Conditional PDF, given an event

$$p_X(x) = P(X = x)$$
 $f_X(x) \cdot \delta \approx P(x \le X \le x + \delta)$

$$p_{X|A}(x) = \mathbf{P}(X = x \mid A)$$
 $f_{X|A}(x) \cdot \delta \approx \mathbf{P}(x \le X \le x + \delta \mid A)$

Conditional PDF, given an event

 $P(X \in B) = \sum p_X(x)$

 $\mathbf{P}(X \in B \mid A) = \sum p_{X|A}(x)$

$$f_{-1}(x) = \mathbf{D}(X - x)$$

$$p_X(x) = P(X = x) f_X(x)$$

$$f_{Y}(x) - \mathbf{P}(X - x)$$
 $f_{Y}(x) = \mathbf{P}(X - x)$

$$f_{2}$$

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

$$p_{X|A}(x) = \mathbf{P}(X = x \mid A)$$

$$f_{X|A}(x) \cdot \delta \approx \mathbf{P}(x \le X \le x + \delta \mid A)$$

$$(X \in$$

$$\mathbf{P}(X \in B) = \int_B f_X(x) \, dx$$

$$P(X \in B) = \int_{B} f_{X}(x) dx$$
$$P(X \in B \mid A) = \int_{B} f_{X|A}(x) dx$$

Conditional PDF, given an event

 $p_X(x) = P(X = x)$

 $p_{X|A}(x) = P(X = x \mid A)$

 $P(X \in B) = \sum p_X(x)$

 $\sum_{x} p_{X|A}(x) = 1$

 $P(X \in B \mid A) = \sum_{x \in A} p_{X|A}(x)$

 $f_{X|A}(x) \cdot \delta \approx \mathbf{P}(x \le X \le x + \delta \mid A)$

 $P(X \in B) = \int_{B} f_X(x) \, dx$

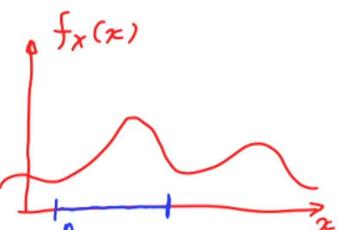
 $\mathbf{P}(X \in B \mid A) = \int_{B} f_{X|A}(x) \, dx$

 $\int f_{X|A}(x) \, dx = 1$

$$X \leq x + 1$$

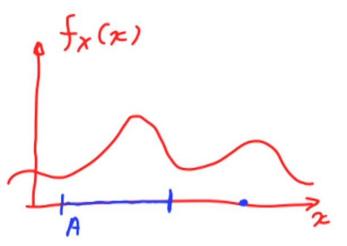
$$f_X(x) \cdot \delta \approx \mathbf{P}(x < X < x + \delta)$$

Conditional PDF of X, given that $X \in A$



Conditional PDF of X, given that $X \in A$

$$P(x \le X \le x + \delta \mid X \in A) \approx f_{X|X \in A}(x) \cdot \delta$$

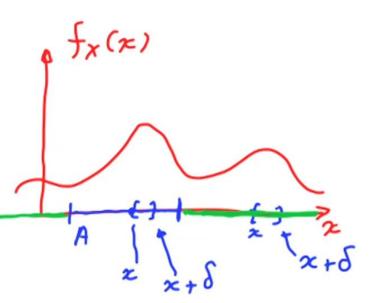


Conditional PDF of X, given that $X \in A$

$$P(x \le X \le x + \delta \mid X \in A) \approx f_{X \mid X \in A}(x) \cdot \delta'$$

$$=\underbrace{\frac{1}{x}(x\leq x\leq z+\delta, X\in A)}_{2}$$

$$= \frac{P(A)}{P(A)} \approx \frac{\int_{x} (x) \delta}{P(A)}$$

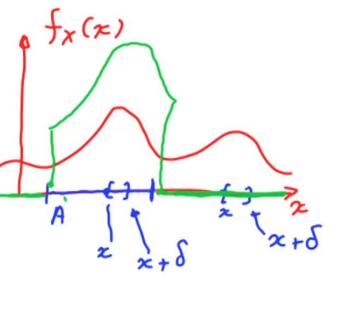


Conditional PDF of
$$X$$
, given that $X \in A$

$$P(x \le X \le x + \delta \mid X \in A) \approx f_{X|X \in A}(x) \cdot \delta'$$

