

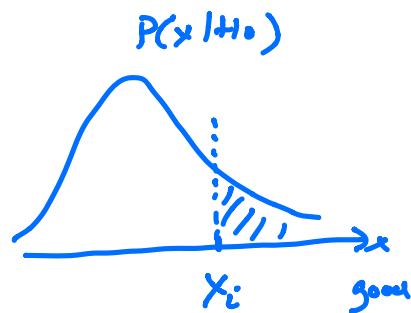
W04 - p-values, significance. The Student's t-test

Experiments - $X \rightarrow \text{Data } \{x_i\} = D$

H_1 - interesting hypothesis about D , that you want to test/validate

H_0 - null hypothesis that "could" also explain your results

H_a is hard to formulate
 H_1 is easier to formulate



take one value x_i :

$$\begin{aligned} \text{p-value}(x_i) &= P(x \geq x_i | H_0) && \text{(right tail)} \\ &= 1 - P(x < x_i | H_0) \\ &= 1 - \text{CDF}_{H_0}(x_i) \end{aligned}$$

$$\text{p-value}(x_i) = P(x \leq x_i | H_0) \quad \text{left tail}$$

$$\text{p-value}(x_i) = P(x \geq x_i \wedge -x_i \leq x | H_0) \quad \text{two tailed test}$$

If p-val is small, there is little chance that H_0 describes the data.

p-value does not use any info about the other hypothesis H_1
p-value rejecting H_0 does not mean H_1 is true

Example experiment failure rate
 $H_0 : f = f_0 = 0.2$ what pstat doc told you
 $H_1 : f > f_0$

$$D = \{ ssffff \}$$

$$\begin{aligned} p\text{-value}(ssffff)_{H_0} &= P(ssffff \mid f=f_0) \\ &\quad + P(sfffff \mid f=f_0) \\ &\quad + P(ffffff \mid f=f_0) \\ &= \frac{5!}{2!3!} f_0^3 (1-f_0)^2 + \frac{5!}{1!4!} f_0^4 (1-f_0) + \frac{5!}{0!5!} f_0^5 \\ &= 0.05792 \\ &\quad (f_0=0.2) \end{aligned}$$

What do you do with this result

$$p\text{-val}(ssffff)_{H_0} = 0.05792$$

do you reject $f=f_0$ or not?

The Bayesian Approach

$$\frac{P(H_0 | ssffff)}{P(H_1 | ssffff)} = \frac{P(ssffff | H_0) P(H_0)}{P(ssffff | H_1) P(H_1)} \quad |_{P(H_0) = P(H_1)}$$

$$= \frac{\int_0^1 df P(ssffff | f) P(f | H_0) df}{\int_0^1 df P(ssffff | f) P(f | H_1) df}$$

$$P(f | H_0) = \begin{cases} \frac{1}{0.21 - 0.19} & 0.19 < f < 0.21 \approx |f - f_0| < 0.1 \\ 0 & \text{otherwise} \end{cases}$$

$$P(f | H_1) = \begin{cases} \frac{1}{1 - 0.2} & f > f_0 = 0.2 \\ 0 & \text{otherwise} \end{cases}$$

$$\frac{P(H_0 | ssffff)}{P(H_1 | ssffff)} = \frac{\frac{1}{0.02} \int_{0.19}^{0.21} P(ssffff | f) df}{\frac{1}{0.80} \int_{0.2}^1 P(ssffff | f) df}$$

$$= \frac{\frac{1}{0.02} \int_{0.19}^{0.21} f^3 (1-f)^2 df}{\frac{1}{0.80} \int_{0.2}^1 f^3 (1-f)^2 df} \approx \frac{20}{80}$$

{ 80:20 chance $f > f_0 \}$

Which method do you prefer?



