

TEOREMA DE BAYES

$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$ BAYES RULE
INVERSE PROBABILITY

CONDITIONAL PROBABILITY:

WHAT IS THE PROBABILITY OF OBSERVING A GIVEN I'VE OBSERVED B?

EXAMPLE I - UNFAIR COIN

I HAVE TWO COINS ON THE TABLE, ONE FAIR THE OTHER UNFAIR. I PICK ONE A RANDOM, THROW IT AND GET TAILS. WHAT IS THE PROBABILITY THAT I THREW THE FAIR COIN?

EXAMPLE II - URNS

ON A TABLE, THERE ARE TWO URNS: THE FIRST(I) HAS 70 BLACK & 30 RED BALLS WHILE THE SECOND(II) HAS 50 BLACK & 50 RED.

YOU PICK AN URN @ RANDOM & DRAW 3 BALLS: 2 BLACK & 1 RED. WHAT IS THE PROBABILITY YOU DREW FROM URN A?

SUCCESS = BLACK

$P(I|x=2) = \frac{P(x=2|I) \cdot P(I)}{P(B)}$

$P(B) = P(x=2) = P(I) \cdot P(x=2|I) + P(II) \cdot P(x=2|II)$
 $= 0.5 \cdot P(x=2|I) + 0.5 \cdot P(x=2|II)$

BAYESIAN VS. FREQUENTIST

INVERSE PROBABILITY - WHAT DOES IT MEAN? FREQUENCY OR BELIEF?

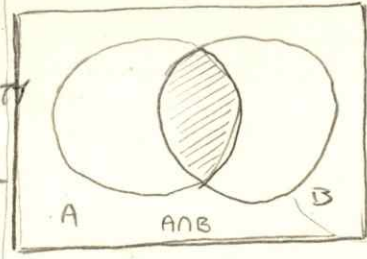
BAYESIAN DATA ANALYSIS

PARAMETERS → PRIORS

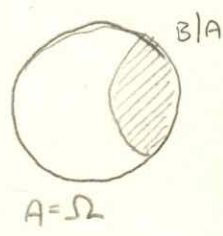
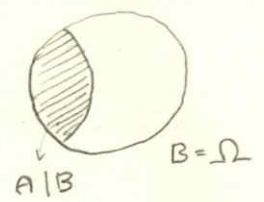
PRIORS + DATA → POSTERIOR

PRIOR: DISTRIBUTION: OUR KNOWLEDGE ABOUT PARAMETERS PRIOR TO ANY OBSERVATION

POSTERIOR DISTRIBUTION: OUR KNOWLEDGE ABOUT PARAMETERS AFTER DATA HAS BEEN OBSERVED.



I. $P(A|B) = \frac{P(A \cap B)}{P(B)}$

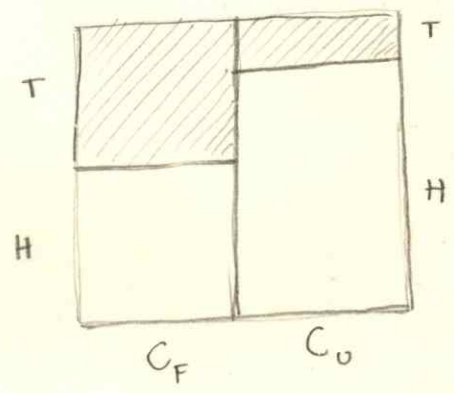


II: $P(B|A) = \frac{P(A \cap B)}{P(A)}$
 ↓ ISOLATE P(A ∩ B)

$P(A \cap B) = P(B|A) P(A)$
 ↓ SUBS. II IN I

$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$

EXAMPLE I



$P(C_F) = P(C_U) = 0.5$

$P(H|C_F) = P(T|C_F) = 0.5$

$P(H|C_U) = 0.7, P(T|C_U) = 0.3$

$P(C_F|H) = \frac{P(H|C_F) \cdot P(C_F)}{P(H)}$

$P(H) = P(C_F) \cdot P(H|C_F) + P(C_U) \cdot P(H|C_U)$
 $= 0.5 \cdot 0.5 + 0.5 \cdot 0.7$
 $= 0.25 + 0.35 = 0.6$

$\frac{0.5 \cdot 0.5}{0.6} = 0.4166$

$P(\theta | \text{data}) = \frac{P(\text{data} | \theta) \cdot P(\theta)}{P(\text{data})}$