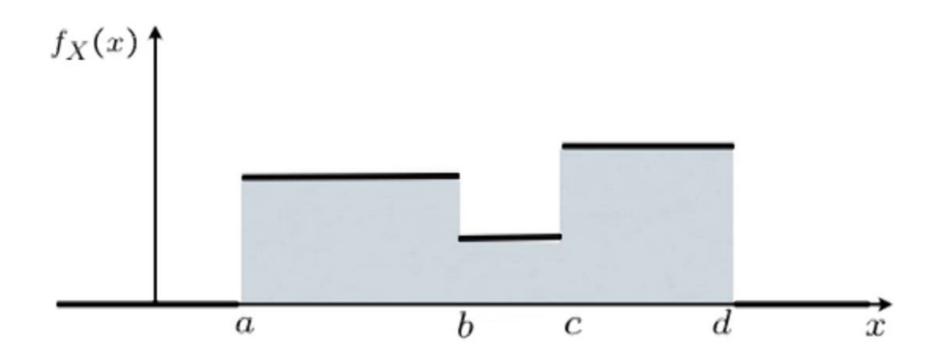
$$= \underbrace{\frac{1}{2} (x \le x \le x + \delta, X \in A)}_{P(A)}$$

$$= \frac{P(A)}{P(A)} \approx \frac{\int_{x} (x) \delta}{P(A)}$$



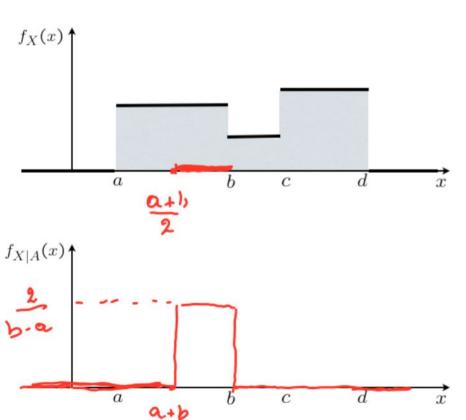
$$f_{X|X\in A}(x) = \begin{cases} 0, & \text{if } x \notin A \\ \frac{f_X(x)}{\mathbf{P}(A)}, & \text{if } x \in A \end{cases}$$

Example



Example

$$A: \quad \frac{a+b}{2} \le X \le b$$



$$E[X \mid A] = \frac{1}{2} \cdot \frac{a+b}{2} + \frac{1}{2}b$$

$$= \frac{1}{4}a + \frac{3}{4}b$$

$$E[X^2 \mid A] = \frac{2}{b-a} \cdot x^2 dx$$

- Do you prefer a used or a new "exponential" light bulb?
- Bulb lifetime T: exponential(λ)

$$P(T > x) = e^{-\lambda x}, \text{ for } x \ge 0$$

- we are told that T > t
- r.v. X: remaining lifetime

- Do you prefer a used or a new "exponential" light bulb?
- Bulb lifetime T: exponential(λ)

$$P(T > x) = e^{-\lambda x}$$
, for $x \ge 0$

- we are told that T > t
- r.v. X: remaining lifetime = T €

$$P(X > x \mid T > t) =$$

- Do you prefer a used or a new "exponential" light bulb? Propabilistically identical!
- Bulb lifetime T: exponential(λ)

