Bayesian Statistics and Hierarchical Bayesian Modeling for Psychological Science

Lecture 04

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https://github.com/lei-zhang/BayesCog_Wien

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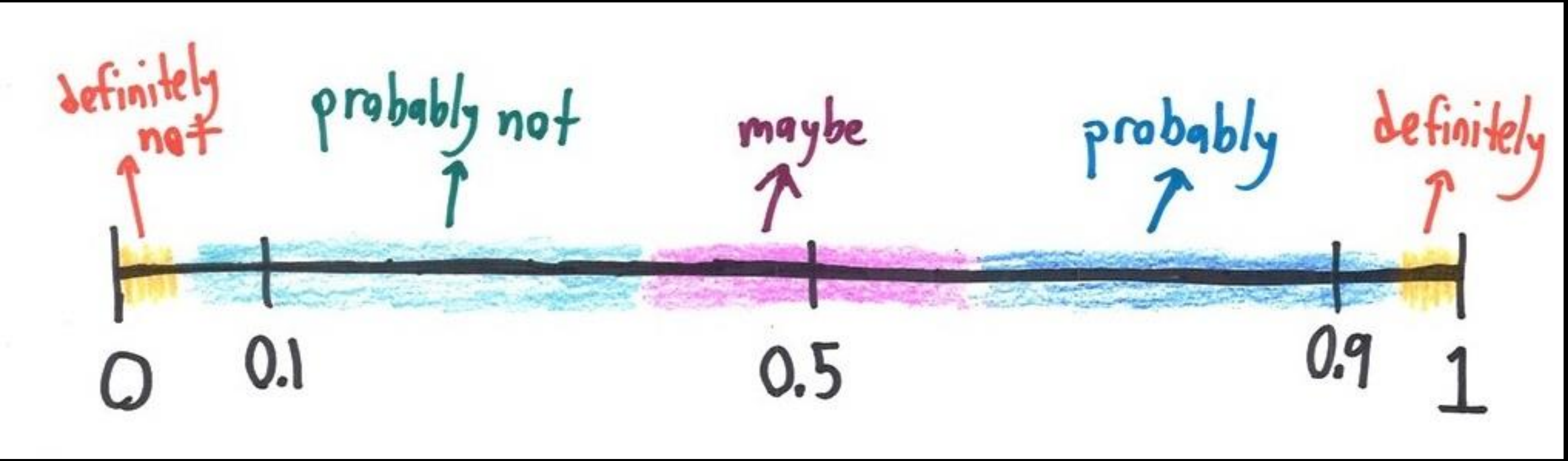
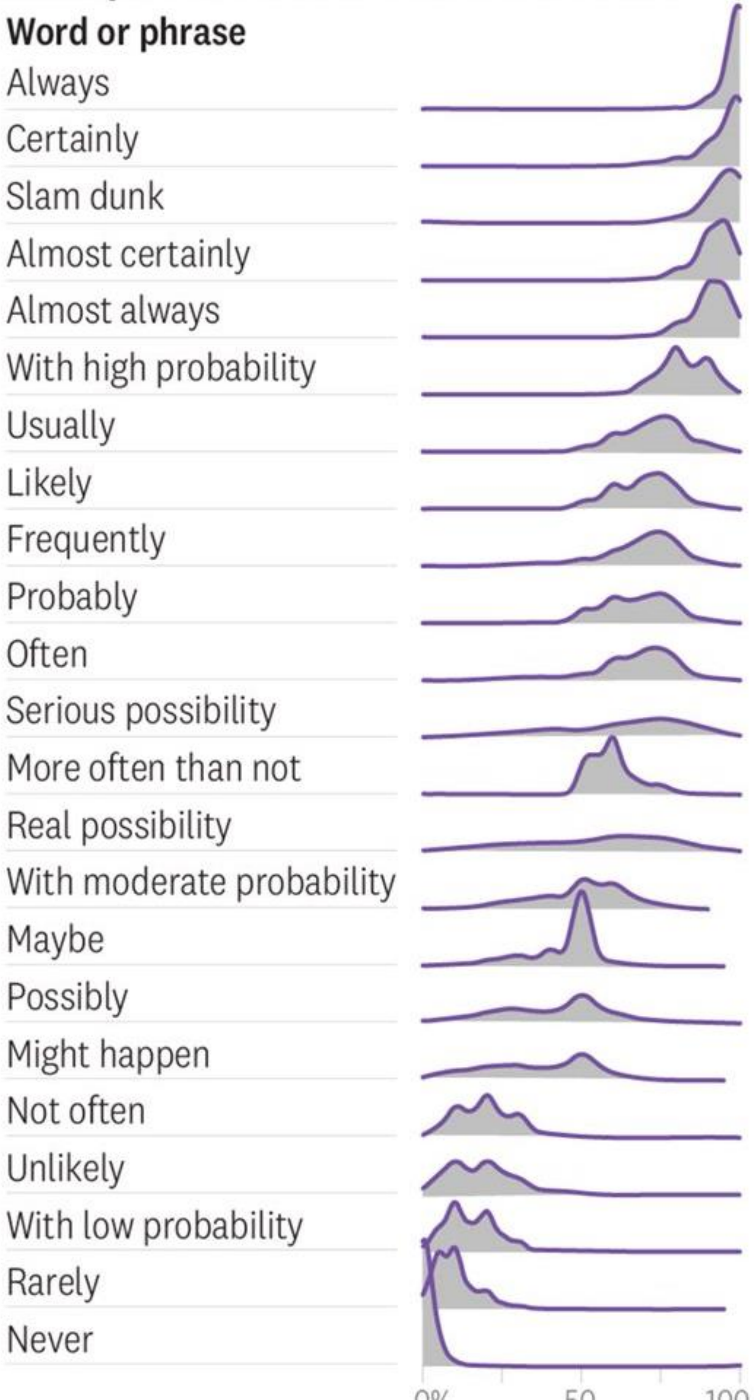
Fakultät für Psychologie

