

Yesterday

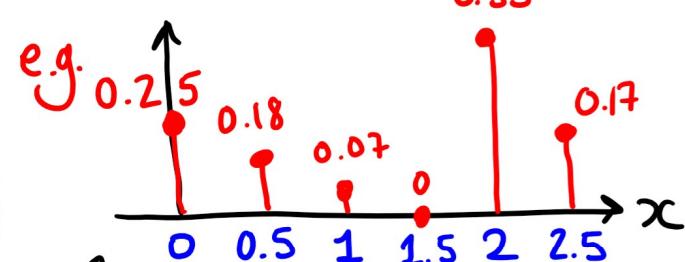
Probability Mass Functions (p.m.f.) $f(x)$

& Cumulative Distribution Functions (c.d.f.) $F(x)$

→ For discrete random variables.

pmf $f(x) = P(X = x)$; $\sum_x f(x) = 1 \text{ & } f(x) \geq 0$.

cdf $F(x) = P(X \leq x)$

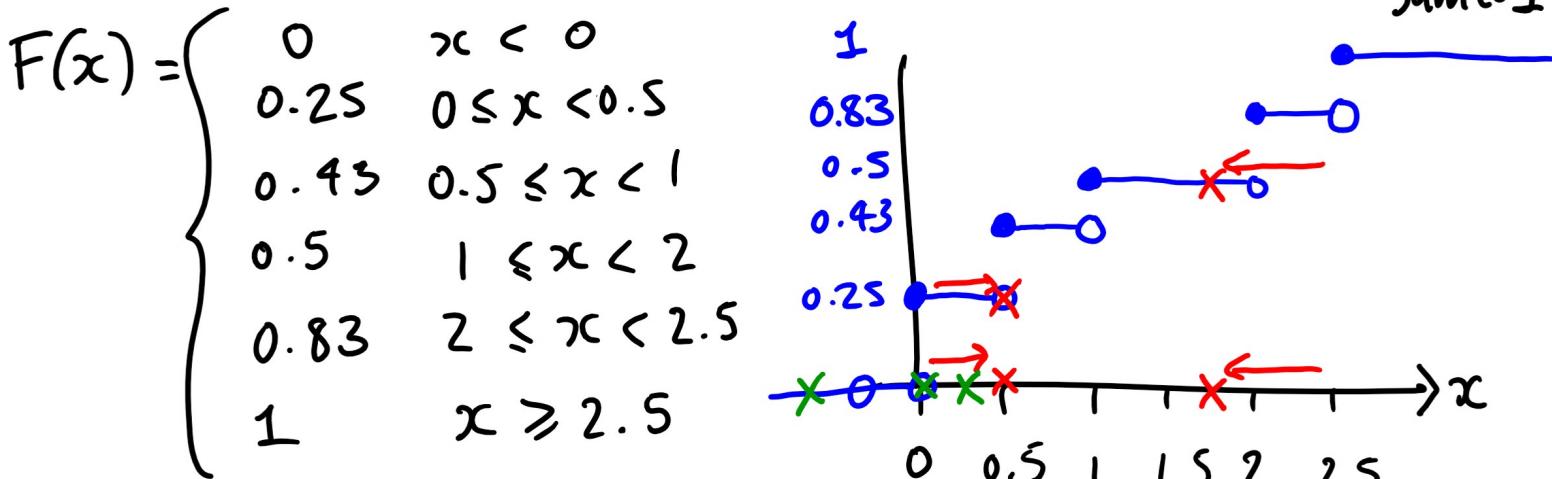


Example

p.m.f. $f(x)$ given by:

x	0	0.5	1	1.5	2	2.5
$f(x)$	0.25	0.18	0.07	0	0.33	0.17

c.d.f.



