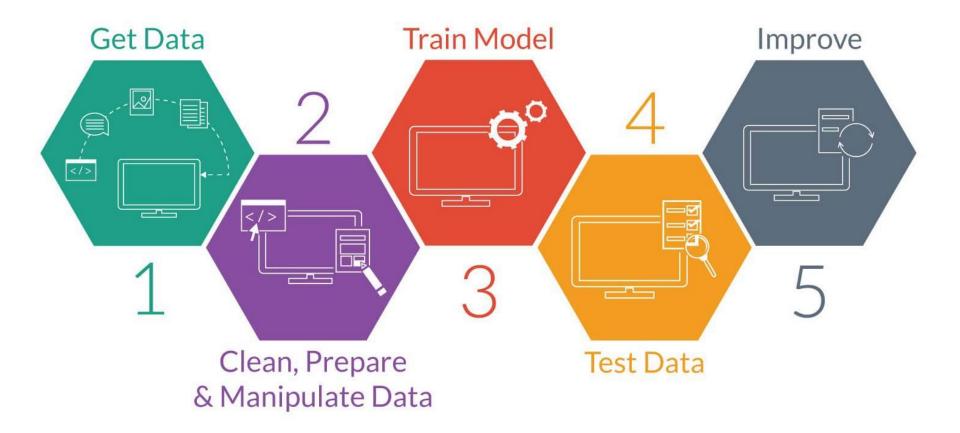
#### **Feature selection**

...or variables, or attributes...

# Boom!



# GIGO

Training	data			
var1	var2	var3	var4	result
5	2	2	1	dead
5	1	2	2	allve
3	3	1	2	dead
5	3	2	1	alive
4	3	2	2	alive

# **Advantages**

- 1. Could improve the results
- 2. Faster training

#### Univariante feature selection

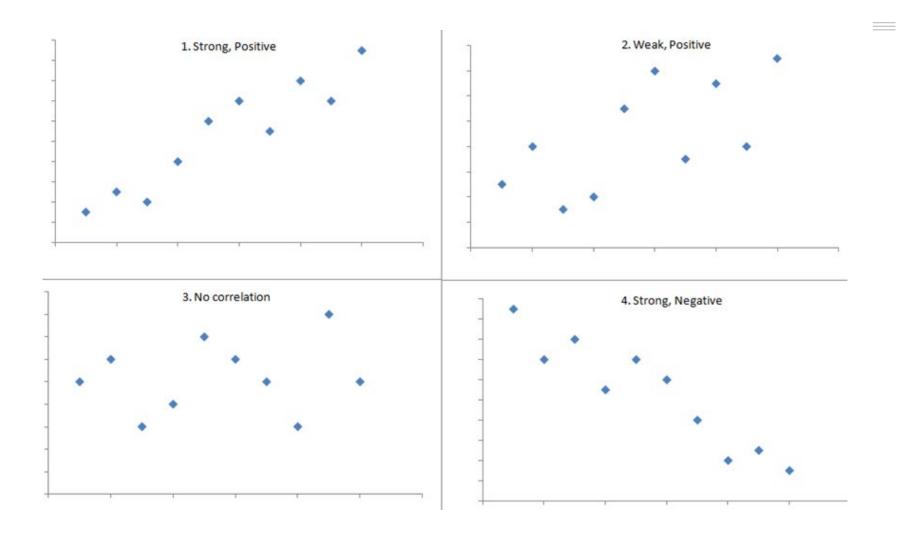
Training	data			
var1	var2	var3	var4	result
5	2	2	1	dead
5	1	2	2	allve
3	3	1	2	dead
5	3	2	1	alive
4	3	2	2	alive

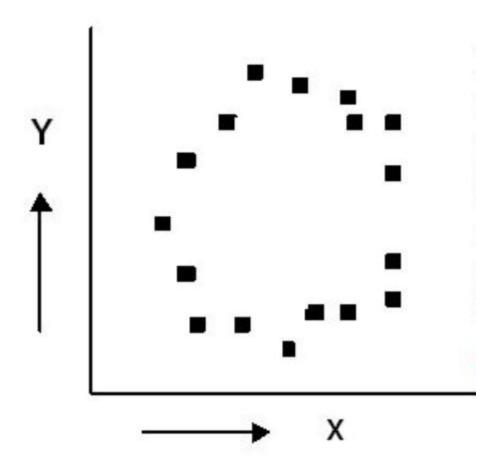
#### Remove no sense variables

#### Low variance variables

#### **Pearson's Correlation**

$$r_{xy} = rac{\sum_{i=1}^{n}(x_i-ar{x})(y_i-ar{y})}{\sqrt{\sum_{i=1}^{n}(x_i-ar{x})^2}\sqrt{\sum_{i=1}^{n}(y_i-ar{y})^2}}$$





# Deal with categorical variables

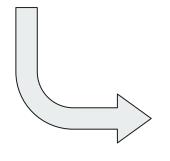
- Numerical Value
- One hot encoding

#### **Get dummies!**

numeric_variable	categorical_variable

0	1	Α
1	2	Α
2	5	В
3	8	Α

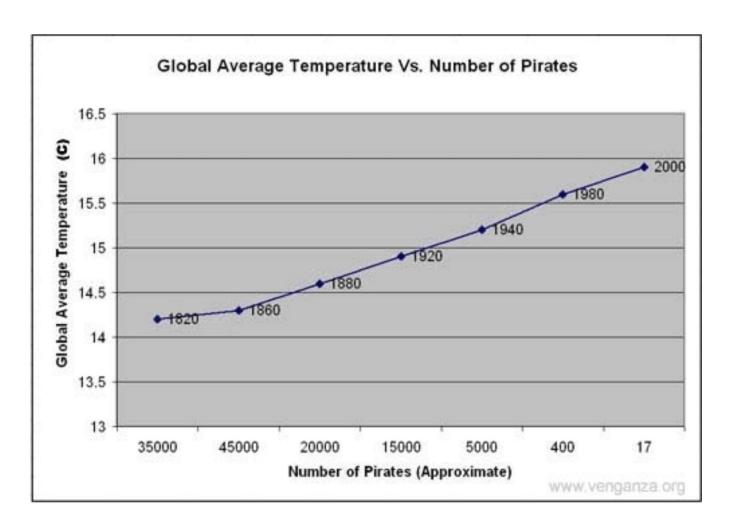
df = pd.get\_dummies(df, columns=[])



	numeric_variable	categorical_variable_A	categorical_variable_B
0	1	1	0
1	2	1	0
2	5	0	1
3	8	1	0

### **Practice time**

#### Correlation isn't causation



# **Chi-square test**

- For categorical variables
- Statistical test
- Likelihood of correlation

### **Practice time**

#### Tree-based feature selection

### **Practice time**

#### Multivariante methods

- Slower than univariante
- More powerful

# Wrapper methods

- Forward Selection
- Recursive Feature elimination

#### **Embedded Methods**

- LASSO
- Ridge Regression

### **Practice time**

Thank you!