Bayesian warm-up?

BASICS OF PROBABILITY



to respondents' estimate of likelihood



Probability

…assigning numbers to a set of possibilities

Properties (Kolmogorov, 1956)

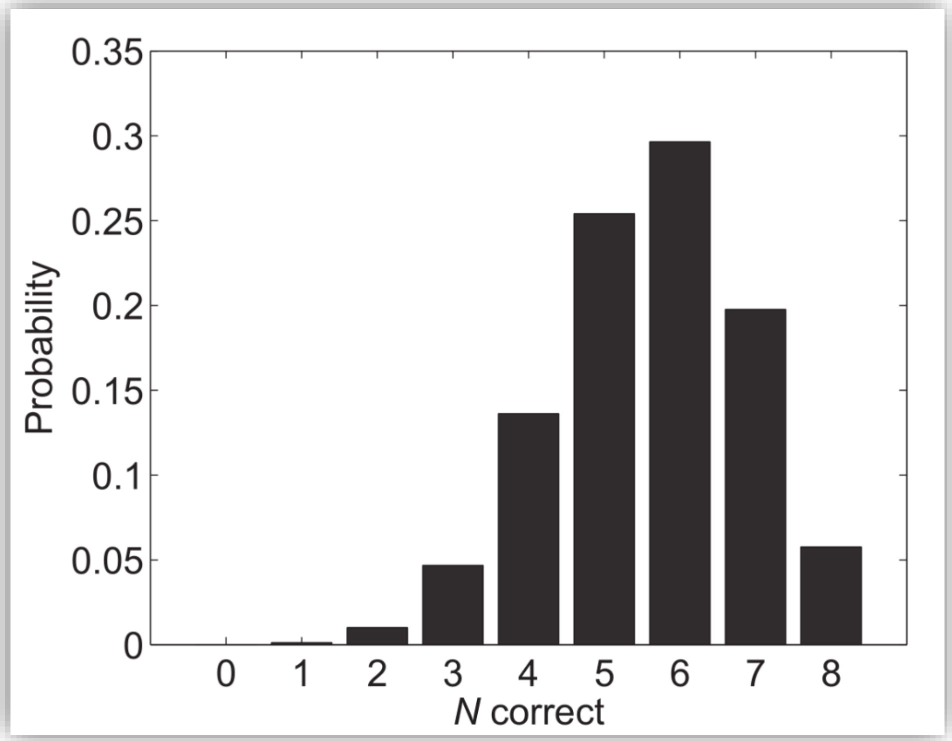
- $p \in [0,1]$
- $\Sigma p = 1$

Probabilities are used to express uncertainty.