Notice : $\int F(x)$ starts at 0 & ends at 1

$F(x)$ is non-decreasing ($x_1 \leq x_2 \Rightarrow f(x_1) \leq f(x_2)$)

$F(x)$ is
Right continuous everywhere
Jumps at x are equal to p.m.f. at x

So from any function $F(x)$ with this list of properties we can describe corresponding pmf $f(x)$ of a

discrete r.v. : $P(X=a) \rightarrow f(a) = F(a) - \lim_{x \rightarrow a^-} F(x)$ ← jumps described formally like this

$$\text{In example above } f(0) = 0.25 - 0 = 0.25$$

$$\text{For } a < b \quad P(a \leq X \leq b) = F(b) - \lim_{x \rightarrow a^-} F(x).$$

Example Given $F(x)$ as follows, find $P(X=1)$,
 $P(1 \leq X \leq 2)$
 $\& P(X > 2)$.

x	$x < 0$	$0 \leq x < 1$	$1 \leq x < 2$	$2 \leq x < 3$	$x \geq 3$
$F(x)$	0	0.1	0.4	0.65	1

$$(a) P(X=1) = F(1) - \lim_{x \rightarrow 1^-} F(x) = F(1) - F(0) \\ = 0.4 - 0.1 \\ = 0.3$$

$$(b) P(1 \leq X \leq 2) = F(2) - \lim_{x \rightarrow 1^-} F(x) = F(2) - F(0) \\ = 0.65 - 0.1 \\ = 0.55$$

$$(c) P(X > 2) = 1 - P(X \leq 2) = 1 - F(2) \\ = 1 - 0.65 = 0.35.$$

3-4 Mean & Variance of a Discrete Random Variable

Weighted average Spread

→ summarise the prob. distribution

Mean "centre" "balance point"

\hookrightarrow where prob. masses would balance

"weighted average"
with probabilities as weights.
 $\hookrightarrow f(x)$ values

Denoted $\mu = E(X) = \sum_x x f(x)$

expected value of X

(we expect
 x to occur
"in the long
run" a
fraction $f(x)$
of the time)

Variance "Spread" "dispersion"

Denoted $\sigma^2 = V(X) = E((X-\mu)^2)$
 $= \sum_x (x-\mu)^2 f(x)$

Standard

Deviation $\sigma = \sqrt{V(X)}$

\hookleftarrow units match those
of X & μ

Alternate formula for Variance

$$\sigma^2 = \sum_x (x - \mu)^2 f(x)$$

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

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

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

$$= \sum_x x^2 f(x) - 2\mu^2 + \cancel{\mu^2} \\ - \cancel{\mu^2}$$

Warning 2 prob. distributions may have same μ & σ^2 (so cannot recover p.m.f. $f(x)$ from μ & σ^2)

See figure 3-6 p. 74 of text.

Example p.m.f. $f(x)$; X represents change in price in \$

x	-2	-1	0	1	2	3
$f(x)$	0.1	0.05	0.13	0.6	0.1	0.02

Calculate μ , σ^2 and σ .

$$\begin{aligned}\mu = E(X) &= \sum_x x \cdot f(x) = (-2)(0.1) + (-1)(0.05) + \\ &\quad 0(0.13) + (1)(0.6) + 2(0.1) \\ &\quad + 3(0.02) \\ &= \$0.61\end{aligned}$$

$$\begin{aligned}\sigma^2 = V(X) &= \sum_x x^2 f(x) - \mu^2 \\ &= (-2)^2(0.1) + (-1)^2(0.05) + 0^2(0.13) + 1^2(0.6) \\ &\quad + 2^2(0.1) + 3^2(0.02) - (0.61)^2 \\ &= 4(0.1) + 0.05 + 0 + 0.6 + 4(0.1) + 9(0.02) \\ &\quad - (0.61)^2\end{aligned}$$

$$= 1.2579 \text{ } \2$

$$\sigma = +\sqrt{1.2579} = \$1.12.$$

This was added
after class due
to a technological
failure preventing
the example from being
completed in class time.