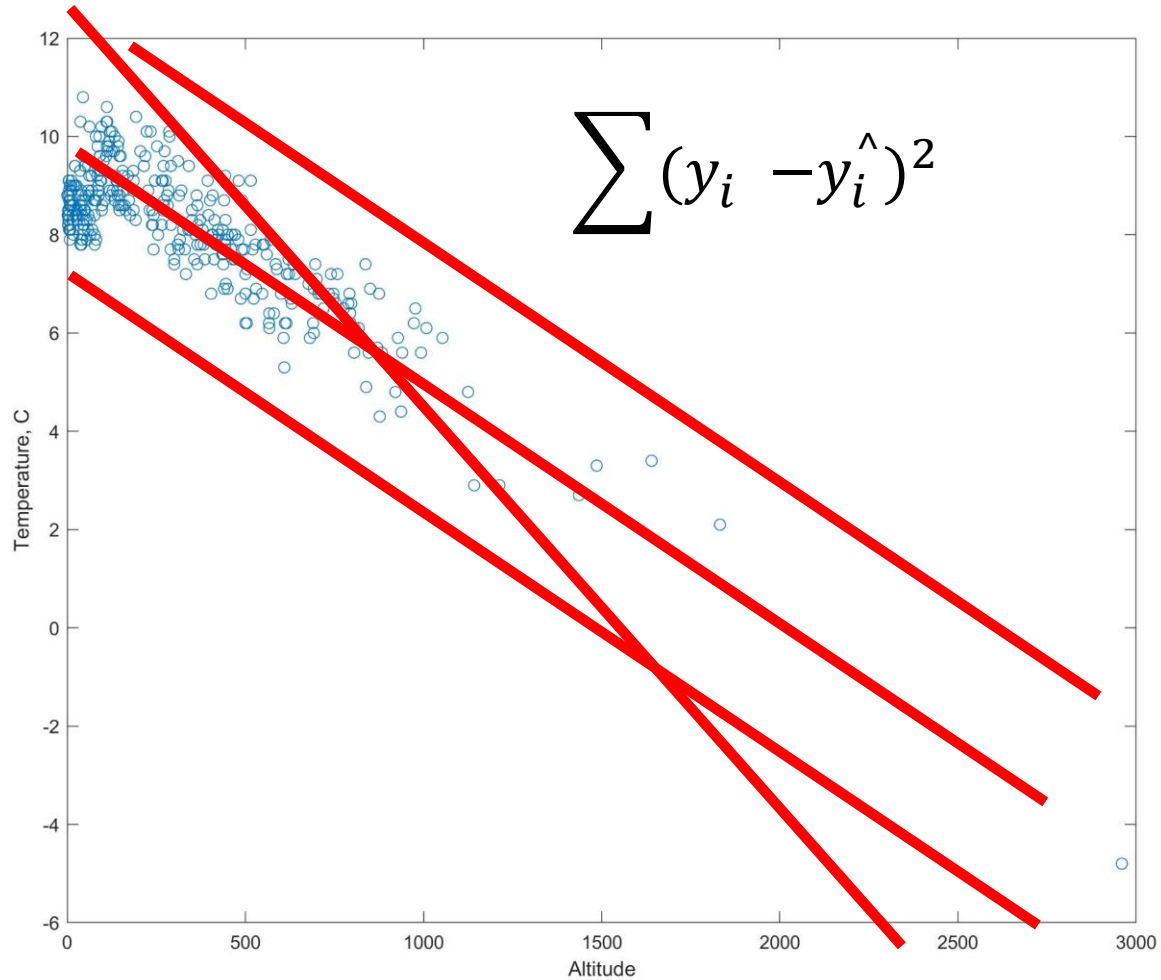
# How to find errors?



