

# **EE2211 Pre-Tutorial 3**

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### Agenda

- Recap
- Self-learning
- Tutorial 3

### Recap

- Linear Algebra
  - Vectors
  - Matrices
  - Linear Equations
- Probability
  - Axioms of Probability
  - Random Variable
  - Basic rules
  - Bayes' rule

#### Vectors

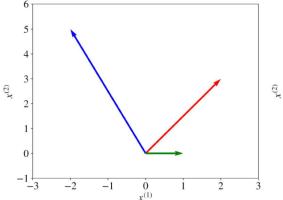
- A vector is an ordered list of scalar values
  - Denoted by a bold character, e.g. x or a
- Vectors can be visualized as, in a multi-dimensional space,
  - arrows that point to some directions, or

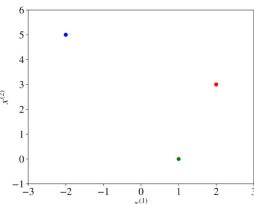
e.g. Velocity vector

– points

e.g. color points

Illustrations of three two-dimensional vectors,  $\mathbf{a} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$ ,  $\mathbf{b} = \begin{bmatrix} -2 \\ 5 \end{bmatrix}$ , and  $\mathbf{c} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ 





### **Matrices**

- A matrix is a rectangular array of numbers arranged in rows and columns
  - Denoted with bold capital letters, such as X or W
  - ➤ An example of a matrix with two rows and three columns:

$$\mathbf{X} = \begin{bmatrix} 2 & 4 & -3 \\ 21 & -6 & -1 \end{bmatrix}$$
 e.g. image frame

#### Note:

- For elements in matrix X, we shall use the indexing  $x_{1,1}$  where the first and second indices indicate the row and the column position.
- ➤ Usually, for input data, rows represent samples and columns represent features

### **Linear Equation**

Linear dependence and independence

• A collection of d-vector  $\mathbf{x}_1, \dots, \mathbf{x}_m$  (with  $m \ge 1$ ) is called linearly dependent if  $\beta_1 \mathbf{x}_1 + \dots + \beta_m \mathbf{x}_m = 0$ 

**Holds for some**  $\beta_1, \dots, \beta_m$  that are not all zeros.

• A collection of d-vector  $\mathbf{x}_1, \cdots, \mathbf{x}_m$  (with  $m \ge 1$ ) is called linearly independent if  $\beta_1 \mathbf{x}_1 + \cdots + \beta_m \mathbf{x}_m = 0$ 

only hold for  $\beta_1 = \cdots = \beta_m = 0$ .

Note: If all rows or columns of a square matrix **X** are linearly **independent**, then **X** is **invertible**.

### **Probability**

 We describe a random experiment by describing its procedure and observations of its *outcomes*.



Outcome

- Outcomes are mutual exclusive in the sense that only one outcome occurs in a specific trial of the random experiment.
  - This also means an outcome is not decomposable.
  - > All unique outcomes form a sample space.







Sample space





- A subset of sample space S, denoted as A, is an event in a random experiment  $A \subset S$ , that is meaningful to an application.
  - Example of an event: faces with numbers no greater than 3









### **Axioms of Probability**

- Assume events  $A \subseteq S$  and  $B \subseteq S$ .
  - 1. [Nonnegativity]  $Pr(A) \ge 0$  for every event A.
  - 2. [Normalization] Pr(S) = 1.
  - 3. [Additivity] If  $A \cap B = \emptyset$ , then  $Pr(A \cup B) = Pr(A) + Pr(B)$ .

otherwise 
$$Pr(A \cup B) = Pr(A) + Pr(B) - Pr(A \cap B)$$

It can be extended to more events:

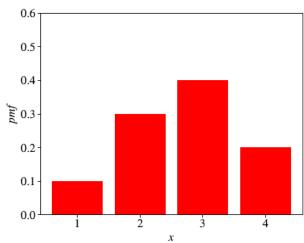
If 
$$A_1, A_2, ...$$
 are disjoint, then  $Pr(\bigcup_{k=1}^{\infty} A_k) = \sum_{k=1}^{\infty} Pr(A_k)$ .

#### Random Variable

- A random variable, usually written as an italic capital letter, like *X*, is a variable whose possible values are numerical outcomes of a random event.
- There are two types of random variables: discrete and continuous.

### Discrete Random Variable

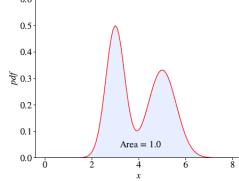
- A discrete random variable (DRV) takes on only a countable number of distinct values such as red, orange, blue or 1, 2, 3.
- The probability distribution of a discrete random variable is described by a list of probabilities associated with each of its possible values.
- This list of probabilities is called a probability mass function (pmf).
  - ➤ Like a histogram, except that here the probabilities sum to 1



A probability mass function

### Continuous Random Variable

- A continuous random variable (CRV) takes an infinite number of possible values in some interval.
  - Examples include height, weight, and time.
  - The number of values of a continuous random variable X is infinite, the probability Pr(X = c) for any c is 0.
  - Therefore, instead of the list of probabilities, the probability distribution of a CRV (a continuous probability distribution) is described by a **probability density** function (pdf).
  - ➤ The pdf is a function whose range is nonnegative and the area under the curve is equal to 1.



A probability density function

#### Two Basic Rules

Sum Rule

$$Pr(X = x) = \sum_{Y} Pr(X = x, Y = y_i)$$

Product Rule

$$Pr(X = x, Y = y) = Pr(Y = y | X = x) P(X = x)$$

### Bayes' Rule

- The conditional probability Pr(Y = y | X = x) is the probability of the random variable Y to have a specific value y, given that another random variable X has a specific value of x.
- The Bayes' Rule (also known as the Bayes' Theorem):

$$Pr(Y = y | X = x) = \frac{Pr(X = x | Y = y) Pr(Y = y)}{Pr(X = x)}$$
posterior
evidence

- ❖A guy was tested positive for a rare disease!
  - In the general population only 0.1% has this disease
- ❖The test is quite accurate
  - It can detect 99% of subjects with the disease
  - It only mis-identifies 2% of subjects without this disease

How worried should this guy be?

	Simple space (1,000 people)	Test Positive	Probability
Disease	1	1	~1/21
No Disease	999	20	~20/21

$$P(H|E) = \frac{P(E|H)P(H)}{P(E|H)P(H) + P(E|\overline{H})P(\overline{H})}$$

P(H): Chance of having this disease = 0.001 (Prior probability)

$$P(\overline{H}) = 0.999$$

P(E|H): Sensitivity = 0.99 (Likelihood)

 $P(E|\overline{H})$ : false positive rate = 0.02

$$P(H|E) = \frac{0.99 \times 0.001}{0.99 \times 0.001 + 0.02 \times 0.999} = 0.0472$$

If tested positive the 2<sup>nd</sup> time

$$P(H|E) = \frac{P(E|H)P(H)}{P(E|H)P(H) + P(E|\overline{H})P(\overline{H})}$$

P(H) = 0.0472 (Posterior probability: tested positive the 1<sup>st</sup> time)

$$P(\overline{H}) = 0.9528$$

P(E|H): Sensitivity = 0.99

 $P(E|\overline{H})$ : False positive rate = 0.02

$$P(H|E) = \frac{0.99 \times 0.0472}{0.99 \times 0.0472 + 0.02 \times 0.9528} = 0.7103$$

### **THANK YOU**