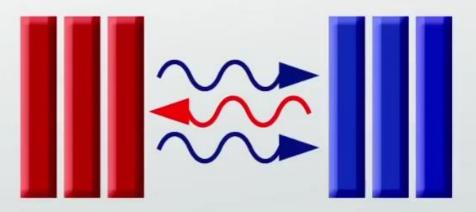
# **Evaluation Parameters for Regression**

- Parameters on which we compare different models and system.
  - Performance parameters:
    - Correlation
    - $\blacksquare$   $\mathbb{R}^2$
    - Accuracy
    - Prediction Time
    - Error (RMSE, MAE, MSA, etc)



### 1. Correlation: Pearson

the relationship between two sets of variables used to describe or predict information



### 1. Correlation: Pearson

#### **Formula**

$$\mathbf{r} = \frac{\mathbf{n}(\mathbf{\Sigma}\mathbf{x}\mathbf{y}) - (\mathbf{\Sigma}\mathbf{x})(\mathbf{\Sigma}\mathbf{y})}{\sqrt{\left[\mathbf{n}\mathbf{\Sigma}\mathbf{x}^2 - (\mathbf{\Sigma}\mathbf{x})^2\right]\left[\mathbf{n}\mathbf{\Sigma}\mathbf{y}^2 - (\mathbf{\Sigma}\mathbf{y})^2\right]}}$$

$$r = \frac{\sum xy - \frac{\sum x \sum y}{n}}{\sqrt{\left(\sum x^2 - \frac{\left(\sum x\right)^2}{n}\right)\left(\sum y^2 - \frac{\left(\sum y\right)^2}{n}\right)}}$$



### 1. Correlation: Pearson

- The closer the number is to positive one, the stronger the positive correlation
- The closer the number is to negative one, the stronger the negative correlation
  - The closer the number is to zero, the weaker the correlation
- 1
- Zero means there
  is no correlation
  between the variables



### 2. Coefficient of determinant

- The coefficient of determination (R2) summarizes the **explanatory power of the regression model**.
- R2 describes the <u>proportion of variance of the dependent variable</u> <u>explained by the regression model</u>.
- If the regression model is perfect then R2 is 1 and if the regression model is a total failure then R2 is zero i.e. no variance is explained by regression.
- It is defined as follows:  $R^2 = r * r$

## 3. Error

Mean squared error

$$ext{MSE} = rac{1}{n} \sum_{t=1}^{n} e_t^2$$

Root mean squared error

$$ext{RMSE} = \sqrt{rac{1}{n}\sum_{t=1}^n e_t^2}$$

Mean absolute error

$$ext{MAE} = rac{1}{n} \sum_{t=1}^n |e_t|$$

Where, e is the error between x and y

## 4

### 4. Accuracy

 The accuracy is calculated as percentage deviation of predicted target with actual target (with or without <u>acceptable</u> <u>error</u>).

### Example:

- Dose given by doctor (100 mg +-10) → OK
- Dose given by doctor (100 mg +-50) → Not OK



### 4. Accuracy

$$Accuracy = \frac{100}{n} \sum_{i=1}^{n} q_i$$

$$q_i = \begin{cases} 1 & ifabs(p_i - a_i) \le err\\ 0 & otherwise \end{cases}$$



### 4. Accuracy: Find

Accuracy with +-5 error

- = IF (ABS (A4 B4)  $\leq$  5, 1, 0)
- = AVERAGE (C4 : C16 ) \* 100

### 4

### 4. Accuracy: Find

		Accuracy				
		With +-5 err	With +- 10 err	With +- 20 err	With +- 30 err	
Actual	Predicted	23.08	38.46	76.92	92.31	
24	21	1	1	1	1	
50	48	1	1	1	1	
42	54	0	0	1	1	
51	32	0	0	1	1	
11	32	0	0	0	1	
24	11	0	0	1	1	
32	38	0	1	1	1	
27	56	0	0	0	1	
59	54	1	1	1	1	
56	49	0	1	1	1	
45	29	0	0	1	1	
16	59	0	0	0	0	
56	37	0	0	1	1	



# **Evaluation Parameters for Binary Classification**

## Binary Classification TRUE/FALSE

Actual	Predicted
1	0
1	1
0	1
1	0
0	1
0	1
0	1
0	0
1	0
0	1

## **Confusion Matrix / Error Matrix**

n= 10	Predicted: YES	Predicted: NO	•
Actual: YES	TP = 1	<b>FN</b> = 3	4
Actual: NO	FP = 5	TN = 1	6
	6	4	

Sensitivity or recall = 
$$\frac{TP}{(TP + FN)}$$



Specificity = 
$$\frac{TN}{(TN + FP)}$$

Accuracy = 
$$\frac{(TP + TN)}{(TP + FN + FP + TN)}$$

Positive Predictive Value or Precision= 
$$\frac{TP}{(TP + FP)}$$

Negative Predictive Value = 
$$\frac{TN}{(TN + FN)}$$

F-measure= 2 \* 
$$\frac{(precision*recall)}{(precision+recall)}$$

### **Sensitivity / Specificity**

- Sensitivity presents true positive rate.
  - $1 \rightarrow 1$
  - Cancer → Cancer
- Specificity presents true negative rate.
  - $0 \rightarrow 0$
  - Non-Cancer → Non-Cancer

## **Confusion Matrix / Error Matrix**

**Predicted Class** 

		Positive	Negative	
Actual Class	Positive	True Positive (TP)	False Negative (FN)  Type II Error	Sensitivity $\frac{TP}{\overline{(TP+FN)}}$
Actual Class	Negative	False Positive (FP)  Type I Error	True Negative (TN)	Specificity $\frac{TN}{(TN + FP)}$
		Precision $\frac{TP}{(TP+FP)}$	Negative Predictive  Value $\frac{TN}{(TN + FN)}$	Accuracy $\frac{TP + TN}{(TP + TN + FP + FN)}$

# Question: Calculate all parameters for

Actual	Predicted
1	0
1	1
0	1
1	0
0	1
0	1
0	1
0	0
1	0
0	1



## **Evaluation Parameters for Multi-Class Classification**

## Model Evaluation Parameters

#### Error Matrix

Actual/	0	1	2	3	4	5
Predicted						
0	1305	1	5	15	15	5
1	0	114	28	0	0	0
2	4	31	551	136	17	2
3	8	2	125	786	150	13
4	22	3	39	206	713	104
5	14	2	17	46	121	314

## Model Evaluation Parameters

For Accuracy

$$Accuracy = \frac{100}{n} \sum_{i=1}^{n} q_i$$

$$q_i = \begin{cases} 1 & p_i = a_i \\ 0 & otherwise \end{cases}$$

## **Model Evaluation Parameters**

Calculate error matrix and Accuracy

Actual	Predicted	
2	0	
1	2	
2	4	
0	4	
3	4	
0	4	
0	0	
0	2	
3	4	
0	0	
4	4	
4	4	

### **Class wise error**

		True/Actual		
		Cat (🐯)	Fish (��)	Hen ( <b>4</b> )
Pr	Cat (👹)	4	6	3
Predicted	Fish (��)	1	2	0
ed	Hen (🐴)	1	2	6

**Number of correctly classified**: 4 + 2 + 6 = 12**Total number** = 25

- Overall Accuracy = 12/25 = 48.0%
- Error (Cat): (1+1)/(4+1+1) = 2/6 = 0.33
- Error (Fish): (6+2)/(6+2+2) = 8/10 = 0.80
- Error (Hen): (3+0)/(3+0+6) = 3/9 = 0.33