$$(\lambda)$$
 $t \approx t + \infty$

$$P(T>x)=e^{-\lambda x}$$
, for $x\geq 0$

- we are told that T > t
- r.v. X: remaining lifetime = T t

$$P(X > x \mid T > t) = e^{-\lambda x}$$
, for $x \ge 0$

$$= \underbrace{\frac{P\left(T-t>x,T>t\right)}{P\left(T>t\right)}} = \underbrace{\frac{P\left(T>t+x,T>t\right)}{P\left(T>t\right)}} = \underbrace{\frac{P\left(T>t+x\right)}{P\left(T>t\right)}}$$

$$= \frac{e^{-\lambda(t+z)}}{e^{-\lambda t}} = e^{-\lambda x}$$

$$f_T(x) = \lambda e^{-\lambda x}$$
, for $x \ge 0$

$$P(0 \le T \le \delta)$$

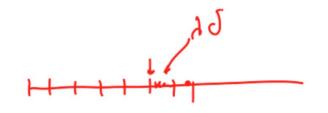
$$P(t \le T \le t + \delta \mid T > t)$$

similar to an independent coin flip, every δ time steps, with $\mathbf{P}(\text{success}) \approx \lambda \delta$

$$f_T(x) = \lambda e^{-\lambda x}$$
, for $x \ge 0$

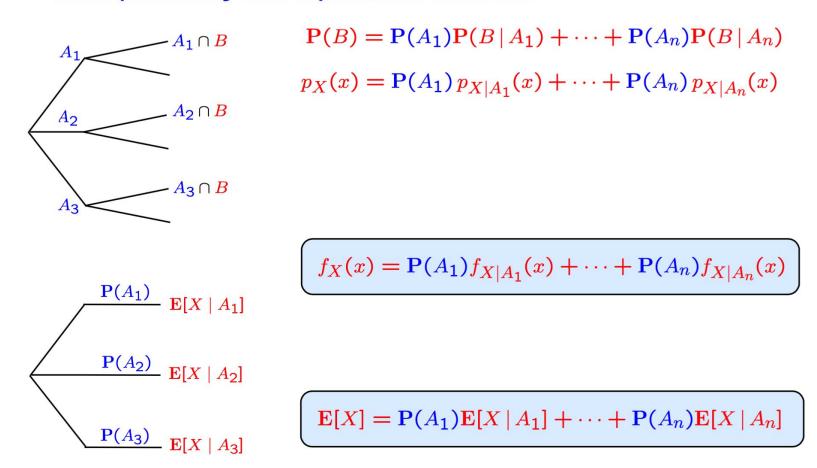
$$P(0 \le T \le \delta) \approx \int_{T} (0) \cdot \delta = \lambda \delta$$

$$P(t \le T \le t + \delta \mid T > t) = 2 \approx \lambda \delta$$



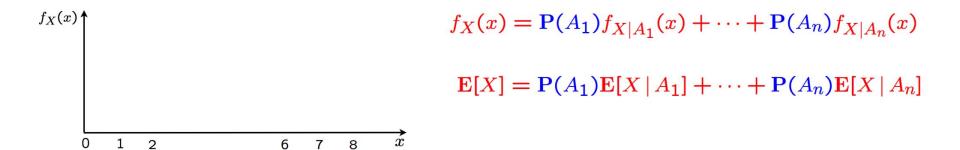
similar to an independent coin flip, every δ time steps, with $\mathbf{P}(\text{success}) \approx \lambda \delta$

Total probability and expectation theorems



Example

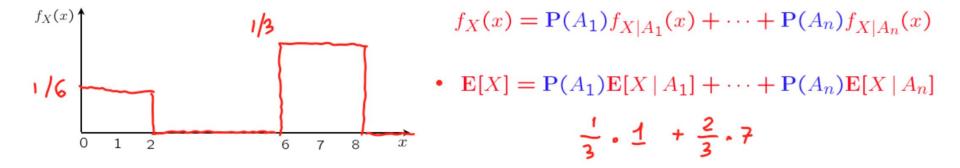
 Bill goes to the supermarket shortly, with probability 1/3, at a time uniformly distributed between 0 and 2 hours from now; or with probability 2/3, later in the day at a time uniformly distributed between 6 and 8 hours from now



