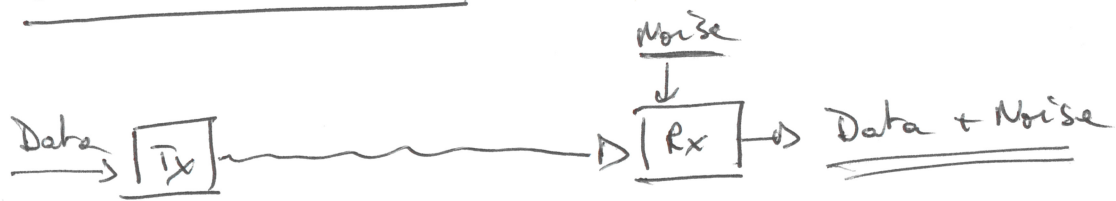
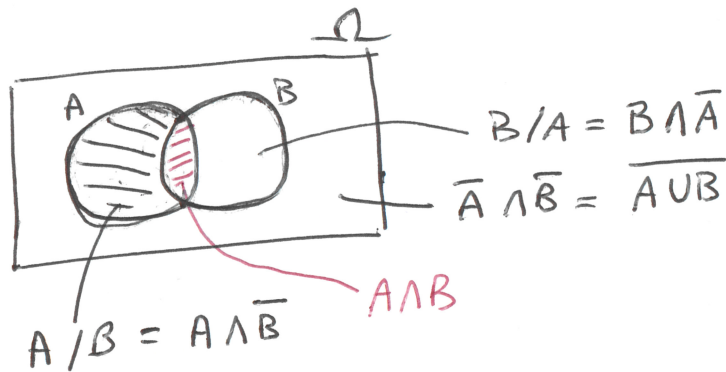


## Lecture 3 and 4



### Set



### disjoint union

$$A = (A \cap \bar{B}) \cup (A \cap B)$$

$$\underline{A \cup B} = B \cup (A \cap \bar{B})$$

### Probability

$$P(E)$$

#### 3 Axioms

$$1) 0 \leq P(E) \leq 1$$

$$2) P(S) = 1$$

$$3) \text{ if } E \cap F = \emptyset, \text{ then } P(E \cup F) = P(E) + P(F)$$

$$\hookrightarrow P(E \cup F) = P(E) + P(F) - P(E \cap F)$$

## Conditional Probability

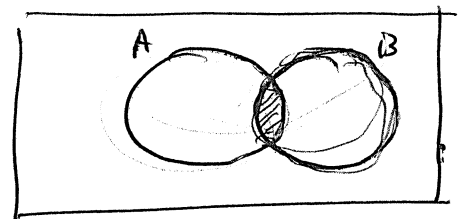
A, B

$$P(A|B) \geq P(A)$$

A pass  
B study

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

when  $P(B) > 0$



•  $B \subseteq A$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(B)}{P(B)} = 1 \geq P(A)$$

•  $A \cap B = \phi$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(\phi)}{P(B)} = 0 \leq P(A)$$

Multiplication  
Law

$$P(A|B)P(B) = P(A \cap B) = P(B \cap A) = P(B|A)P(A)$$

A pass  
B study

if  $P(A) = P(B)$

$$\rightarrow P(A|B) = P(B|A)$$

• A, B independent

$$\boxed{P(A|B) = P(A)}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$\begin{aligned} P(A \cap B) &= P(A|B) P(B) \stackrel{\text{ind.}}{=} \frac{P(A) P(B)}{P(B)} \\ &= P(A) P(B) \Rightarrow \boxed{P(B|A) = P(B)} \end{aligned}$$