# How to define errors? Ordinary Least Square



# How to find the best fitted line?



# Simple vs. Multiple Linear Regression

- Simple Linear Regression

$$y = b_0 + b_1 * x_1$$

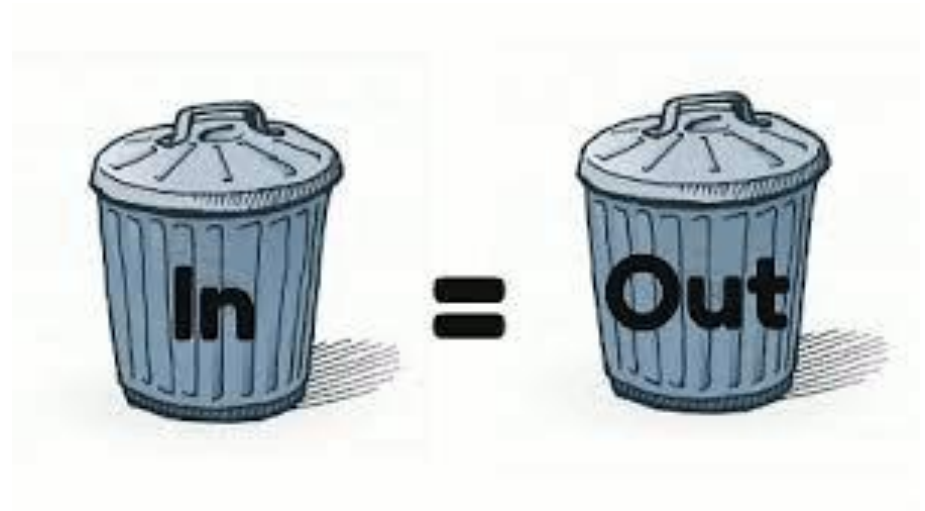
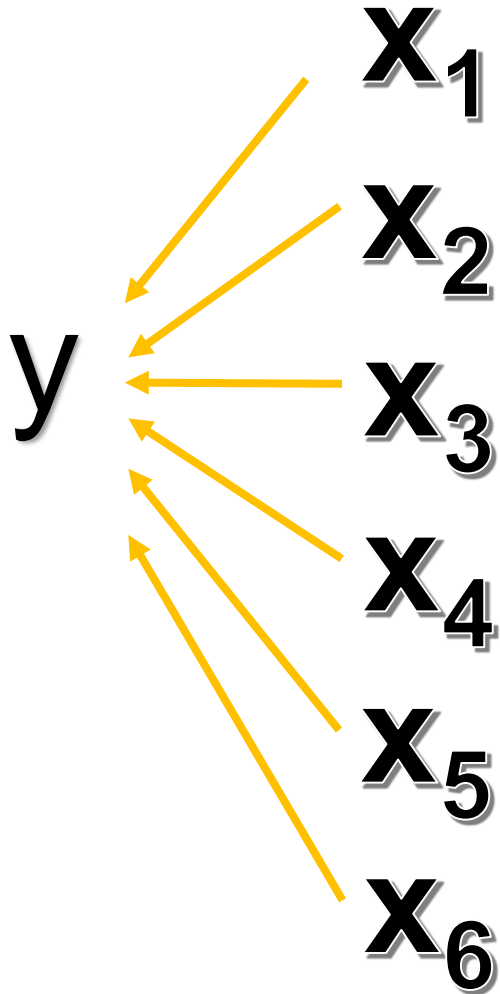
- Multiple Linear Regression

$$y = b_0 + b_1 * x_1 + b_2 * x_2 + b_3 * x_3 + \dots + b_n * x_n$$


# Assumption of Linear Regression

- Linearity
- Homoscedasticity
- Multivariate normality
- Independence of errors
- Lack of multicollinearity

# Building a Multiple Regression Model



# Variable Selection Methods

- All-in
  - Backward Elimination
  - Forward Selection
  - Bidirectional Elimination
  - Score comparison
- 
- Stepwise Method

