

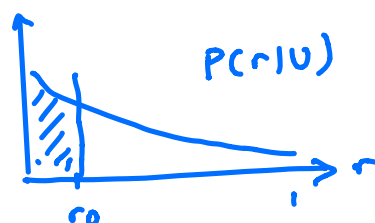
Using p-values to estimate false positives

RNA chemical modifications reactivity (SHAPE, DMS)

$0 \leq r \leq 1$  reactivity

$$H_1: \begin{cases} r \leq r_0 & \text{residue is P} \\ r > r_0 & \text{" is U} \end{cases}$$

$H_0$ : residues are Unpaired



← obtained by getting the reactivities of many residues that are U

$$p\text{-value}(r_0)_{H_0} = P(r \leq r_0 | U) = CDF_{H_0}(r_0)$$

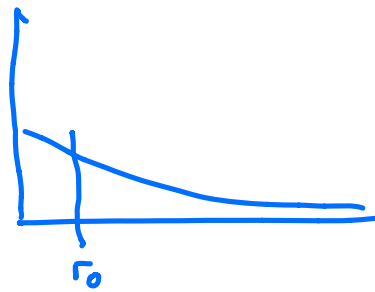
reactivity $r_0$	p-value	2% chance an Un has reactivity $\leq 0.0029$	expected FP
0.0029	0.02		N 0.02
0.0034	0.05		N 0.05
0.0042	0.10		N 0.10

If I test N residues  $\rightarrow$  n x pval expected False Positives  
if all tested were derived from the null hypothesis (were unpaired)

Test N residue S

$F^0$  called P  
( $r \leq r_0$ )

$$pral^0 = P(r \leq r_0 | U)$$



\*  $pral^0$  is the fraction of N expected to be FP

$$FP^0 = pral^0 \cdot N$$

False Discovery Rate  $FDR^0 = \frac{FP^0}{F^0} \Rightarrow$  fraction of  $F^0$  expected to be FP

\*  $FDR^0$  is the fraction of  $F^0$  expected to be FP

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• If you are testing an RNA w/ known structure

$$N = 265$$

$$T = 160 (P)$$

$T =$  fraction of N that are P

$TNF^0 =$  True positives with  $r \leq r_0$

$$sen^0 = \frac{TNF^0}{T} = \text{Fraction of P that you will detect with } r \leq r_0$$

| <u>r</u> | <u>pral<sup>0</sup></u> | <u>fdr<sup>0</sup></u> | <u>sen<sup>0</sup> (%)</u> | <u>F<sup>*</sup></u> | <u>FP<sup>*</sup></u> | <u>T<sup>*</sup></u> |
|----------|-------------------------|------------------------|----------------------------|----------------------|-----------------------|----------------------|
| 0.0029   | 0.02                    | 0.17                   | 16.9                       | 29                   | 2                     | 27                   |
| 0.0034   | 0.05                    | 0.30                   | 23.1                       | 42                   | 15                    | 37                   |
| 0.0042   | 0.10                    | 0.45                   | 30.6                       | 59                   | 25                    | 49                   |
| 0.0063   | 0.28                    | 0.67                   | <u>50.0</u>                | 108                  | 73                    | 79                   |

detecting 50% of P, implies ~ 28% of 265 are FP  
~ 67% of 108 are FP

"P-values should be treated with caution"

- no info about  $H_1$   
rejecting  $H_0$  does not mean  $H_1$  is true
- p-values are not the probability of any model

$$p\text{-value} = 0.02 \not\Rightarrow \left. \begin{array}{l} P(H_0 \text{ is true}) = 0.02 \\ P(H_1 \text{ is true}) = 0.98 \end{array} \right\}$$

- p-values cannot be used to compare different models
- "I run test, and the p-value is  $10^{-5}$ "  
should you be surprised?

$CDF_X(x)$  is a  $U[0:1]$

p-values are uniformly distributed!!  $[0:1]$

If data follows the null hypothesis  
you should be as surprised of a  
p-value of 0.05 as one of 0.99!

- when someone uses a p-value, request that  
they tell you under which null hypothesis they  
are operating!

## How to treat p-values with caution

- (1) precisely define the null hypothesis and its pdf

$$P(x|H_0)$$

- (2) When possible, estimate the  $P(x|H_0)$  yourself, don't rely on test with obscure or unspecified assumptions. They all have them!

- (3) P-values do not validate, nor refute any hypothesis. They give you a sense of the # of FP you should expect, based on which you may take further actions.

- (4) What to do if you have more than 1 alternative hypothesis?

The bayesian way is clear:

$$P(H_1|D) = \frac{P(D|H_1) \cdot P(H_1)}{P(D|H_1)P(H_1) + P(D|H_2)P(H_2) + P(D|H_3)P(H_3)}$$

but using p-values?