• n events mutually independent

$$P(A_1 \cap A_2 \cap \dots \cap A_n) = P(A_1) P(A_2) \dots P(A_n)$$

ex

$$A = \{1, 2, 3\}$$

$$B = \{2, 4\}$$

$$S = \{1, 2, 3, 4, 5, 6\}$$

$$\boxed{P(A) = \frac{3}{6} = \frac{1}{2}}$$

$$P(B) = \frac{2}{6} = \frac{1}{3}$$

$$P(A \cap B) = \frac{1}{6}$$

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{1/6}{2/6} = \frac{1}{2} = P(A)$$

$$P(B|A) = \frac{P(B \cap A)}{P(A)} = \frac{1/6}{3/6} = \frac{1}{3} = P(B)$$

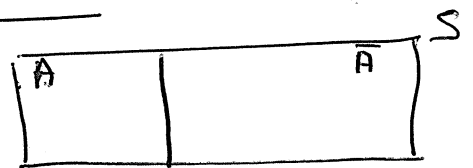
$$A \cup \bar{A} = S$$

$$P(A \cup \bar{A}) = P(A) + P(\bar{A}) = P(S) = 1$$

A: 3                      A: 2

$$P(\bar{A}) = 1 - P(A)$$

$$P(\bar{A} | B) = 1 - P(A | B)$$



$$A \cap \bar{A} = \emptyset$$

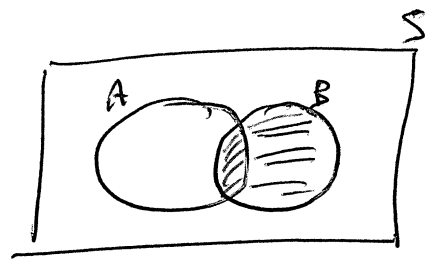
A and  $\bar{A}$  form a partition of S

Show that  $P(\bar{A}|B) = 1 - P(A|B)$

$$B = (B \cap A) \cup (B \cap \bar{A})$$

$$\frac{P(B)}{P(B)} = \frac{P(B \cap A)}{P(B)} + \frac{P(B \cap \bar{A})}{P(B)}$$

$$1 = P(A|B) + P(\bar{A}|B)$$



$$A \cup \bar{A} = S$$

$$B \cup \bar{B} = S$$

$$A = (A \cap B) \cup (A \cap \bar{B})$$

$$B = (B \cap A) \cup (B \cap \bar{A})$$

$$\bar{A} = (\bar{A} \cap B) \cup (\bar{A} \cap \bar{B})$$

$$\bar{B} = (\bar{B} \cap A) \cup (\bar{B} \cap \bar{A})$$

$$P(A) + P(\bar{A}) = 1$$

$$P(B) + P(\bar{B}) = 1$$

$$P(A) = P(A \cap B) + P(A \cap \bar{B})$$

$$P(B) = P(B \cap A) + P(B \cap \bar{A})$$

$$P(\bar{A}) = P(\bar{A} \cap B) + P(\bar{A} \cap \bar{B})$$

$$P(\bar{B}) = P(\bar{B} \cap A) + P(\bar{B} \cap \bar{A})$$

Probability Tables

	A	$\bar{A}$	
B	$P(A \cap B)$	$P(\bar{A} \cap B)$	$P(B)$
$\bar{B}$	$P(A \cap \bar{B})$	$P(\bar{A} \cap \bar{B})$	$P(\bar{B})$
	$P(A)$	$P(\bar{A})$	1

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

ex 100 components two defects A, B

2 components have defect A and B  
 6 " " A only  
 4 " " B only

$A \cap B$   $P(A \cap B) = \frac{2}{100}$   
 $A \cap \bar{B}$   $P(A \cap \bar{B}) = \frac{6}{100}$   
 $B \cap \bar{A}$   $P(B \cap \bar{A}) = \frac{4}{100}$

	A	$\bar{A}$	
B	$\frac{2}{100}$	$\frac{4}{100}$	$\frac{6}{100}$
$\bar{B}$	$\frac{6}{100}$	$\frac{88}{100}$	$\frac{94}{100}$
	$\frac{8}{100}$	$\frac{92}{100}$	1

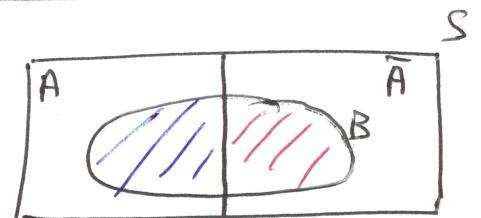
$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{2/100}{6/100} = \frac{1}{3}$$