Probability Functions

discrete events – we talk about mass

Run a test and record each student's correct responses

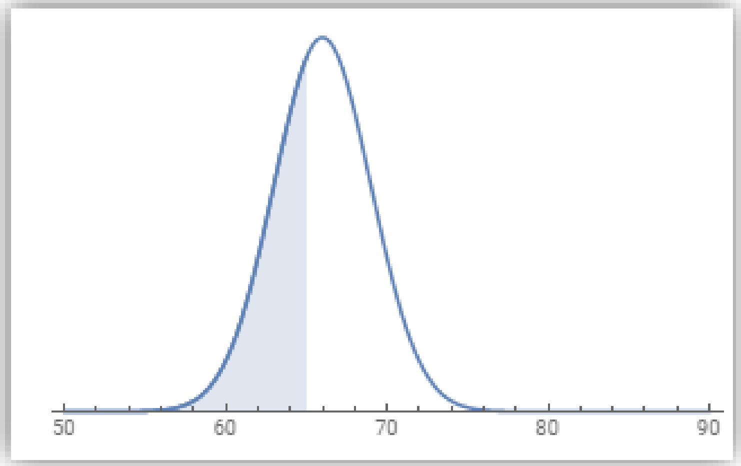


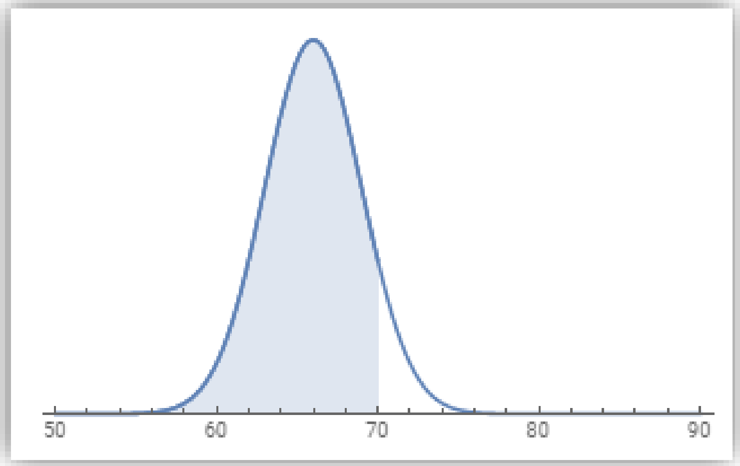
Lewandowsky & Farrell (2011)

Probability Functions

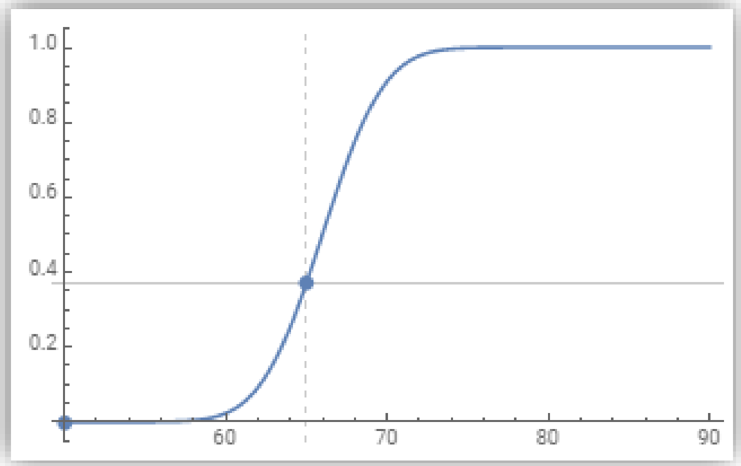
continuous events – we talk about density

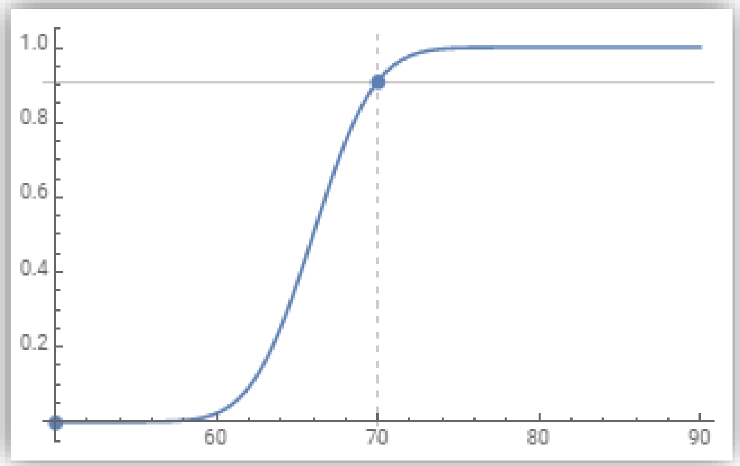
probability density function
(PDF)





cumulative distribution function
(CDF)