Example

 Bill goes to the supermarket shortly, with probability 1/3, at a time uniformly distributed between 0 and 2 hours from now; or with probability 2/3, later in the day at a time uniformly distributed between 6 and 8 hours from now

$$f(A_1) = \frac{1}{3}$$
 $f_{X|A_1} \sim unif[0,2]$ $f(A_2) = \frac{2}{3}$ $f_{X|A_2} \sim U[6,8]$



Jointly continuous r.v.'s and joint PDFs

$$p_X(x)$$
 $f_X(x)$ $p_{X,Y}(x,y)$ $f_{X,Y}(x,y)$

$$p_{X,Y}(x,y) = \mathbf{P}(X = x \text{ and } Y = y) \ge 0$$

$$f_{X,Y}(x,y) \geq 0$$

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

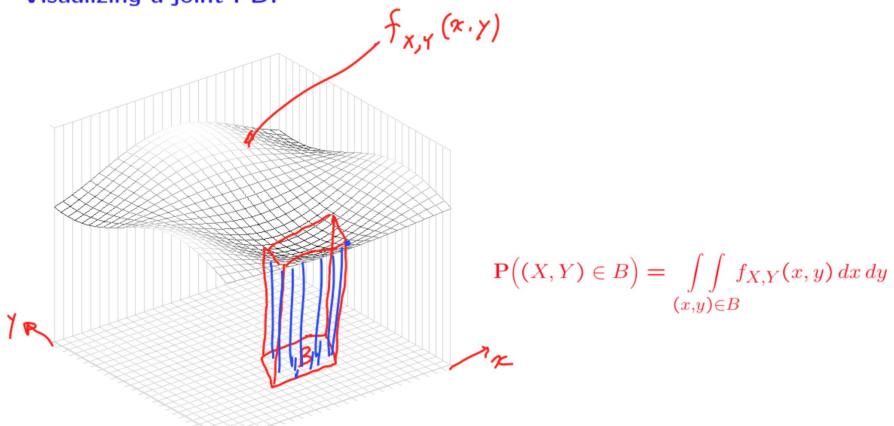
$$\mathbf{P}((X,Y) \in B) = \int \int f_{X,Y}(x,y) \, dx \, dy$$

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

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{X,Y}(x,y) \, dx \, dy = 1$$

Definition: Two random variables are **jointly continuous** if they can be described by a joint PDF

Visualizing a joint PDF



On joint PDFs

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

$$P(a \le X \le b, c \le Y \le d) = \int_c^d \int_a^b f_{X,Y}(x,y) \, dx \, dy$$

$$P(a \le X \le a + \delta, c \le Y \le c + \delta) \approx f_{X,Y}(a,c) \cdot \delta^2$$

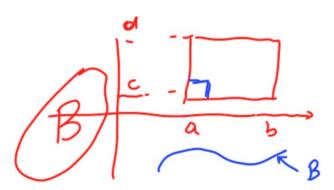
 $f_{X,Y}(x,y)$: probability per unit area

$$area(B) = 0 \Rightarrow P((X,Y) \in B) = 0$$

On joint PDFs

$$\mathbf{P}((X,Y) \in B) = \int_{(x,y) \in B} f_{X,Y}(x,y) \, dx \, dy$$

$$P(a \le X \le b, c \le Y \le d) = \int_c^d \int_a^b f_{X,Y}(x,y) \, dx \, dy$$



$$P(a \le X \le a + \delta, c \le Y \le c + \delta) \approx f_{X,Y}(a,c) \cdot \delta^2$$