Total Probability

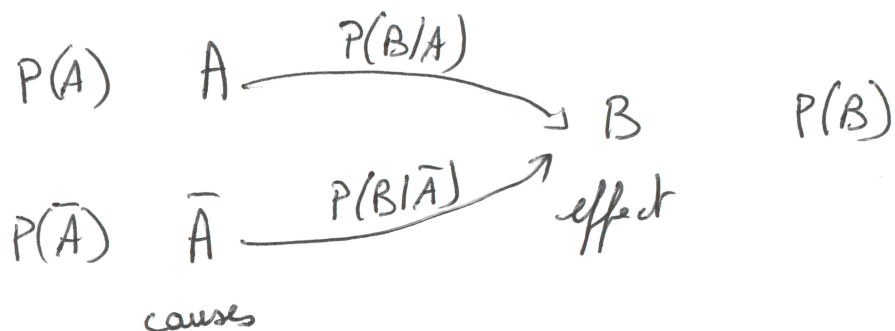
A, B

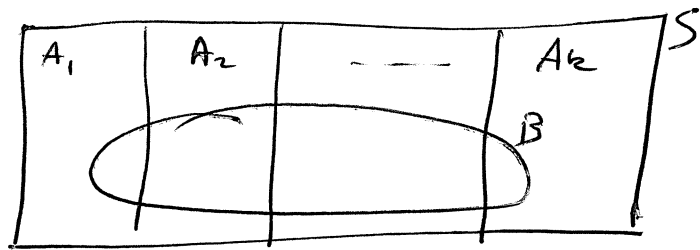
$$B = (B \cap A) \cup (B \cap \bar{A})$$

$$\begin{aligned}
 P(B) &= P(B \cap A) + P(B \cap \bar{A}) \\
 &= P(B|A)P(A) + P(B|\bar{A})P(\bar{A})
 \end{aligned}$$



A and  $\bar{A}$  form a partition of S

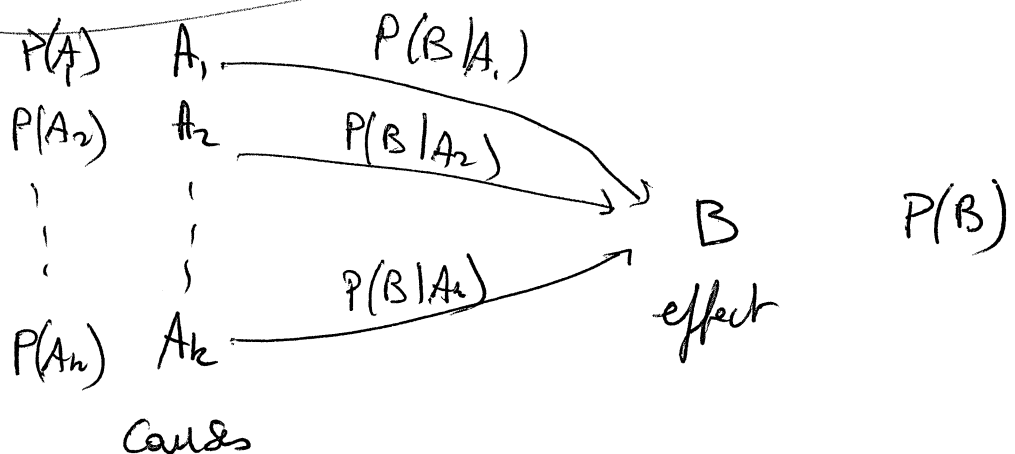




$A_1, A_2, \dots, A_k$  form a partition of  $S$

$$B = (B \cap A_1) \cup (B \cap A_2) \cup \dots \cup (B \cap A_k)$$

$$\begin{aligned} P(B) &= P(B \cap A_1) + P(B \cap A_2) + \dots + P(B \cap A_k) \\ &= P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_k)P(A_k) \\ &= \sum_{i=1}^k P(B|A_i)P(A_i) \end{aligned}$$

