

# 3703-3J04 PROBABILITY & STATISTICS FOR CO1 - Lecture 8 (CIVIL) ENGINEERING

Last time PROBABILITY MASS FUNCTION (p.m.f.):  
 $f(x) = P(X=x)$

CUMULATIVE DISTRIBUTION FUNCTION (c.d.f.):  
 $F(x) = P(X \leq x)$

MEAN:  $\mu = E(X) = \sum_x x f(x)$

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

Example  $X =$  price difference in \$ with p.m.f.  $f(x)$

given by:

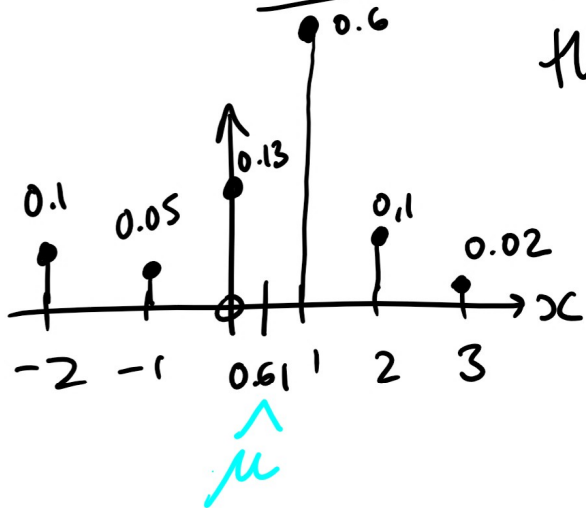
$x$	-2	-1	0	1	2	3
$f(x)$	0.1	0.05	0.13	0.6	0.1	0.02

Determine  $\mu$ ,  $\sigma^2$ ,  $\sigma$   $\leftarrow$  standard deviation.

Solution  $\mu = \sum_x x f(x) = (-2)(0.1) + (-1)(0.05) + 0$   
 $+ (1)(0.6) + (2)(0.1) + 3(0.02)$

$$= \underline{\underline{\$0.61}}$$

Notice  $X$  never takes value 0.61 but this is still the mean.



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

$$= 4(0.1) + 1(0.05) + 0 + 1(0.6) + 4(0.1) + 9(0.02) - (0.61)^2$$

$$= \underline{\underline{1.2579}} \text{ \$}^2$$

$$\sigma = \sqrt{V(X)} = \sqrt{1.2579} = \underline{\underline{\$1.12}}$$

In general, the expected value of any function  $h(X)$  of our discrete r.v.  $X$  with p.m.f.  $f(x)$  is given by

$$E(h(X)) = \sum_x h(x) f(x).$$

Notice  $h(X)$  is itself a <sup>(discrete)</sup> r.v.  $Y$  with its own

p.m.f.  $f_h(y)$ . So  $E(Y) = \sum_y y f_h(y)$ .

It can be shown that these give same answer.

Also notice: Our alt. formula for variance is

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

$$\Rightarrow \boxed{\sigma^2 = V(X) = E(X^2) - [E(X)]^2}$$

Observation If  $a, b$  are constants, then

$$(1) E(aX + b) = aE(X) + b$$

$$(2) V(aX + b) = a^2 V(X)$$

(Scaling  $X$  by  $a$  scales the mean, and shifting by  $b$  shifts it by  $b$ .)

(Scaling  $X$  by  $a$  scales variances by  $a^2$  but shifting by  $b$  does not change how spread out the distribution is.)

## Special Probability Distributions

### [3.6] Binomial Distribution

- Some number  $n$  of independent "trials"  
(single experiment that can be repeated)

→ each trial has 2 mutually exclusive possible outcomes (e.g. 0 or 1, arrival or not, Heads or Tails, win or loss, yes or no, faulty or not, success or failure)

↳ This is called a Bernoulli trial

→ probability of success is  $p$  (same every time)

$X = \#$  successes is called a binomial random variable with parameters  $(n, p)$

[and has a binomial (probability) distribution]  
and has p.m.f.  $f(x) = \binom{n}{x} p^x (1-p)^{n-x}$   
for each  $x = 0, \dots, n$ .

$C_x^n$ : # sequences with  $x$  successes &  $n-x$  failures from  $n$  trials

Example Multiple Choice Test with  $n = 6$  questions

5 options for each question; randomly guess answers

So  $p = 0.2 = \frac{1}{5}$ .  $X = \#$  correct answers;

$$P(X=0) = f(0) = \binom{6}{0} (0.2)^0 (0.8)^6 = 0.26$$

$$P(X=1) = f(1) = \binom{6}{1} (0.2)^1 (0.8)^5 = 0.39$$

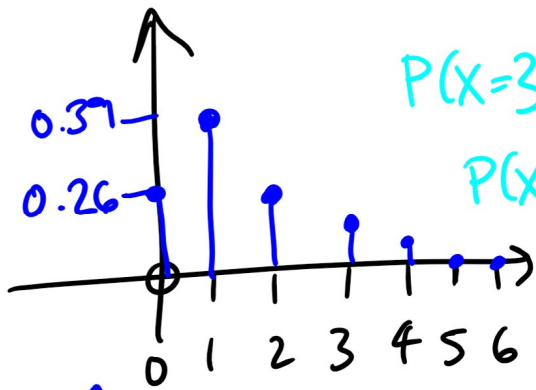
$$P(X=2) = f(2) = \binom{6}{2} (0.2)^2 (0.8)^4 = 0.25$$

$$P(X=3) = f(3) = \binom{6}{3} (0.2)^3 (0.8)^3 = 0.08$$

$$P(X=4) = f(4) = \binom{6}{4} (0.2)^4 (0.8)^2 = 0.02$$

$$P(X=5) = f(5) = \binom{6}{5} (0.2)^5 (0.8)^1 = 0.00$$

$$P(X=6) = f(6) = \binom{6}{6} (0.2)^6 (0.8)^0 = 0.00$$



Notice the shape



Notice that for binomial distribution ← prob. distr. of a bin. r.v.

the c.d.f. ( $P(X \leq x_0) = F(x_0) = \sum_{x \leq x_0} f(x)$ )

$$= \sum_{x \leq x_0} \binom{n}{x} p^x (1-p)^{n-x}$$

Example In the above example,

find  $P(X=4)$ ,  $P(X \leq 3)$ ,  $P(2 \leq X \leq 5)$   
and  $P(X \geq 3)$

Solution

$$P(X=4) = f(4) = \underline{\underline{0.02}}$$

$$P(X \leq 3) = \sum_{x \leq 3} f(x) = f(0) + \dots + f(3)$$

$$P(2 \leq X \leq 5) = \sum_{x=2}^5 f(x) = \underline{\underline{0.35}} = \underline{\underline{0.98}}$$

$$P(X \geq 3) = 1 - P(X \leq 2) = 1 - (f(0) + f(1) + f(2))$$

↑ We could also take  $\sum_{x=3}^6 f(x) = f(3) + f(4) + f(5) + f(6)$ .

(Know that both options are available so that you can do whichever is easiest.!) )