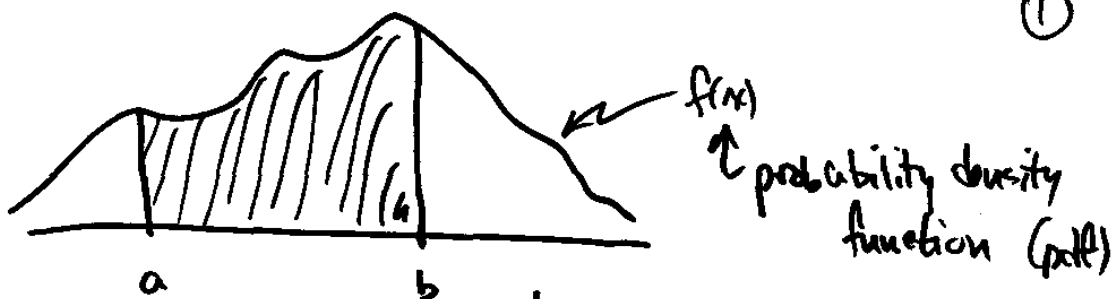


Continuous random variables

Stat 451
10-18-16

①



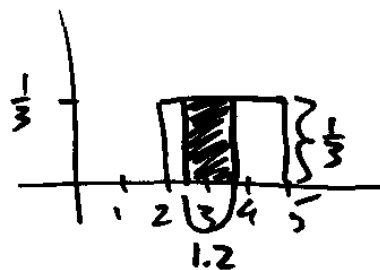
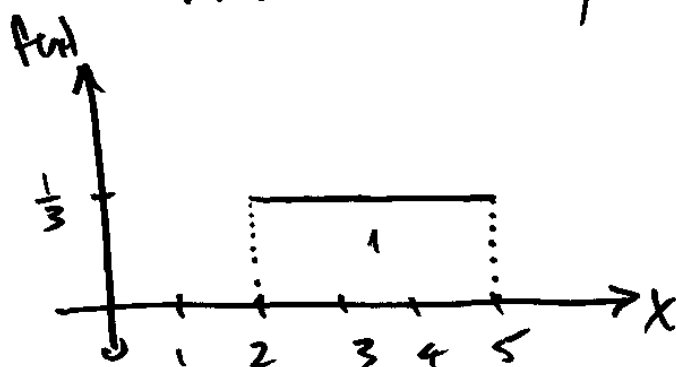
$$P(a < X < b) = \int_a^b f(x) dx$$

Properties of pdfs:

1. $f(x) \geq 0 \quad \forall x$
2. $\int_{-\infty}^{\infty} f(x) dx = 1$

②

Example: Suppose that X takes on all possible values from 2 to 5, with constant density throughout.



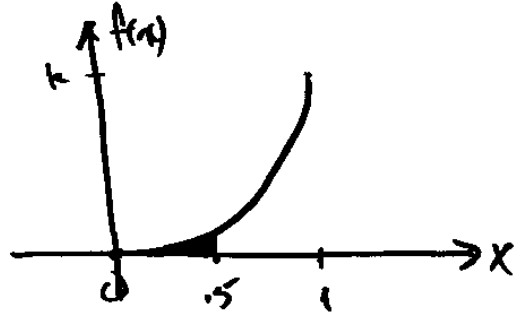
Find the prob. that X lies between 2.5 and 3.7

$$P(2.5 < X < 3.7) = \int_{2.5}^{3.7} \frac{1}{3} dx = \frac{1}{3} x \Big|_{2.5}^{3.7} = \frac{1}{3} (3.7 - 2.5) = .4$$

Example: $f(x) = kx^2$ $0 < x < 1$ ($f(x) = 0$ elsewhere) (3)

① Find k

② $P(X < .5)$



$$\begin{aligned} \textcircled{1} \quad \int_0^1 kx^2 dx &= 1 = k \left. \frac{x^3}{3} \right|_0^1 \\ &= \frac{k}{3}(1-0) = \frac{k}{3} \quad \therefore k=3 \end{aligned}$$

$$\textcircled{2} \quad P(X < .5) = \int_0^{.5} 3x^2 dx = x^3 \Big|_0^{.5} = (.5)^3 - 0 = .125$$

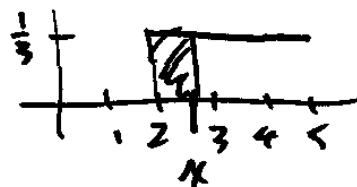
The cumulative distribution function (cdf) (4)
was $P(X \leq x) = F(x)$

For continuous random variables, $F(x) = \int_{-\infty}^x f(t) dt$

Look at the graphs of our 2 examples.