# Building A Model

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## Backward Elimination

**STEP 1:** Select a significance level to stay in the model (e.g.  $SL = 0.05$ )



**STEP 2:** Fit the full model with all possible predictors



**STEP 3:** Consider the predictor with the highest P-value. If  $P > SL$ , go to STEP 4, otherwise go to FIN



**STEP 4:** Remove the predictor



**STEP 5:** Fit model without this variable\*

# Building A Model

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## Forward Selection

**STEP 1:** Select a significance level to enter the model (e.g.  $SL = 0.05$ )



**STEP 2:** Fit all simple regression models  $y \sim x_n$ . Select the one with the lowest P-value



**STEP 3:** Keep this variable and fit all possible models with one extra predictor added to the one(s) you already have



**STEP 4:** Consider the predictor with the lowest P-value. If  $P < SL$ , go to STEP 3, otherwise go to FIN

# Regressions

- Simple Linear Regression

$$y = b_0 + b_1 * x_1$$

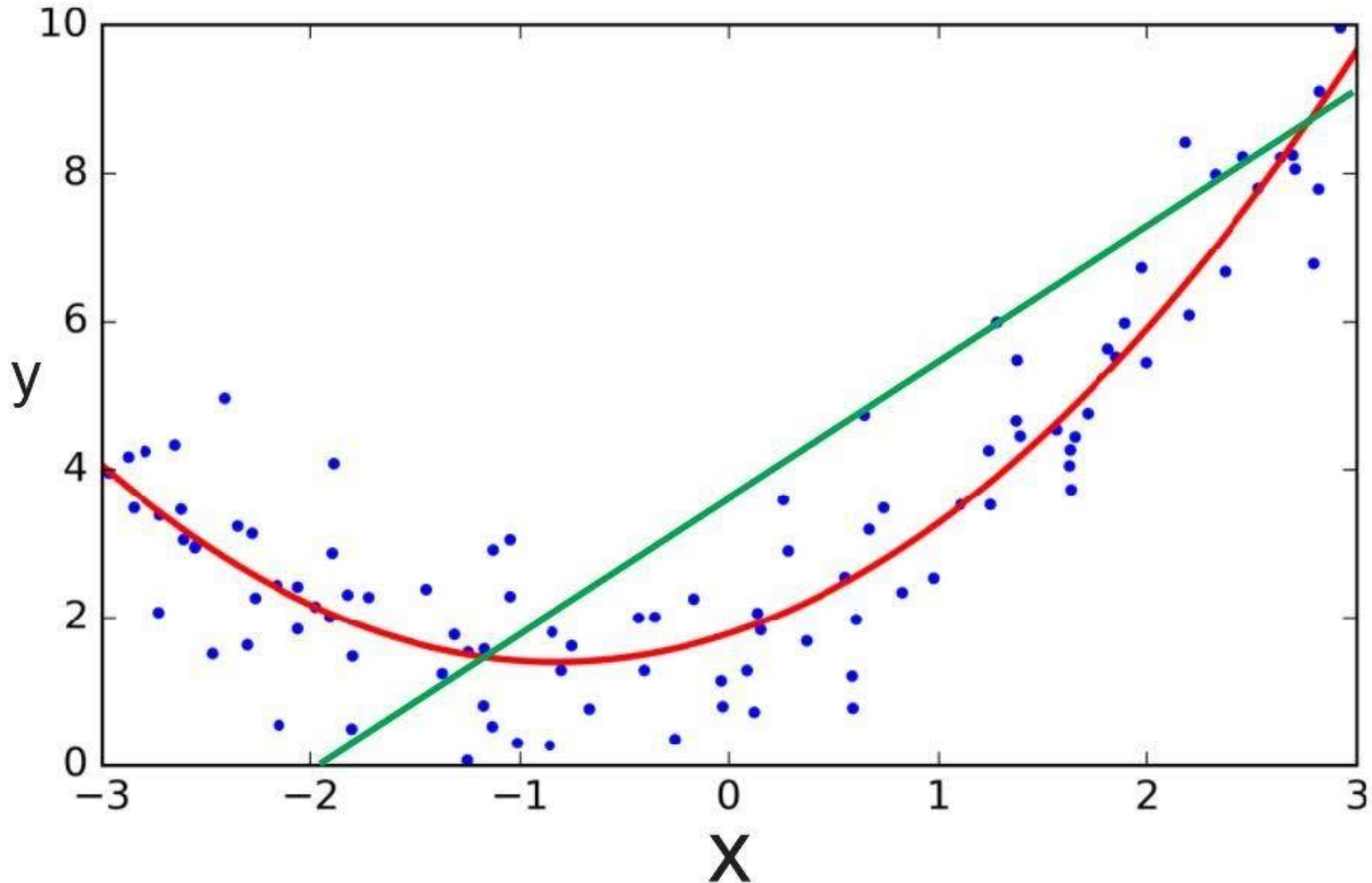
- Multiple Linear Regression

$$y = b_0 + b_1 * x_1 + b_2 * x_2 + b_3 * x_3 + \dots + b_n * x_n$$

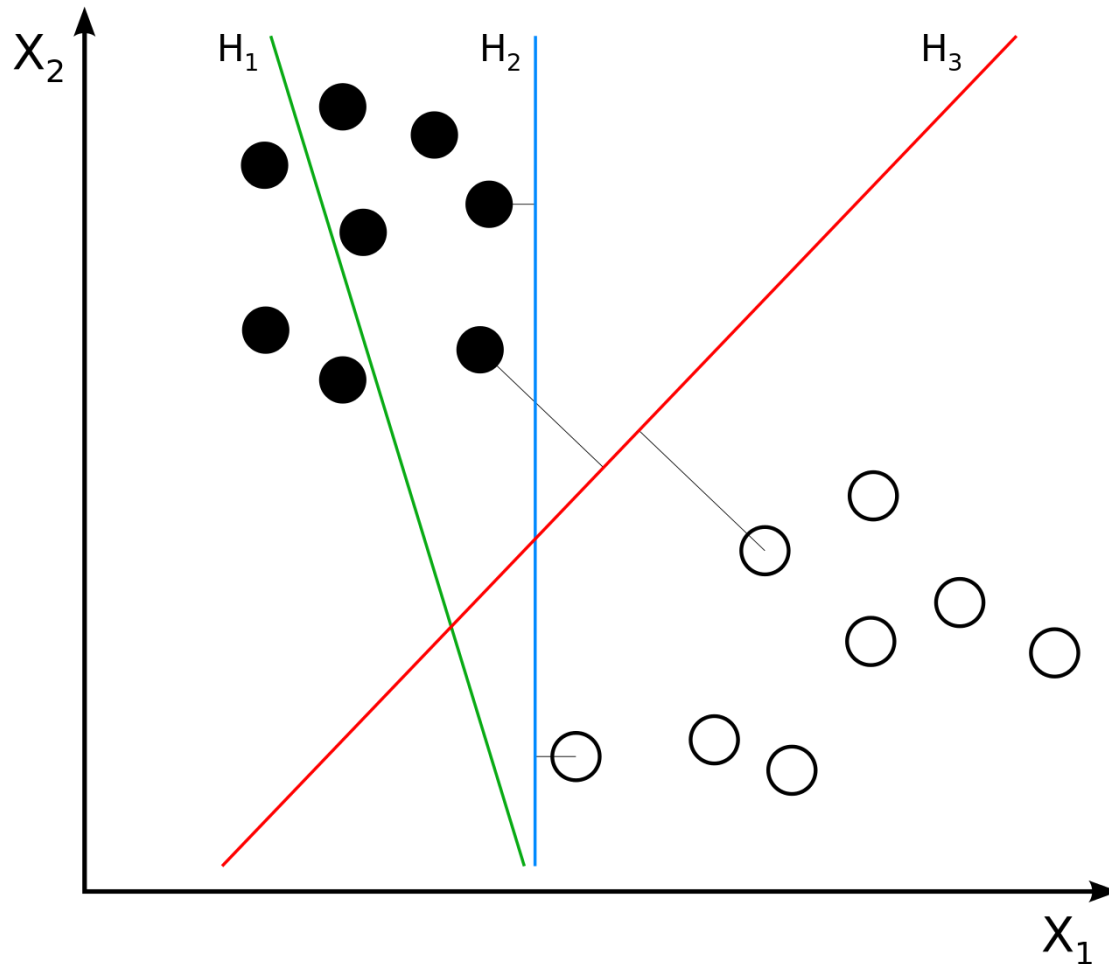
- Polynomial Linear Regression

$$y = b_0 + b_1 * x_1 + b_2 * x_1^2 + b_3 * x_1^3 + \dots + b_n * x_1^n$$

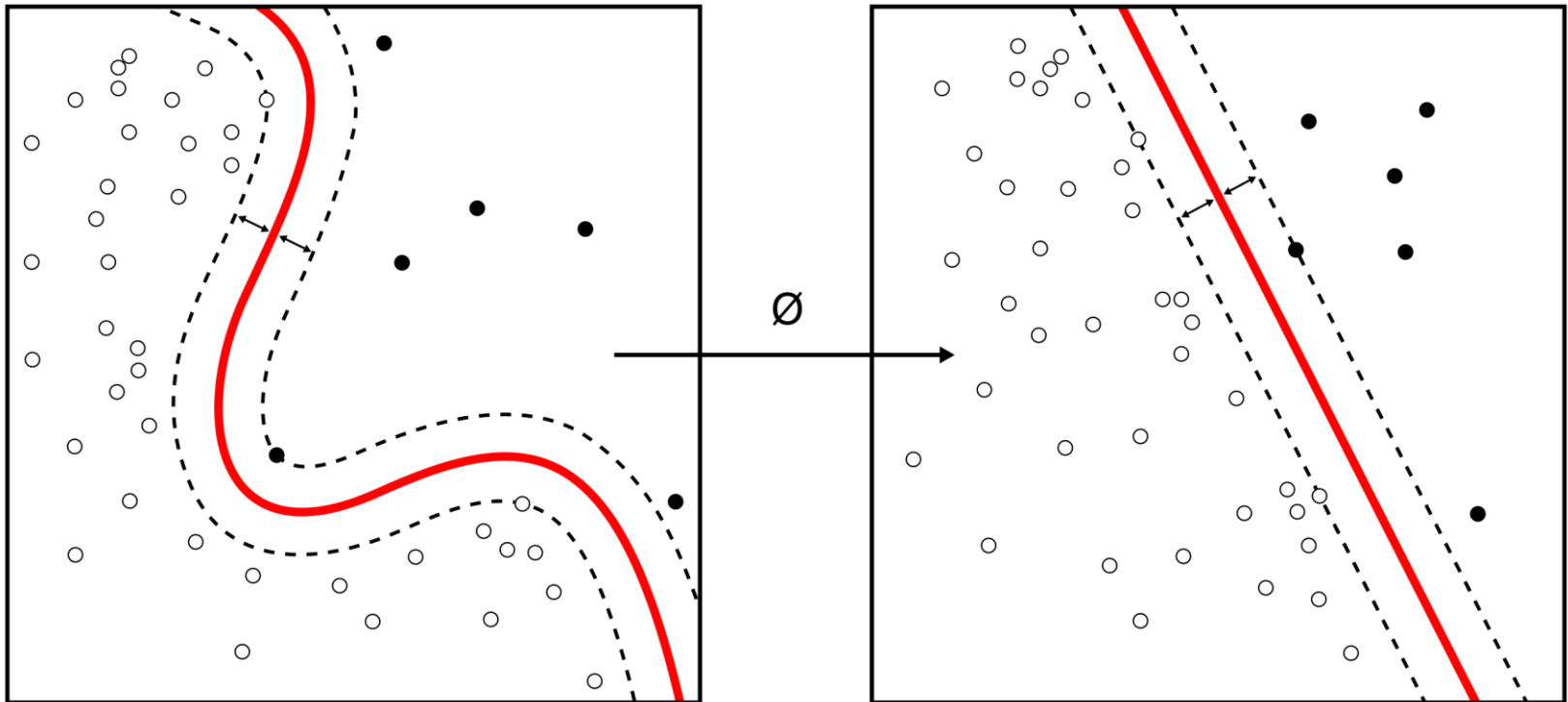
# Polynomial vs. Linear Regression



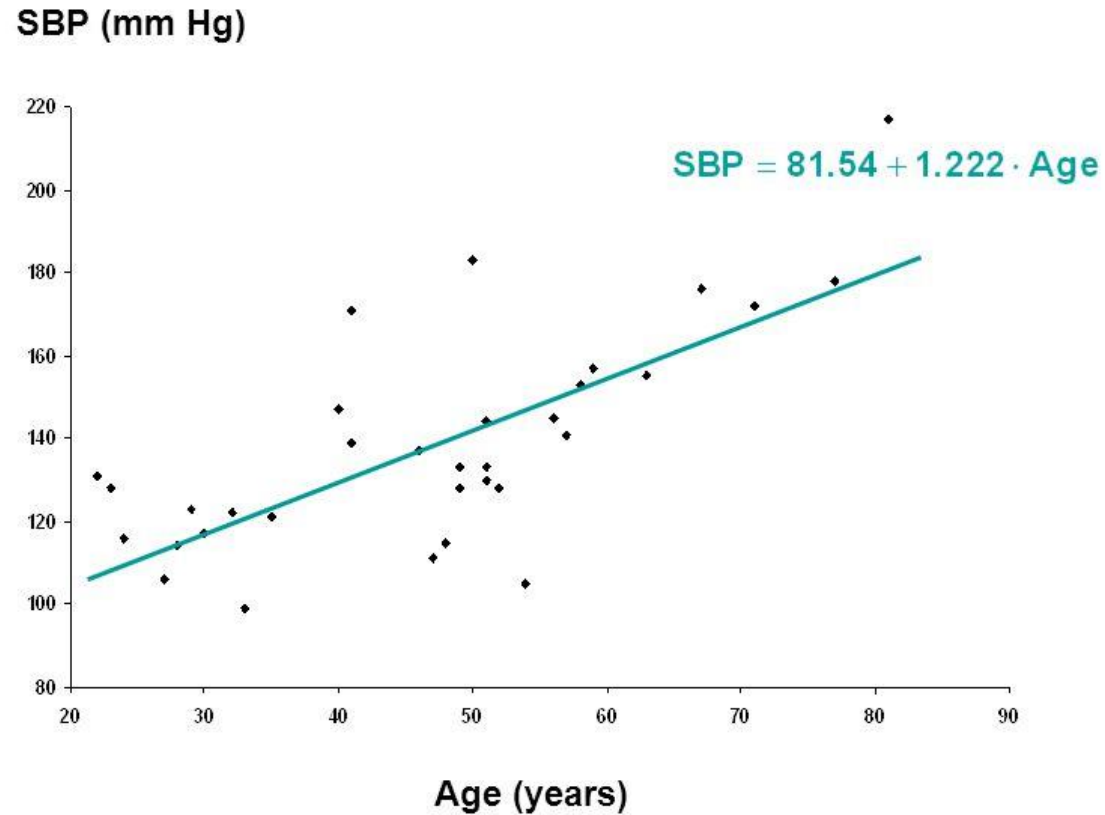