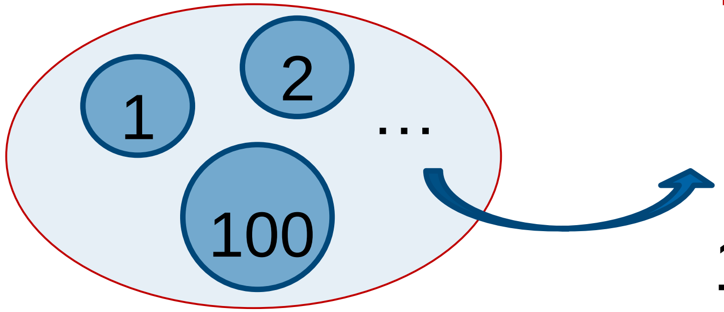




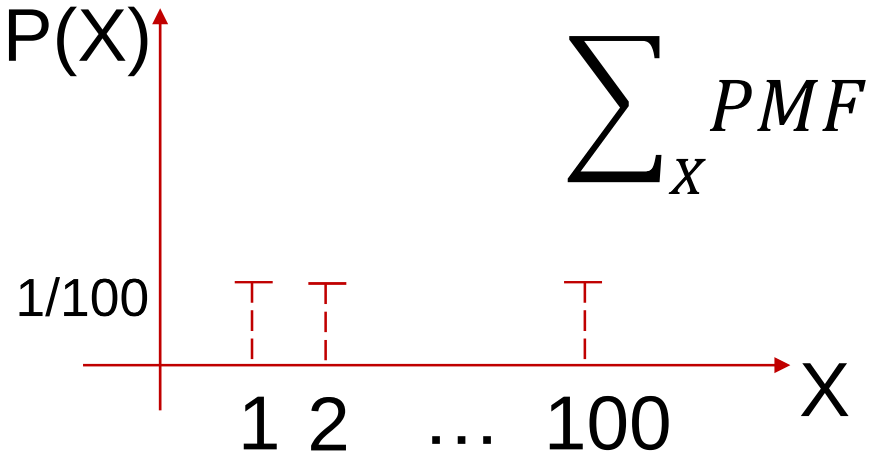
https://demonstrations.wolfram.com/ConnectingTheCDFAndThePDF/