$$f_{X,Y}(x,y)$$
: probability per unit area

$$area(B) = 0 \Rightarrow P((X,Y) \in B) = 0$$

From the joint to the marginals

$$p_X(x) = \sum_y p_{X,Y}(x,y)$$

$$f_X(x) = \int f_{X,Y}(x,y) \, dy$$

$$p_Y(y) = \sum_x p_{X,Y}(x,y)$$

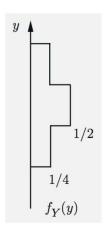
$$f_Y(y) = \int f_{X,Y}(x,y) \, dx$$

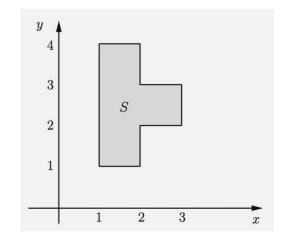
Uniform joint PDF on a set S

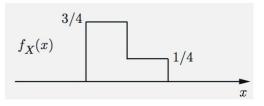
$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{\text{area of } S}, & \text{if } (x,y) \in S, \\ 0, & \text{otherwise.} \end{cases}$$

$$f_X(x) = \int f_{X,Y}(x,y) \, dy$$

$$f_Y(y) = \int f_{X,Y}(x,y) dx$$







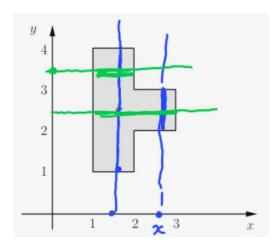
Uniform joint PDF on a set S

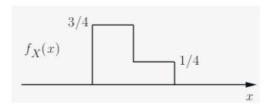
$$f_{X,Y}(x,y) = \begin{cases} \frac{1}{\text{area of } S}, & \text{if } (x,y) \in S, \\ 0, & \text{otherwise.} \end{cases}$$

$$\frac{1}{\text{area of } S}, & \text{otherwise.}$$

$$\frac{1}{\text{area of } S}, & \text{otherwise.}$$

$$\frac{1}{\text{area of } S}, & \text{otherwise.}$$





Conditional PDFs, given another r.v.

$$p_{X|Y}(x \mid y) = P(X = x \mid Y = y) = \frac{p_{X,Y}(x,y)}{p_{Y}(y)}, \quad \text{if } p_{Y}(y) > 0$$

$$p_{X|X}(x \mid y) = P(X = x \mid Y = y) = \frac{p_{X,Y}(x,y)}{p_{Y}(y)}, \quad \text{if } p_{Y}(y) > 0$$

$$p_{X|Y}(x \mid y) = \frac{p_{X|X}(x)}{p_{X|Y}(x \mid y)}$$

$$egin{align} p_{X,Y}(x,y) & f_{X,Y}(x,y) \ & p_{X\mid A}(x) & f_{X\mid A}(x) \ & p_{X\mid Y}(x\mid y) & f_{X\mid Y}(x\mid y) \ \end{array}$$

Definition:
$$f_{X|Y}(x \mid y) = \frac{f_{X,Y}(x,y)}{f_Y(y)}$$
 if $f_Y(y) > 0$

$$P(x \le X \le x + \delta | A) \approx f_{X|A}(x) \cdot \delta$$
, where $P(A) > 0$

$$P(x \le X \le x + \delta \mid y \le Y \le y + \epsilon)$$

Definition:
$$P(X \in A \mid Y = y) = \int_A f_{X \mid Y}(x \mid y) dx$$

Comments on conditional PDFs

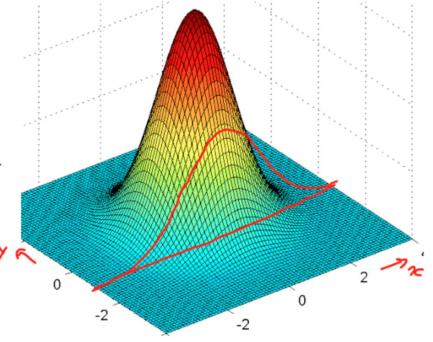
$$f_{X|Y}(x \mid y) = \frac{f_{X,Y}(x,y)}{f_{Y}(y)}$$

