

Lecture 1

Please see the wikipedia entry on "Sensitivity and Specificity"

warning using conventional definitions,

p. 17 top of the course lecture notes is wrong: there are two reasonable ways

to define the false positive rate, and the official definition seems counter-intuitive to me but we need to use

what test says

	truth	
	+	-
+	true + TP	false + FP
-	false - FN	true - TN

official false positive rate:

$$FPR = \frac{FP}{FP + TN}$$

incorrect definition

p. 17 lecture notes

$$"FPR" = \frac{FP}{TP + FP}$$

what is computed on p. 17 of the lecture notes is officially called the false discovery rate

rate
$$FDR = \frac{FP}{FP+TP}$$

official
$$FDR = \frac{594}{594 + 98,406}$$

$= 0.006$
 $= 0.6\%$

truth
really really
 $\theta = 1$ $\theta = 0$

test says $\hat{y}_i = 1$ \oplus	999	594	1,593
$\hat{y}_i = 0$ \ominus	1	98,406	98,407
	1,000	99,000	100,000

official
$$FDR =$$

$$\frac{594}{999 + 594} = 0.373$$

$$= 37.3\%$$

FNR rate definition on

p. 18 of the lecture notes is also

unstandard: official
$$FNR = \frac{FN}{FN+TP}$$

what's called FNR on p. 18 is false omission rate

officially
$$FOR = \frac{FN}{FN+TN} = \frac{1}{1+98,406} = 1.0 \cdot 10^{-5} = 0.1\%$$

information flow: $\xrightarrow{\text{time}}$

A $A|B$ $A|B|C$ ^(*)

$P(A) = ?$ $P(A|B)$ $P(A|B|C) = ?$

B arrives C arrives

we know from Boyer's theorem directly

$$\text{that } P(A|B) = \frac{P(A)P(B|A)}{P(B)} \quad (*)$$

what we showed in the corrected version of the 7 Jan 21 lecture

$$\text{was } P(A|B|C) = \frac{P(A|B)P(C|AB)}{P(C|B)} \quad (**)$$

in going from (*) to (**), B gets moved to the other side of the conditioning bar and C takes its place

$1 = \text{really HIV}$
 $0 = \text{not}$

what says | truth
 1 = 0 | 0
 0 = 0 | 0
 1 = 1 | 1
 0 = \vdots | \vdots

1 row for each subject

$n = 100,000$

Grant

0	0	n_{00}
\vdots	\vdots	
0	1	n_{01}
\vdots	\vdots	
1	0	n_{10}
\vdots	\vdots	
1	1	n_{11}
\vdots	\vdots	

1 column for each variable

truth

	$\theta = 1$	$\theta = 0$
what test says $y_1 = 1$	999	594
what test says $y_1 = 0$	1	98,406
	1000	99,000

prevalence

prev. $P(\theta = 1 | B) = 0.01$

1,593 tests. $P(y_1 = 1 | \theta = 1) = 0.999$

Specificity =

$P(y_1 = 0 | \theta = 0)$

= 0.994

$P(\theta = 1 | y_1 = 1, B) = \frac{999}{1,593} = 0.627$

= 62.7%

$P(y_1 = 1 | B) = \frac{1,593}{100,000}$

real-world consequences of FP

bad doctor says Bob HIV+ when actually HIV-

test says $y_1 = 0$ which is right

truth

	$\theta = 1$	$\theta = 0$
test says $y_1 = 1$	✓	FP mistake
test says $y_1 = 0$	FN mistake	✓ test right

real-world consequences of FN

Bob told HIV ⊖ when really HIV ⊕ really bad ⑥

Bayesian decision theory

$$(\underline{A}, \mathcal{P}, \mathcal{B}) \leftarrow R = (Q, C)$$

$$[\underline{a} | \underline{B}], \underline{u(a, \theta | B)}$$

action space

utility function

several actors in this drama:

$$u(a, \theta | B) \in \mathbb{R}$$

convention
large u better than small

- * Bob
- * FDA
- * you (Bob's doctor)
- * Bob's health care system (\$)
- * * company making Determine test

one big problem:

all actors in this drama ⁽⁷⁾ may have different utility functions

only one actor

then:

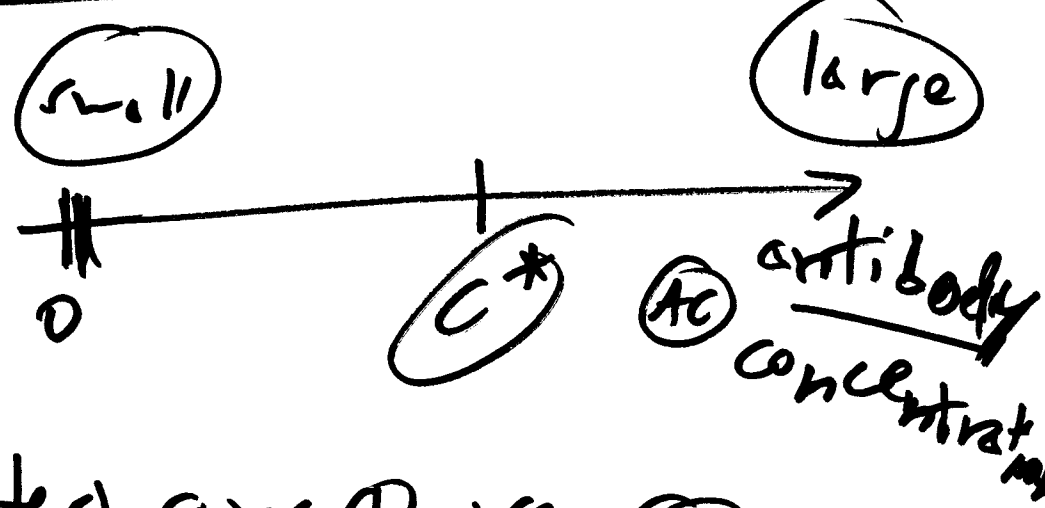
if \equiv optimal action $a^* \in (A|B)$

~~maximizes~~ maximizes $E[U(a, \theta | B)]$
 $(\theta | B)$ unknown

cut do that

possible actions for ~~company~~ company

change c^*



test says (+) iff $AC > c^*$

as $c^* \uparrow$ FP \downarrow FN \uparrow
 $c^* \downarrow$ FP \uparrow FN \downarrow

MEU: maximize expected utility

with multiple actors in draws,

MEU with a single utility function only works when all actors agree on a common specification of $U(q, \theta | \underline{B})$

when they don't agree (everyone is in a situation of partial competition and partial cooperation with everyone else): MEU doesn't work

in situation \square the current best ^⑨
math is Bayesian game theory