# Support Vector Regression



# Support Vector Regression

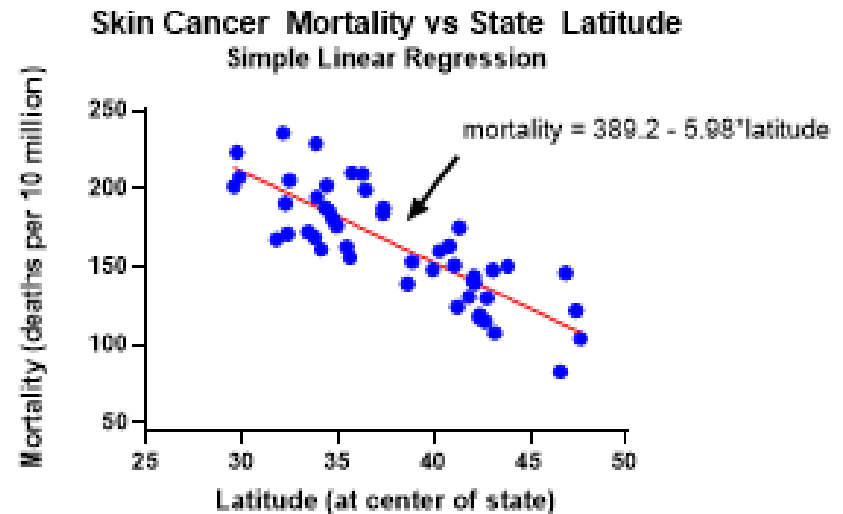
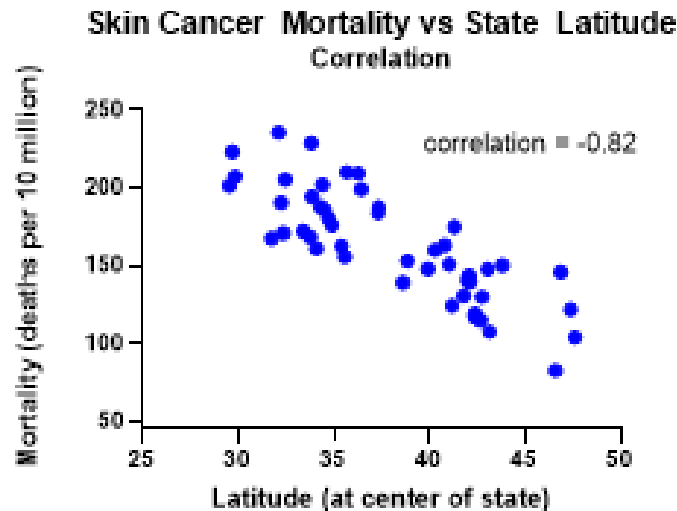


# Applications of Regression



adapted from Colton T. Statistics in Medicine. Boston: Little Brown, 1974

# Correlation vs. Coefficient



<https://www.graphpad.com/support/faq/what-is-the-difference-between-correlation-and-linear-regression/>

# Summary of Regression Methods

- Widely used in medical research
- Easy implementation
- Good interpretability
- Limited learnability