- * p-value = 0.0579
 - nothing about H_1
 - easier but hard to interpret
- $p_{\text{val}} \neq P(H_0 \text{ is true})$
- $1 - p_{\text{val}} \neq P(H_1 \text{ is true})$

* Bayes

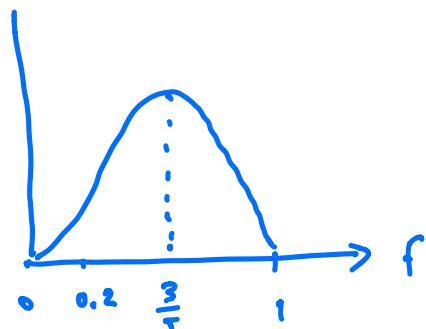
+ model comparison

$$H_1 : H_0 \\ 80 \quad 20$$

+ posteriors

$$P(f \mid \text{ssffff})$$

$$f^* = \frac{n}{N}$$

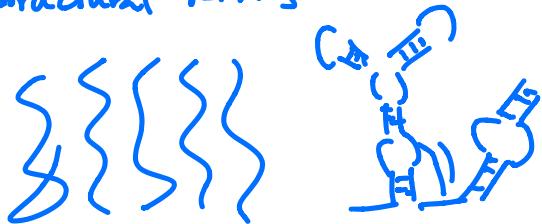


p-value is indirect

If I were to repeat a $N=5$ experiment many times and $f=f_0$, there will be a ~ 620 chance of obtaining at least 3 failures (ssfft or stffff or ttttt)

Student's t-test a widely used p-value

structural RNAs



RNase P RNA
(ribozyme)

265 residues

160 pairs (80 bp)

103 unpaired

base pairs
A:U U:A
C:G G:C
G:U U:G

Given the RNase P sequence, what is the structure?

Experimental chemical modification reactivities

r = reactivity ($0 \leq r \leq 1$)

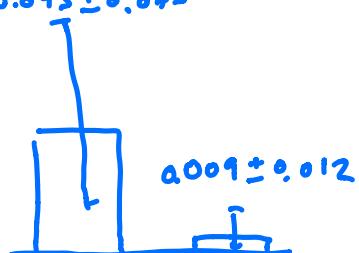
r measures "flexibility": is this a synonym of being paired/unpaired?

DNS, SHAPE (selective 2'-hydroxylation analyzed by -Seq + primer extension)

$$D = \{r_1, \dots, r_{265}\}$$

$$D_u = \{103\ r's\}$$

$$D_p = \{165\ r's\}$$



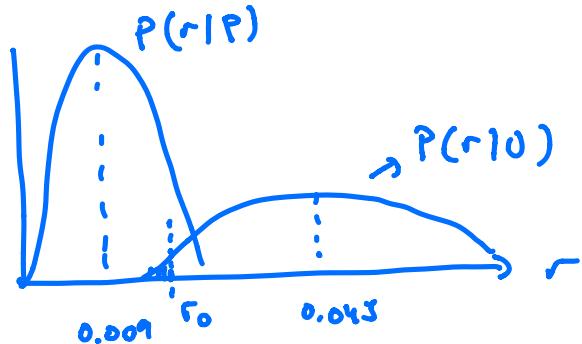
Student-t test
(D_u, D_p)

$$p\text{-val} = 4.3 \cdot 10^{-9}$$

Great! SHAPE should be enough to distinguish P/U

① Look at the data and assumptions

Ideal scenario



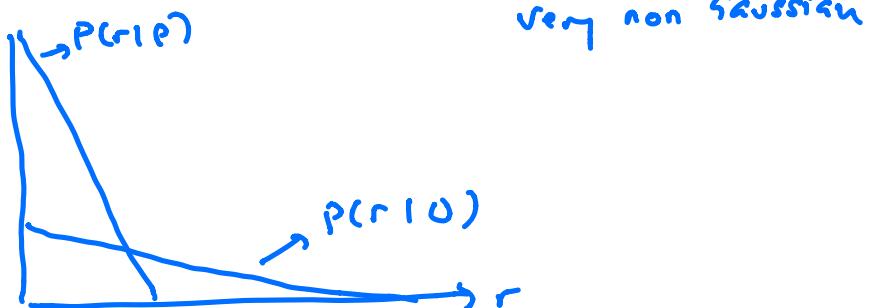
$r < r_0$ call it P

$r > r_0$ call it O

$$p\text{-val}(r_0)_0 = P(r < r_0 | O) = \text{small} \quad (\text{False Positives})$$