Another example

Discrete



X	P(X)
1	1/100
2	1/100
…	…
100	1/100



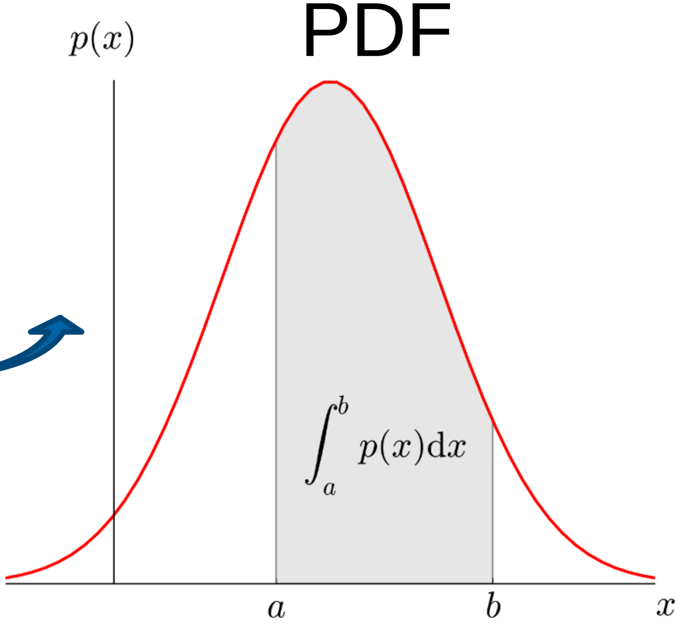
$$\sum_X PMF(X) = 1$$

Continuous

Height

X	P(X)
1.8 m	0

PDF



P given by the area

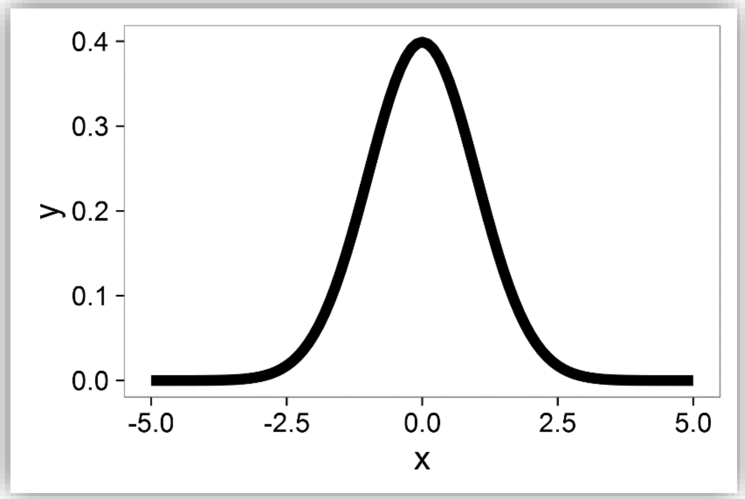
$$1.75 \leq X \leq 1.85$$

Playing with Probability Functions in R

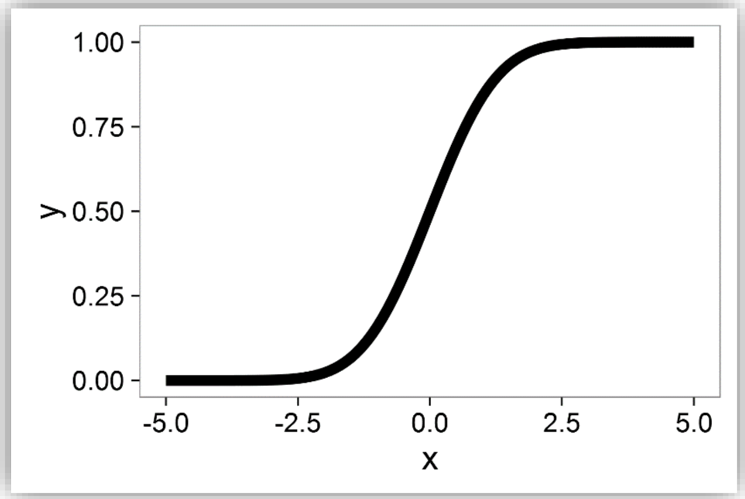

```
dnorm() – PDF
pnorm() – CDF
qnorm() – quantile, inverse cdf
rnorm() – random number generator
```

Example: Normal(0,1)