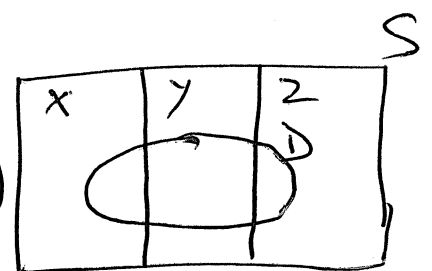


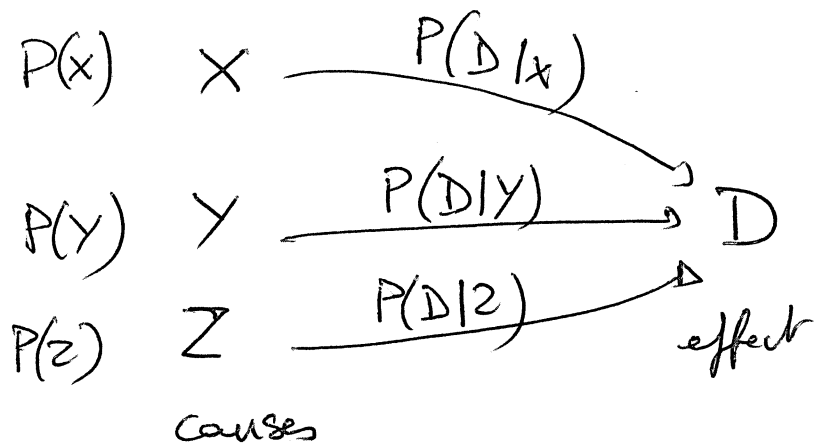
ex 3 machines  $X, Y, Z$

$X$	produces	50%	of the components,	3%	are defective
$Y$	"	30%	" " ,	4%	"
$Z$	"	20%	" " ,	5%	"

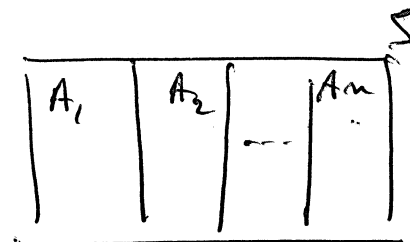
$D$  defective component  $P(D) = ?$

$$\begin{aligned} D &= (D \cap X) \cup (D \cap Y) \cup (D \cap Z) \\ P(D) &= P(D|X)P(X) + P(D|Y)P(Y) + P(D|Z)P(Z) \\ &= 0.03 \times 0.5 + 0.04 \times 0.3 + 0.05 \times 0.2 \end{aligned}$$





$A_1, A_2, \dots, A_m$  partition of  $S$



$$\frac{P(A_k)}{P(B|A_k)} \implies P(B)$$

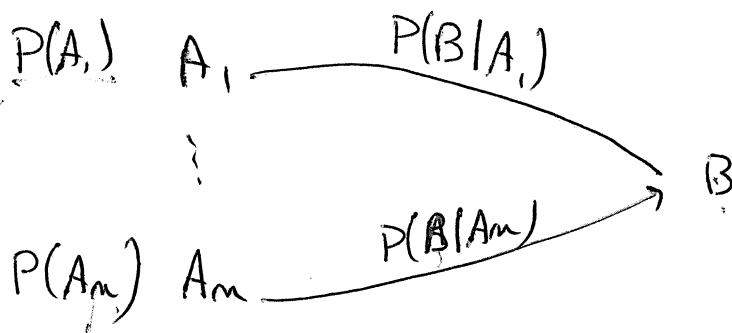
$$P(A_k | B)$$

Bayes Theorem

$$P(A_k \cap B) = P(B|A_k) P(A_k) = P(A_k | B) P(B)$$

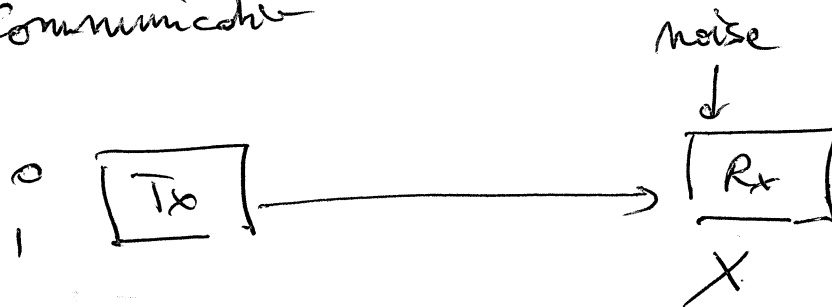
$$P(A_k | B) = \frac{P(B|A_k) P(A_k)}{P(B)}$$

