

Cumulative Probabilities

- alternative way of describing probabilities associated with a random variable

some value

$$P(X \leq \underset{\downarrow}{x})$$

- "up to and including" some value of X , rather than the probability of an exact value

$$= \sum_{x_i \leq x} f(x_i)$$

Sampling with replacement: item is selected;
Placed back into the batch (randomly!);
next selection has same probability distribution
as previous selection .. independent

Sampling without replacement : item is selected;

not placed back in; probabilities associated with next selection depend on outcome of previous selection
.. Not independent !!

ex: 850 manufactured parts
.. known: 50 contain flaws

- assembly requires two parts, selected at random without replacement
- let \times : non-conforming parts
- write a cumulative distribution for the # of non-conforming parts in an assembly

First: write sample space of outcomes!

$$S \{ gg \quad gb \quad bg \quad bb \}$$

- now let's determine probabilities associated with these outcomes

$$P\{gg\} = \frac{800}{850} \cdot \frac{799}{849}$$

↑ good parts ↑ good parts left after 1st selection
total parts parts left after first selection!
 $= 0.886 \text{ or } 88.6\%$

$$P\{gb\} = \frac{800}{850} \cdot \frac{50}{849} = 0.05543$$

↑
parts left or 5.5%

$$P\{bg\} = \frac{50}{850} \cdot \frac{800}{849} = 0.05543$$

↑
all 800 good ones
still there

$$P\{bb\} = \frac{50}{850} \cdot \frac{49}{849} = 0.003$$

↑
bad ones left
or 0.3%

now we can put together a probability distribution

↳ possible values of X



$$S\{0, 1, 2\}$$

$$P(0) = P\{gg\} = 0.886$$

$$P(1) = P\{gb, bg\} = 0.05543 + 0.05543 = 0.111$$

↑
exactly

$$P(2) = P\{bb\} = 0.003$$

Non-conforming
parts ↓

Now we compute cumulative probabilities

$$P(X \leq x) \quad (\text{up to and including})$$

$$= F(x)$$

↑ capital F notation

$$F(0) = P(\underbrace{X \leq 0}_{\text{up to one bad one}}) = P(0) = \underline{0.886}$$

$$F(1) = P(X \leq 1) = P(0) + P(1) = 0.886 + 0.111$$

$$F(2) = P(X \leq 2) = P(0) + P(1) + P(2) = \overbrace{1}^{\sim} \quad - 0.997$$

or $\overbrace{100\%}^{\sim}$

Mean and Variance of Discrete Random Variable

- We already know mean and variance as sample parameters \bar{x} and s^2
- We can think of population mean μ^2 and variance σ^2 in a new way

define mean or expected value of discrete random variable X :

$$\mu = E(X) = \sum x f(x)$$

Weighted sum of possible values of X
↳ by respective probabilities

define Variance of X :

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

Weighted sum of squares of errors

~ quantifies dispersion or scatter in data

Computational formula:

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

not in summation!!!

ex: camera problem

recall probability distribution: $P(0) = 0.008$

$$P(1) = 0.096$$

$$P(2) = 0.384$$

$$P(3) = 0.512$$

~ the expected value of the # of passing cameras

= \bar{X}

$$\mu = E(X) = \sum x f(x)$$

$$= 0 \cdot 0.008 + 1 \cdot 0.096 + 2 \cdot 0.384 + 3 \cdot 0.512$$

$E(X) = 2.4$ passing cameras

.. it is perfectly okay that 2.4 passing cameras is not a possible outcome in the discrete sample space!

$$\begin{aligned}\sigma^2 &= V(X) = \sum x^2 f(x) - \mu^2 \\ &= 0^2 \cdot 0.008 + 1^2 \cdot 0.096 + 2^2 \cdot 0.384 + 3^2 \cdot 0.512 \\ &\quad - 2.4^2 \\ V(X) &= 0.48 \text{ (passing cameras)}^2\end{aligned}$$

$$\sigma = +\sqrt{\sigma^2} = \sqrt{0.48} = \underline{0.6928 \text{ passing cameras}}$$