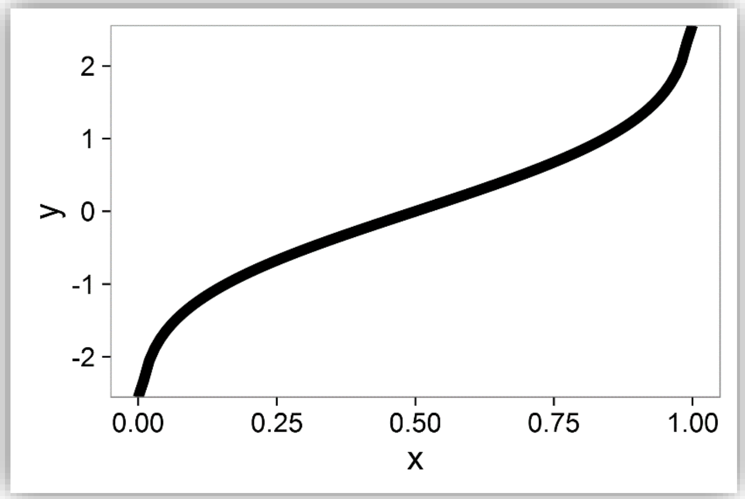
`dnorm()`



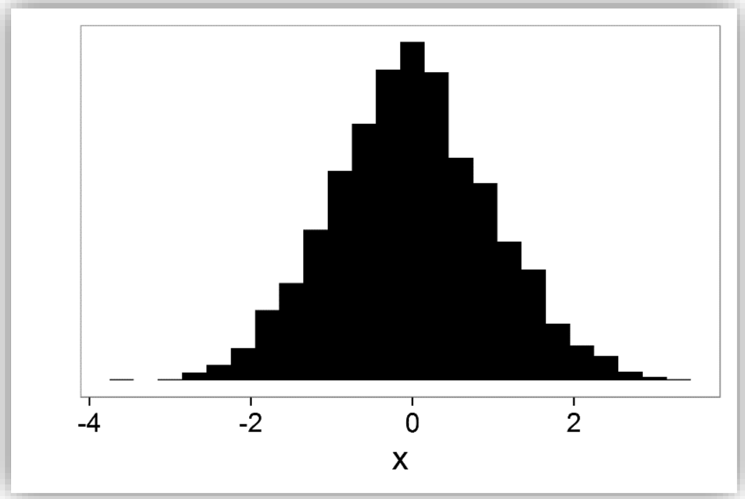
`pnorm()`



`qnorm()`



`rnorm()`



Joint, Marginal and Conditional Probability

Joint Probability

$p(A, B) = p(B, A)$

- e.g., $p(\text{rain, cold})$: $p(\text{rain})$ AND $p(\text{cold})$

Marginal Probability

$p(A)$ – 'p of A irrespective of B'

- e.g., $p(\text{rain})$: $p(\text{rain, cold}) + p(\text{rain, not cold})$

Conditional Probability

$p(A|B)$ – 'p of A given B' – event B is fixed, not uncertainty

$p(A,B) = p(A|B)p(B)$

- e.g., $p(\text{rain, cold}) = p(\text{rain}|\text{cold})p(\text{cold})$

Example I: discrete

rain

X

cold Y

		1	0
Y	1	0.5	0.1
	0	0.1	0.3

Joint probability :

$$P(X=0, Y=1) =$$

$$\sum_{x,y} P(X=x, Y=y) = 1$$

Marginal probability :

$$P(Y=1) =$$

$$P(X=0) =$$

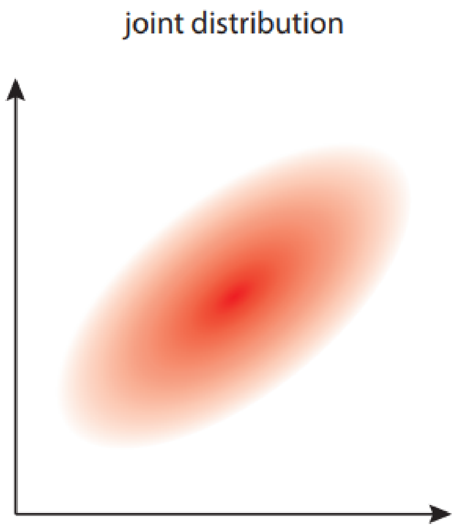
$$P(X=x) = \sum_{y} P(X=x, Y=y)$$

Conditional probability :

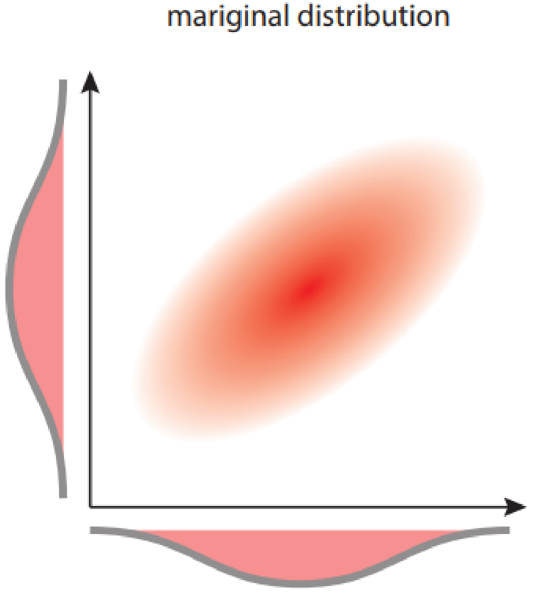
$$P(X=1|Y=1) =$$

$$P(X=x|Y=y) = \frac{P(X=x, Y=y)}{P(Y=y)} = \frac{P(X=x, Y=y)}{\sum_x P(X=x, Y=y)}$$