$$P(A_k | B) = \frac{P(B|A_k) P(A_k)}{\sum_{i=1}^m P(B|A_i) P(A_i)}$$



$$P(A_k | B) =$$

ex Communication



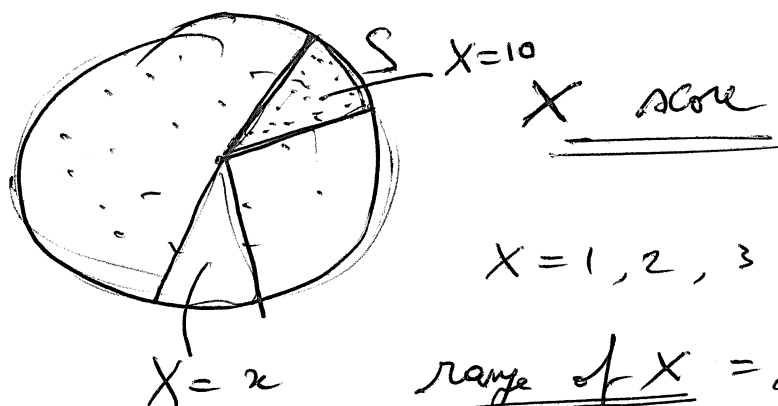
$$P(1|X) = \frac{P(X|1) P(1)}{P(X|0) P(0) + P(X|1) P(1)}$$

Random Variable (RV)

$$X: \underbrace{S} \rightarrow \underline{R}$$

$$X = \underline{x}$$

ex darts



range of X = {1, 2, 3, ..., 60}

Discrete

• Probability mass function (PMF)

$x$	$x_1$	$x_2$	$x_3$	...	$x_m$
$f_X(x)$	$P(X=x_1)$	$P(X=x_2)$	...		$P(X=x_m)$
$P(X=x)$	$f_X(x_1)$	$f_X(x_2)$	...		$f_X(x_m)$
	$p_1$	$p_2$	...		$p_m$

$$\textcircled{f_X}(x)$$

$$P(X=x)$$



## Properties

$$1) \underline{f_X(x_i)} = P(X=x_i) = p_i \geq 0$$

$$2) \sum_{i=1}^{\infty} f_X(x_i) = \sum_{i=1}^{\infty} P(X=x_i) = 1$$

ex

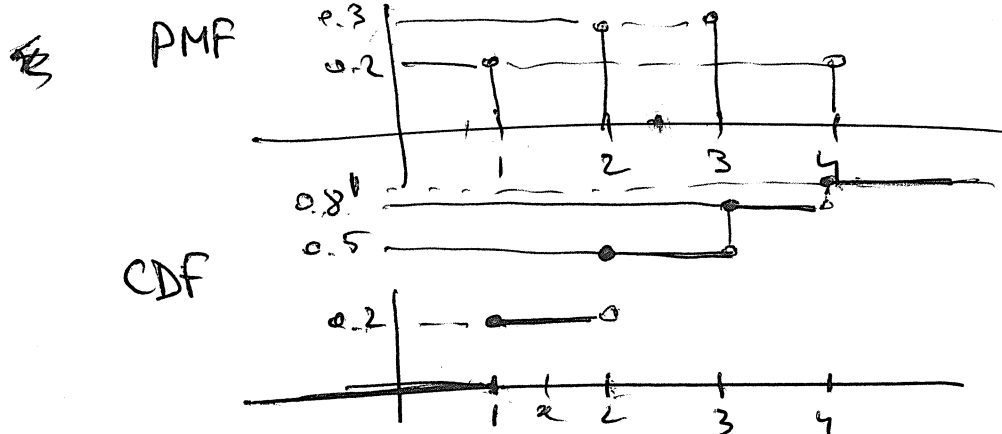
$x$	1	2	3	4
$f_X(x)$	0.2	0.3	0.3	0.2

• Cumulative distribution function (CDF)

