

LECTURE 9

- **Readings:** Sections 3.4-3.5

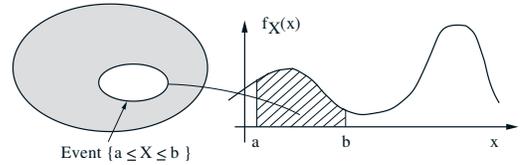
Outline

- PDF review
- Multiple random variables
 - conditioning
 - independence
- Examples

Summary of concepts

$p_X(x)$	$F_X(x)$	$f_X(x)$
$\sum_x x p_X(x)$	$\mathbf{E}[X]$	$\int x f_X(x) dx$
	$\text{var}(X)$	
$p_{X,Y}(x, y)$		$f_{X,Y}(x, y)$
$p_{X A}(x)$		$f_{X A}(x)$
$p_{X Y}(x y)$		$f_{X Y}(x y)$

Continuous r.v.'s and pdf's



$$\mathbf{P}(a \leq X \leq b) = \int_a^b f_X(x) dx$$

- $\mathbf{P}(x \leq X \leq x + \delta) \approx f_X(x) \cdot \delta$
- $\mathbf{E}[g(X)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$

Joint PDF $f_{X,Y}(x, y)$

$$\mathbf{P}((X, Y) \in S) = \int \int_S f_{X,Y}(x, y) dx dy$$

- Interpretation:

$$\mathbf{P}(x \leq X \leq x + \delta, y \leq Y \leq y + \delta) \approx f_{X,Y}(x, y) \cdot \delta^2$$

- Expectations:

$$\mathbf{E}[g(X, Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) f_{X,Y}(x, y) dx dy$$

- From the joint to the marginal:

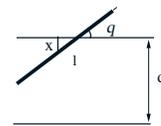
$$f_X(x) \cdot \delta \approx \mathbf{P}(x \leq X \leq x + \delta) =$$

- X and Y are called **independent** if

$$f_{X,Y}(x, y) = f_X(x) f_Y(y), \quad \text{for all } x, y$$

Buffon's needle

- Parallel lines at distance d
- Needle of length ℓ (assume $\ell < d$)
- Find \mathbf{P} (needle intersects one of the lines)



- $X \in [0, d/2]$: distance of needle midpoint to nearest line
 - **Model:** X, Θ uniform, independent
- $$f_{X,\Theta}(x, \theta) = \quad 0 \leq x \leq d/2, 0 \leq \theta \leq \pi/2$$

- Intersect if $X \leq \frac{\ell}{2} \sin \Theta$

$$\begin{aligned} \mathbf{P}\left(X \leq \frac{\ell}{2} \sin \Theta\right) &= \int \int_{x \leq \frac{\ell}{2} \sin \theta} f_X(x) f_{\Theta}(\theta) dx d\theta \\ &= \frac{4}{\pi d} \int_0^{\pi/2} \int_0^{(\ell/2) \sin \theta} dx d\theta \\ &= \frac{4}{\pi d} \int_0^{\pi/2} \frac{\ell}{2} \sin \theta d\theta = \frac{2\ell}{\pi d} \end{aligned}$$

Conditioning

- Recall

$$P(x \leq X \leq x + \delta) \approx f_X(x) \cdot \delta$$

- By analogy, would like:

$$P(x \leq X \leq x + \delta | Y \approx y) \approx f_{X|Y}(x | y) \cdot \delta$$

- This leads us to the **definition**:

$$f_{X|Y}(x | y) = \frac{f_{X,Y}(x, y)}{f_Y(y)} \quad \text{if } f_Y(y) > 0$$

- For given y , conditional PDF is a (normalized) "section" of the joint PDF

- If independent, $f_{X,Y} = f_X f_Y$, we obtain

$$f_{X|Y}(x | y) = f_X(x)$$

Joint, Marginal and Conditional Densities

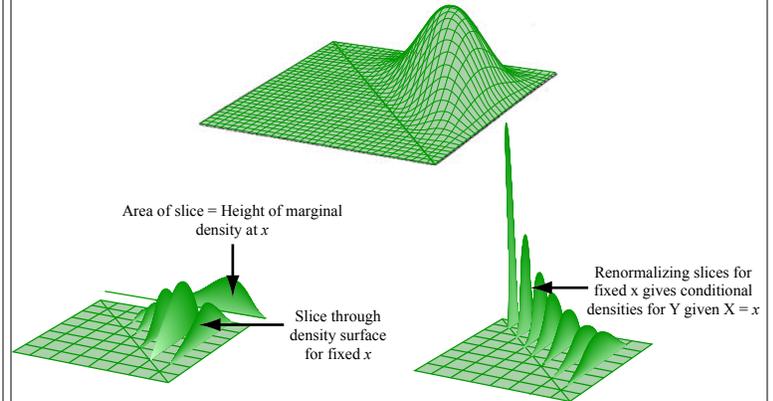
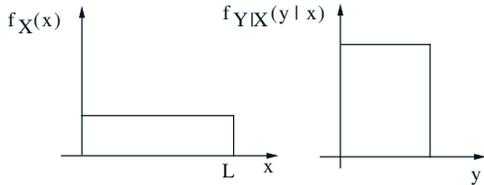


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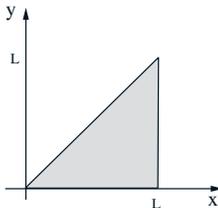
Stick-breaking example

- Break a stick of length ℓ twice:
break at X : uniform in $[0, 1]$;
break again at Y , uniform in $[0, X]$



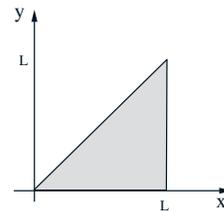
$$f_{X,Y}(x, y) = f_X(x) f_{Y|X}(y | x) =$$

on the set:



$$E[Y | X = x] = \int y f_{Y|X}(y | X = x) dy =$$

$$f_{X,Y}(x, y) = \frac{1}{\ell x}, \quad 0 \leq y \leq x \leq \ell$$



$$\begin{aligned} f_Y(y) &= \int f_{X,Y}(x, y) dx \\ &= \int_y^\ell \frac{1}{\ell x} dx \\ &= \frac{1}{\ell} \log \frac{\ell}{y}, \quad 0 \leq y \leq \ell \end{aligned}$$

$$E[Y] = \int_0^\ell y f_Y(y) dy = \int_0^\ell y \frac{1}{\ell} \log \frac{\ell}{y} dy = \frac{\ell}{4}$$

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