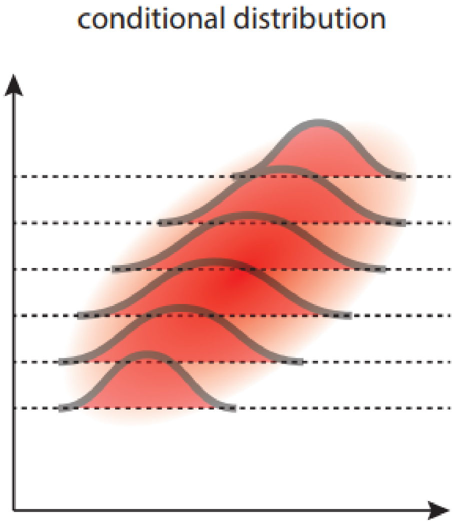
Example I: continuous



The "co-distribution" of x and y.



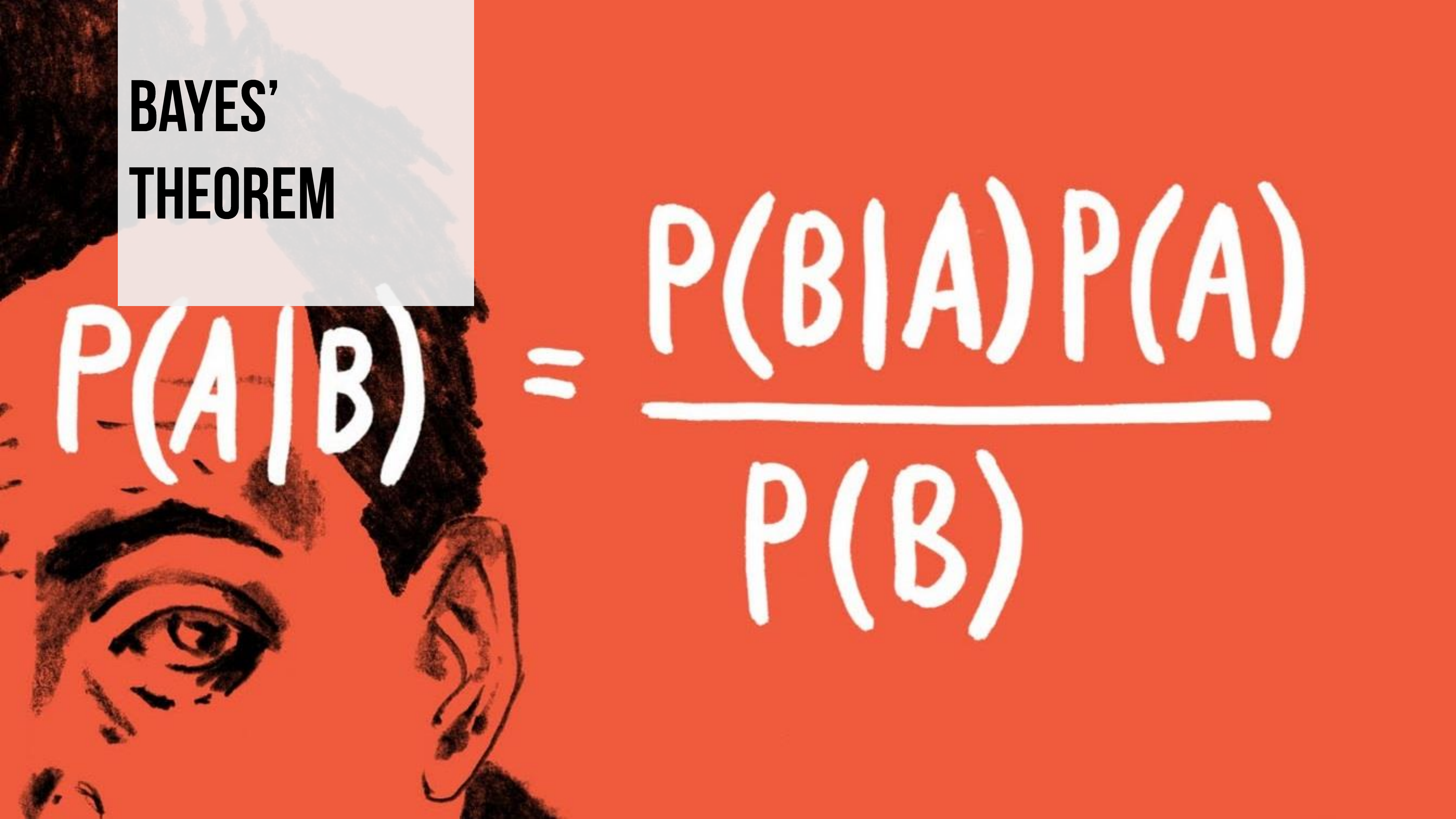
The density of x- (or y-) values, without knowing the other's value.



