

P-value is a conditional distribution " $P(\text{data}/H_0)$ "

Stat 88: Prob. & Mathematical Statistics in Data Science



<https://xkcd.com/892/>

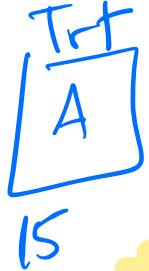
Lecture 33 : 4/15/2024 (PART 1)

Section 9.3, 9.4

A/B testing & ~~confidence intervals~~

A/B testing: comparing two distributions

- Data 8, section 12.3, randomized controlled trial to see if botulinum toxin could help manage chronic pain.
- 31 patients → 15 in treatment group, 16 in control group. 2 patients in the control group reported pain relief and 9 in the treatment group.
- A/B testing is a term used to describe hypothesis tests which involve comparing the distributions of two random samples. (Earlier we had one sample and made a hypothesis about its distribution.)
- In particular, we can conduct an A/B test for hypothesis tests involving results of randomized controlled trials, A is the control group and B the treatment group.

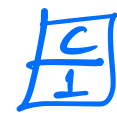


31 tickets



tkt 1

tkt 2



← orig.

tkt 31

Fisher's exact test

- Control group: 16 patients, 2 reported relief
- Treatment group: 15 patients, 9 reported relief
- H_0 : The treatment has no effect (there would have been 11 patients reporting pain relief no matter what, and it just so happens that 9 of them were in the treatment group)
- H_A : The treatment has an effect

Total # of patients reporting pain relief is 11 (out of 31)

Perm test shuffle labels & re compute diff. in prop.

Test statistic be the # of patients that reported pain relief in sample (Treatment)

If H_0 is true then X is just the # of successes in a SRS of size 15 out of 31 total with 11 successes in pop. of 31.

$X \sim HG(N=31, G=11, n=15)$ $E(X|H_0) = \frac{15 \cdot 11}{31} \approx 5.32$

4/12/24

using H_0 .

Now let's use test statistic $T = X - 5.32$, observed value of $T = 9 - 5.32 = 3.68$

$$\begin{aligned}
 \text{P-value} &= P(|T| \geq 3.68) = P(X \leq 5.32 - 3.68) \\
 &\quad + P(X \geq 5.32 + 3.68) \\
 \text{obs. diff to } 3.68 & \\
 &= P(X \leq 1.64) + P(X \geq 9)
 \end{aligned}$$

$$\begin{aligned}
 &= P(X \leq 1) + P(X \geq 9) \\
 &\approx 0.00915
 \end{aligned}$$

Reject H_0 ; the trt (appears) to have an effect.

Exercise for multiple chocolates

Example: The Lady Tasting Tea

- The first person to describe this sort of hypothesis test was the famous British statistician Ronald Fisher. In his book *The Design of Experiments*, he describes a tea party in which a lady of his acquaintance claimed that she could tell from tasting a cup of tea if the milk had been poured first or the tea.
- Fisher immediately set up an experiment in which she was given multiple cups of tea and asked to identify which of them had had the tea poured first. She tasted 8 cups of tea, of which 4 had the tea poured first, and identified 3 of them correctly. Does this data support her claim?

		Actually happened	
		Tea first	Milk first
Lady says	Tea first		
	Milk first		

Exercise

Example: Gender bias ?

- Rosen and Jerdee conducted several experiments using male bank supervisors (this was in 1974) who were given a personnel file and asked to decide whether to promote or hold the file. 24 were randomly assigned to a file labeled as that of a male employee and 24 to a female.
- 21 of the 24 males were promoted, and 14 of the females. Is there evidence of gender bias?

Moving on to Chapter 10

Chapter 10: Probability Density

Earlier we saw discrete random variables that took ^{particular} discrete values on \mathbb{R} and had discrete prob. mass functions.

Now instead of restricting the r.v. to particular values, we allow it to take any value in a given interval.

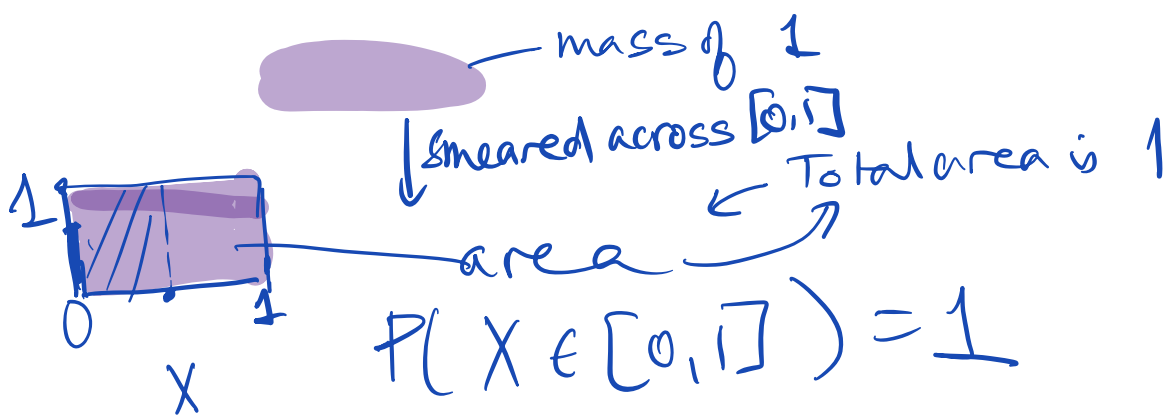
For example we might have a r.v. X s.t. X is in $[0, 1]$ ($X \in [0, 1] \rightarrow 0 \leq X \leq 1$)