$$P(X \leq x)$$

$x_1, x_2, x_3, \dots, x_m$   
in increasing order

$$P(X \leq x_j) = \sum_{i=1}^j P(X=x_i) = \sum_{i=1}^j p_i$$



• Expectation / Theoretical mean  $P(X \leq 2)$

$$\mu = \underline{E(X)} = \sum_x \underline{x} \underline{f_X(x)}$$

$x$	1	2	3	4
$f_X(x)$	0.2	0.3	0.3	0.2

$$\begin{aligned} \mu = E(X) &= 1 \times 0.2 + 2 \times 0.3 \\ &\quad + 3 \times 0.3 + 4 \times 0.2 \\ &= 2.5 \end{aligned}$$

Property

$$\begin{aligned}
 E(ax+b) &= \sum_x (ax+b) f_x(x) \\
 &= \sum_x a x f_x(x) + \sum_x b f_x(x) \\
 &= a \underbrace{\left[ \sum_x x f_x(x) \right]}_{E(X)} + b \underbrace{\left[ \sum_x f_x(x) \right]}_1 \\
 &= \underline{a E(X) + b}
 \end{aligned}$$

$$E(g(x)) = \sum_x g(x) \underline{f_x(x)} \quad \begin{array}{l} g(x) = ax+b \\ g(x) = x^2 \end{array}$$

$$\underline{E(x^2) = \sum_x x^2 f_x(x) \neq [E(x)]^2 = \left[ \sum_x x f_x(x) \right]^2}$$

Variance

$$\sigma^2 = \text{Var}(x) = E([x-\mu]^2) = \sum_x (x-\mu)^2 f_x(x)$$

$\mu = 2.5$  ←

$x$	1	2	3	4
$f_x(x)$	0.2	0.3	0.3	0.2
$f_x(x)$	0.1	0.4	0.4	0.1

$$\begin{aligned}
 \text{Var}(x) &= \sum (x-\mu)^2 f_x(x) \\
 &= (1-2.5)^2 \cdot 0.2 \\
 &\quad + (2-2.5)^2 \cdot 0.3 \\
 &\quad + (3-2.5)^2 \cdot 0.3 \\
 &\quad + (4-2.5)^2 \cdot 0.2
 \end{aligned}$$

$\sigma$  standard deviation

$$\text{Var}(X) = \sum_x (x - \mu)^2 f_X(x) \geq 0$$

$$= \sum_x (x^2 - 2\mu x + \mu^2) f_X(x)$$

$$= \underbrace{\sum_x x^2 f_X(x)}_{E(X^2)} - \underbrace{\sum_x 2\mu x f_X(x)}_{-2\mu \underbrace{\sum_x x f_X(x)}_{E(X)=\mu}} + \underbrace{\sum_x \mu^2 f_X(x)}_{\mu^2 \sum_x f_X(x) = \mu^2}$$

$$= E(X^2) - 2\mu^2 + \mu^2 = E(X^2) - \mu^2$$

$$= E(X^2) - E(X)^2 \geq 0$$

$$\underline{\underline{E(X^2) \geq E(X)^2}}$$

$$\text{Var}(aX+b) = E([ (aX+b) - E(aX+b) ]^2)$$

$$= E([ aX + b - aE(X) - b ]^2)$$

$$= E(a^2 (X - E(X))^2)$$

$$= a^2 \underbrace{E((X - \mu)^2)}_{\text{Var}(X)} = \underline{\underline{a^2 \text{Var}(X)}}$$

function of  $a$ !  
not a function  
of  $b$ !

$$\sum_x a^2 (x - \mu)^2 f_X(x)$$

$$= a^2 \underbrace{\sum_x (x - \mu)^2 f_X(x)}_{\text{Var}(X)}$$