•
$$f_{X|Y}(x \mid y) \ge 0$$

• Think of value of Y as fixed at some y shape of $f_{X|Y}(\cdot \mid y)$: slice of the joint

Multiplication rule:

$$f_{X,Y}(x,y) = f_Y(y) \cdot f_{X|Y}(x \mid y)$$
$$= f_X(x) \cdot f_{Y|X}(y \mid x)$$



Independence

$$p_{X,Y}(x,y) = p_X(x) p_Y(y)$$
, for all x, y

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

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

• equivalent to: $f_{X|Y}(x|y) = f_X(x)$, for all y with $f_Y(y) > 0$ and all x

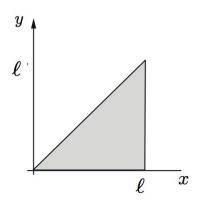
If X, Y are independent:
$$E[XY] = E[X]E[Y]$$

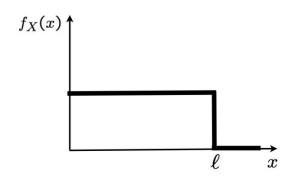
$$var(X + Y) = var(X) + var(Y)$$

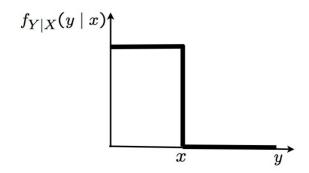
g(X) and h(Y) are also independent: $\mathbf{E}[g(X)h(Y)] = \mathbf{E}[g(X)] \cdot \mathbf{E}[h(Y)]$

- Break a stick of length ℓ twice
 - first break at X: uniform in $[0,\ell]$
 - second break at Y: uniform in [0, X]

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

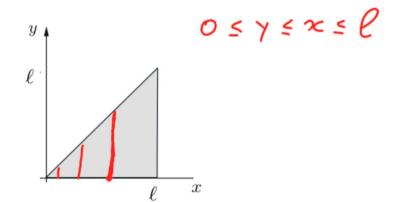


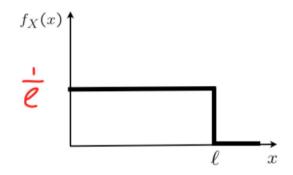


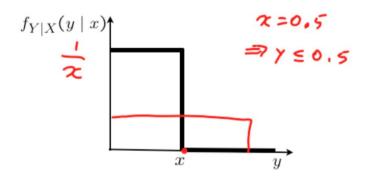


- O T X
- Break a stick of length ℓ twice
- first break at X: uniform in $[0,\ell]$
- second break at Y: uniform in [0, X]

$$f_{X,Y}(x,y) = f_X(x)f_{Y|X}(y \mid x) = \frac{1}{\ell x}$$





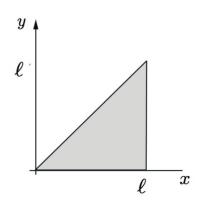


$$f_{X,Y}(x,y) = \frac{1}{\ell x}, \qquad 0 \le y \le x \le \ell$$

$$f_Y(y) =$$

$$\mathbf{E}[Y] =$$

Using total expectation theorem:



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

$$f_{Y}(y) = \int_{\ell} f_{X,Y}(z,\gamma) dz = \int_{\ell} \frac{1}{\ell z} dz = \frac{1}{\ell} \log \left(\frac{\ell}{\gamma}\right)$$

$$E[Y] = \int_{0}^{\ell} \gamma \frac{1}{\ell} \log \left(\frac{\ell}{\gamma}\right) d\gamma$$

Using total expectation theorem:

• Using total expectation theorem:
$$E[Y] = \int_{0}^{1} \frac{1}{e} E[Y|X=x] dx = \int_{0}^{1} \frac{x}{2} dx = \frac{1}{2} E[x] = \frac{1}{2} \cdot \frac{e}{2} = \frac{1}{4}$$