Instead of a prob. mass function, we will define a PROBABILITY DENSITY function.

Imagine the total prob. mass of 1 is smeared over the interval according to the density function. Such a r.v. whose probabilities are defined this way is called a

CONTINUOUS RANDOM VARIABLE

& the probabilities are computed by areas.

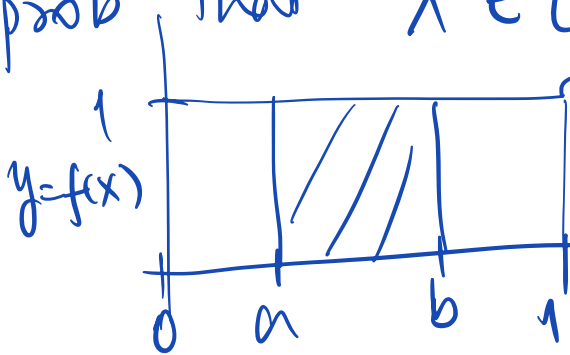


$$P(X \in [0, \frac{1}{2}]) = \frac{1}{2}$$

If the density function is flat,

We call X a uniform r.v. & the

prob that $X \in [a, b]$ is just the area of rectangle whose base is interval $[a, b]$



$$P(X \in [a, b]) = b - a$$

$$P(X \in [0, \frac{1}{2}]) = \frac{1}{2}$$

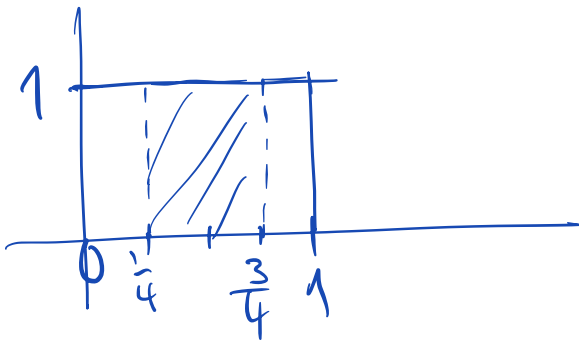
$$P(X \in [\frac{1}{4}, \frac{3}{4}]) = \frac{1}{2}$$

$$P(X \in [\frac{1}{2}, 1]) = \frac{1}{2}$$

If X is a uniform r.v. (density function is flat)

then the only thing that matters is width of the interval, not where it is in $[0, 1]$

$$P\left(\frac{1}{4} < X < \frac{3}{4}\right) = P\left(X \in \left(\frac{1}{4}, \frac{3}{4}\right)\right) \\ = \frac{1}{2} = P\left(X \in \left[\frac{1}{4}, \frac{3}{4}\right]\right)$$

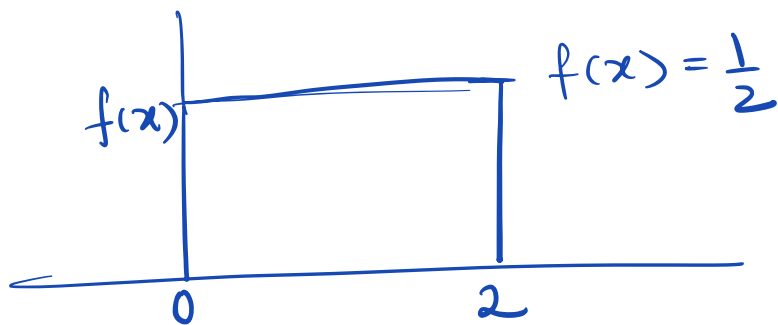


Define a prob. density function $f(x)$ by

$$\textcircled{1} f(x) \geq 0$$

$$\textcircled{2} \int_{-\infty}^{\infty} f(x) dx = 1$$

Cdf is defined to be $F(x) = \int_{-\infty}^x f(t) dt$



$$f(x) = \begin{cases} \frac{1}{2}, & 0 \leq x \leq 2 \\ 0 & \text{o/w} \end{cases}$$

$$F(x) = \int_{-\infty}^x f(t) dt = \int_0^x \frac{1}{2} dt$$

$$P(1.2) = P(X \leq 1.2) = \int_{-\infty}^{1.2} \frac{1}{2} dt = \int_0^{1.2} \frac{1}{2} dt = \frac{1.2}{2} = 0.6$$