

**EE 464**

**Spring 2003**

**Lecture Notes Part 8d**

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## 8.6.2 Total Probability and Bayes' Theorem

Let  $A_1, \dots, A_n$  be a partition of  $\Omega$ . Then

$$P(B) = P(B|A_1)P(A_1) + \dots + P(B|A_n)P(A_n).$$

Let  $B = \{X \leq x\}$ . Then

$$P(X \leq x) = P(X \leq x|A_1)P(A_1) + \dots + P(X \leq x|A_n)P(A_n).$$

$$\Rightarrow F(x) = F(x|A_1)P(A_1) + \dots + F(x|A_n)P(A_n)$$

$$\Rightarrow f(x) = f(x|A_1)P(A_1) + \dots + f(x|A_n)P(A_n).$$

Also,

$$P(A|X \leq x) = \frac{P(A, X \leq x)}{P(X \leq x)} = \frac{P(X \leq x|A)P(A)}{P(X \leq x)} = \frac{F(x|A)}{F(x)}P(A).$$

Similarly,

$$\begin{aligned} P(A|x_1 < X \leq x_2) &= \frac{P(A, x_1 < X \leq x_2)}{P(x_1 < X \leq x_2)} = \frac{P(x_1 < X \leq x_2|A)P(A)}{P(x_1 < X \leq x_2)} \\ &= \frac{F(x_2|A) - F(x_1|A)}{F(x_2) - F(x_1)}P(A). \end{aligned}$$

In the above we have conditioned on events like  $\{X \leq x\}$  or  $\{x_1 < X \leq x_2\}$ . We cannot use the above development directly when conditioning on the event  $\{X = x\}$  (since this event has zero probability in the continuous case). We define

$$P(A|X = x) = \lim_{\Delta x \rightarrow 0} P(A|x < X \leq x + \Delta x).$$

Recall,

$$P(A|x_1 < X \leq x_2) = \frac{F(x_2|A) - F(x_1|A)}{F(x_2) - F(x_1)}P(A).$$

Let  $x_1 = x$ ,  $x_2 = x + \Delta x$ . Then

$$\begin{aligned} P(A|x < X \leq x + \Delta x) &= \frac{F(x + \Delta x|A) - F(x|A)}{F(x + \Delta x) - F(x)}P(A) \\ &= \frac{\frac{F(x + \Delta x|A) - F(x|A)}{\Delta x}}{\frac{F(x + \Delta x) - F(x)}{\Delta x}}P(A). \end{aligned}$$

Now let  $\Delta x \rightarrow 0$  to get

$$P(A|X = x) = \frac{f(x|A)}{f(x)}P(A)$$
$$\Rightarrow \int_{-\infty}^{\infty} P(A|X = x)f(x)dx = P(A) \int_{-\infty}^{\infty} f(x|A)dx = P(A)$$

or

$$P(A) = \int_{-\infty}^{\infty} P(A|X = x)f(x)dx.$$

This is the continuous version of the total probability theorem. Compare to

$$P(A) = \sum_{i=1}^n P(A|B_i)P(B_i)$$

where  $B_1, \dots, B_n$  form a partition of  $\Omega$ .

Recall,

$$P(A|X = x) = \frac{f(x|A)}{f(x)}P(A).$$

Then

$$f(x|A) = \frac{P(A|X = x)f(x)}{P(A)}.$$

Using the above expression for  $P(A)$  we get

$$f(x|A) = \frac{P(A|X = x)f(x)}{\int_{-\infty}^{\infty} P(A|X = x)f(x)dx}.$$

This is the continuous version of Bayes' theorem. Compare to

$$P(B_i|A) = \frac{P(A|B_i)P(B_i)}{\sum_{i=1}^n P(A|B_i)P(B_i)}$$

where  $B_1, \dots, B_n$  form a partition of  $\Omega$ .