$$P(r < r_0 | P) \approx 0.9 \quad (\text{True Positives})$$

Real data



T-test assumes data is Gaussian!
and it uses the means only

What does the Student's distribution have to do

with anything? using 2.3, 3.2

assume $\{x_i\}$ follows a Gaussian dist $\mathcal{N}(\mu, \sigma)$

$$P(\mu, \sigma | \{x_i\}) \propto P(\{x_i\} | \mu, \sigma) P(\mu, \sigma)$$
$$= \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right)^N e^{-\frac{\sum(x_i - \mu)^2}{2\sigma^2}} P(\mu) P(\sigma)$$

$$P(\mu) = \begin{cases} \frac{1}{\mu + \mu_0} & \mu < \mu_0 \\ 0 & \text{otherwise} \end{cases}$$

$$P(\sigma) = \begin{cases} C & \sigma > 0 \\ 0 & \sigma < 0 \end{cases}$$
$$e^{-\frac{\sum(x_i - \mu)^2}{2\sigma^2}}$$

$$P(\mu, \sigma | \{x_i\}) \propto \sigma^{-N} e^{-\frac{\sum(x_i - \mu)^2}{2\sigma^2}}$$

$$P(\sigma | \{x_i\}) \propto \int_0^\infty d\sigma \sigma^{-N} e^{-\frac{\sum(x_i - \mu)^2}{2\sigma^2}}$$

$$t = \frac{\sqrt{\sum(x_i - \mu)^2}}{\sigma} \quad d\sigma = -\frac{\sqrt{\sum(x_i - \mu)^2}}{t^2} dt$$

$$P(\sigma | \{x_i\}) \propto \left[\sum(x_i - \mu)^2 \right]^{\frac{-N+1}{2}} \int_0^\infty t^{N-2} e^{-t^2/2} dt$$

Student's distribution.

IS this p-value that compare the means

really what we want to know?

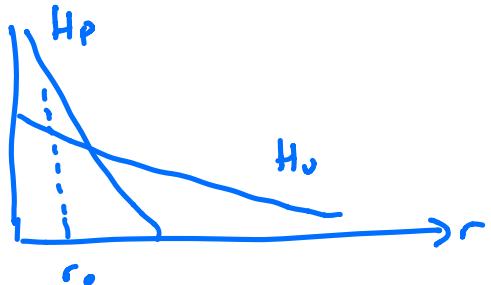
$$\mu_1 = 0.045$$

$$t\text{test} \propto C(\mu_1 - \mu_2)^2$$

$$\mu_2 = 0.009$$

\hookrightarrow follows Student's dist

what I want to know is
if I make the assumption



$r < r_0$ is P

$r > r_0$ is U

What errors do I make?

$$p\text{-value}(r_0) = P(r \leq r_0 | H_0) = CDF_{H_0}(r_0) \quad \text{prob that a random sample has } r \leq r_0$$

It tests

$$n \sim n.pval(r_0) = \text{Expected # FP if all ns follow null}$$

Then pick p-value based on how many FP you

are willing to tolerate

reject	p-value	
0.0029	0.02	2% FP
0.0034	0.05	5% FP
0.0042	0.10	10% FP

N = # of test

\hat{p}^* = p-value at r^*

F^* = # of N w/ $r \leq r^*$

T^* = $T \cap F^*$

$T = 165$ (true)

$$FDR = \frac{\hat{p}^* \cdot N}{F^*}$$

fraction of positive calls
that should be corrected
to be U

p^* fraction of tested (N) expected to be false calls
fdr fraction of results (F^*) expected to be false calls

r	p^*	fdr	sen
0.0029	0.2	17.4	16.9
0.0034	0.05	30.0	23.1
0.0042	0.10	44.8	30.6
0.0063	0.28	66.5	50.0

to get 50% of real paired bases, more than half
of the called paired are going to be wrong