The probability distribution of x, given that we know the value of y.

Credit: Eric Ma

BAYES' THEOREM



Bayes' theorem

$$p(A,B) = p(B,A)$$

$$p(A,B) = p(A|B)\,p(B)$$

$$p(B,A) = p(B|A)\,p(A)$$

$$p(A|B)\,p(B) = p(B|A)\,p(A)$$

$$\boxed{p\left(A \mid B\right) = \frac{p\left(B \mid A\right) p\left(A\right)}{p\left(B\right)}}$$

Bayesian warm-up?

Second Example

Row	Column			Marginal
	. . .	*c*	. . .	
⋮		⋮		
r	. . .	$p(r, c) = p(r\|c)\, p(c)$	. . .	$p(r) = \sum_{c^*} p(r\|c^*)\, p(c^*)$
⋮		⋮		
Marginal		$p(c)$		

Kruschke (2015)

Second Example

	Hair color				
Eye color	Black	Brunette	Red	Blond	Marginal (Eye color)
Brown	0.11	0.20	0.04	0.01	0.37
Blue	0.03	0.14	0.03	0.16	0.36
Hazel	0.03	0.09	0.02	0.02	0.16
Green	0.01	0.05	0.02	0.03	0.11
Marginal (hair color)	0.18	0.48	0.12	0.21	1.0

$$p(A \mid B) = \frac{p(B \mid A)\, p(A)}{\sum_{A^*} p(B \mid A^*)\, p(A^*)}$$

Kruschke (2015)

Exercise VI

Suppose that in the general population, the probability of having a rare disease is 1/1000. We denote the true presence or absence of the disease as the value of a parameter, ϑ, that can have the value ϑ = ☹ if disease is present in a person, or the value ϑ = ☺ if the disease is absent. The base rate of the disease is therefore denoted $p(\vartheta =$ ☹ $) = 0.001$.

Suppose(1): a test for the disease that has a 99% hit rate: $p(T = + | \vartheta =$ ☹ $) = 0.99$

Suppose(2): the test has a false alarm rate of 5%: $p(T = + | \vartheta =$ ☺ $) = 0.05$

Q: Suppose we sample a person at random from the population, administer the test, and it comes up positive. What is the posterior probability that the person has the disease?

Kruschke (2015)

Exercise VI

Q: What is the posterior probability that the person has the disease?

→ $p(\vartheta = ☹ \mid T = +)$

Exercise VI

Test result	Disease: $\theta = ☹$ (present)	Disease: $\theta = ☺$ (absent)	Marginal (test result)
$T = +$	$p(+\|☹)\, p(☹)$ $= 0.99 \cdot 0.001$	$p(+\|☺)\, p(☺)$ $= 0.05 \cdot (1 - 0.001)$	$\sum_\theta p(+\|\theta)\, p(\theta)$
$T = -$	$p(-\|☹)\, p(☹)$ $= (1 - 0.99) \cdot 0.001$	$p(-\|☺)\, p(☺)$ $= (1 - 0.05) \cdot (1 - 0.001)$	$\sum_\theta p(-\|\theta)\, p(\theta)$
Marginal (disease)	$p(☹) = 0.001$	$p(☺) = 1 - 0.001$	1.0

$$p(\theta = ☹ \,|\, T = +) = \frac{p(T=+|\theta=☹)\, p(\theta=☹)}{\sum_\theta p(T=+|\theta)\, p(\theta)}$$

$$= \frac{0.99 \cdot 0.001}{0.99 \cdot 0.001 + 0.05 \cdot (1 - 0.001)}$$

$$= 0.019$$

Kruschke (2015)


ANY
QUESTIONS
?

Happy Computing!