

STT 465

Lecture 2:

Belief, probability, independence, conditional  
independence, exchangeability

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# Beliefs & Probability

⇒ Beliefs and probability (read the chapter in the book)

⇒ Review of probability

(1) Events  $\{A_i\}$  & Sample space  $S=\{A_1, \dots\}$

(2) Probability (map from events to numbers in the  $[0,1]$  interval that follows a few rules).

(3)  $P(A_i) \geq 0$

(4)  $P(S)=1$

(5) Probability of the union:  $P(A_i \cup A_j) = P(A_i) + P(A_j) - P(A_i \cap A_j)$  [ $\cup$ =OR,  $\cap$ =&]

⇒ Marginal, conditional and joint probabilities

(6) Marginal probability:  $P(A)$

(7) Joint probability:  $P(A \& B) = P(A, B)$  [we will commonly use ',' for joint]

(8) Conditional probability:  $p(A | B)$

(9) Factorization:  $P(A, B) = P(A) P(B | A) = P(B) P(A | B)$

(10) Rule of marginal probability:  $P(A) = P(A) P(A | B) + P(\text{Not } B) P(B | \text{Not } A)$

⇒ Bayes rule:  $p(A | B) = P(B | A) P(A) / P(B)$

# Examples

⇒ Review income/education example (p 16 )

⇒ Example: genetics of disease

Genotype	Status
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AA	H
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AB   BA	H
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BB	D
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$p(A)=0.8$

⇒ Determine

- (1) The (marginal) probability of each genotype under random mating,  $p(G)$
- (2) The probability of disease of a randomly sampled individual,  $P(D)$
- (3) The conditional probability of disease given genotypes,  $p(D|G)$
- (4) The joint probability of disease and genotypes,  $p(D,G)$

# Bayesian Learning

- ⇒ Compute the probability that a parent is a carrier given that the first offspring is healthy  $p(P=AB \mid O_1=H)$
- ⇒ How would that probability be updated if you know that the second offspring is also healthy?
- ⇒ What about if you know that the third offspring developed the disease?

# Conditional Independence

⇒ X,Y are independent if  $p(X,Y)=p(X)p(Y)$

⇒ X,Y are said conditionally independent if, given a third variable (Z), X and Y are independent, that is:

Conditional Independence:  $p(X,Y|Z)=P(X|Z)*P(Y|Z)$

⇒ Note: conditional independence does not imply marginal independence

⇒ HW1-Q1: Consider the disease model of slide 3, with  $p(A)=0.8$ , for this model:

(1.1) Derive the conditional probability of disease given the genotype of one of the parents, that is  $p(S_i|G_0)$ .

(1.2) Derive and report the joint probability of the disease status of two half sibs (i.e., two individuals that share one parent) given the genotype of the known parent, that is  $p(S_1, S_2|G_0)$ .

(1.3) Derive and report the joint probability of the disease status of two half sibs ,that is  $p(S_1, S_2)$ .

(1.4) Are  $S_1$  and  $S_2$  independent?

(1.5) Derive and report  $p(S_2|S_1)$ .

Notation:  $G_0$  denotes the genotype of the known parent and can take values {AA,AB,BB}.  
 $S_i$  denotes the health status of the ith progeny ( $i=1,2$ ), it can take values {H,D}.

# Remarks

- ⇒ The joint probability contains all the information needed to arrive at the marginal s and conditionals
- ⇒ We can arrive at the joint distribution from the marginal only with knowledge of the conditional distributions (under independence this is trivial).

# Exchangeability & Conditional Independence

⇒ Bruno de Finetti established an important relationship between exchangeable sequences of RVs and Conditional independence (de Finetti's theorem)

⇒ Exchangeable Sequence of Random Variables

- Sequence of Random Variables  $Y_1, Y_2, \dots, Y_n$
- Joint Distribution  $p(Y_1, Y_2, \dots, Y_n)$
- Permutation:  $\pi = \{\pi_1, \pi_2, \dots, \pi_n\}$

⇒ A sequence of RV is said to be exchangeable if

$p(Y_{\pi_1}, Y_{\pi_2}, \dots, Y_{\pi_n})$  is the same for any permutation  $\pi$

⇒ Note: sequence of IID RVs are exchangeable (discuss), but the converse is not TRUE

⇒ Importantly, independence is not required for exchangeability (discuss MVN case)

# de Finetti's Theorem

⇒ Bruno de Finetti established an important relationship between exchangeable sequences of RVs and Conditional independence (de Finetti's theorem)

⇒ If  $Y_1, Y_2, \dots, Y_n$  is an exchangeable sequence of random variables then, for some parameter  $(\theta)$ , some prior density,  $p(\theta)$  and some conditional density  $p(Y_i | \theta)$ , the joint distribution of the sequence can be expressed as:

$$p(Y_1, Y_2, \dots, Y_n) = \int \left\{ \prod_{i=1}^n p(Y_i | \theta) \right\} p(\theta) d\theta$$

⇒ Note:

$$[\text{Allways}]: p(Y_1, Y_2, \dots, Y_n) = \int p(Y_1, Y_2, \dots, Y_n, \theta) d\theta = \int p(Y_1, Y_2, \dots, Y_n | \theta) p(\theta) d\theta$$

⇒ De Finetti's theorem tells us that if the sequence is exchangeable, we can assume conditional IID

$$p(Y_1, Y_2, \dots, Y_n | \theta) = \prod_{i=1}^n p(Y_i | \theta)$$