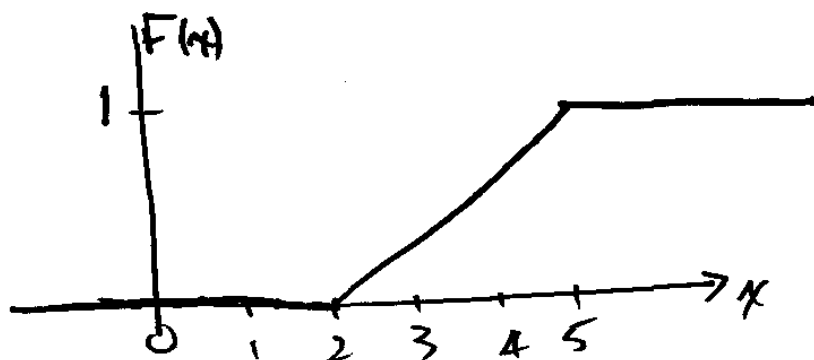
Ex1: $f(x) = \frac{1}{3}$, $2 < x < 5$

$$F(x) = \begin{cases} 0 & x \leq 2 \\ \int_2^x \frac{1}{3} dt & 2 < x < 5 \\ 1 & x \geq 5 \end{cases}$$



(5)

$$F(x) = \begin{cases} 0 & x \leq 2 \\ \frac{1}{3}(x-2) & 2 < x < 5 \\ 1 & x \geq 5 \end{cases} \quad \frac{1}{3}t \Big|_2^x = \frac{1}{3}(x-2)$$



Note: this is nondecreasing & continuous

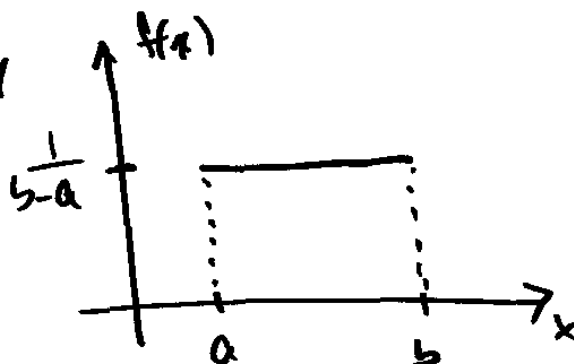
(6)

The uniform distribution

X takes on values in the interval (a, b)

with constant density

$$f(x) = \frac{1}{b-a}, \quad a < x < b$$



Recall $\mu = E(X) = \begin{cases} \sum_{\text{all } x} x p(x) & X \text{ discrete} \\ \int_{-\infty}^{\infty} x f(x) dx & X \text{ continuous} \end{cases}$

For the uniform distribution,

(7)

$$\mu = \int_a^b x \frac{1}{b-a} dx = \frac{1}{b-a} \frac{x^2}{2} \Big|_a^b$$

$$= \frac{1}{b-a} \frac{1}{2} (b^2 - a^2)$$

$$= \frac{1}{\cancel{b-a}} \frac{1}{2} (\cancel{b-a})(b+a) = \frac{a+b}{2}$$

Recall $\sigma^2 = \text{Var}(X) = E(X^2) - \mu^2$

$$\text{and } E(X^2) = \begin{cases} \sum_{\text{all } x} x^2 p(x) & X \text{ discrete} \\ \int_{-\infty}^{\infty} x^2 f(x) dx & X \text{ continuous} \end{cases}$$

$$E(X^2) = \int_a^b x^2 \frac{1}{b-a} dx$$

(8)

$$= \frac{1}{b-a} \frac{x^3}{3} \Big|_a^b = \frac{1}{b-a} \frac{1}{3} (b^3 - a^3)$$

$$= \frac{1}{\cancel{b-a}} \cdot \frac{1}{3} (\cancel{b-a})(b^2 + ab + a^2) = \frac{1}{3} (b^2 + ab + a^2)$$

$$\sigma^2 = \frac{1}{3} (b^2 + ab + a^2) - \left[\frac{1}{2} (a+b) \right]^2$$

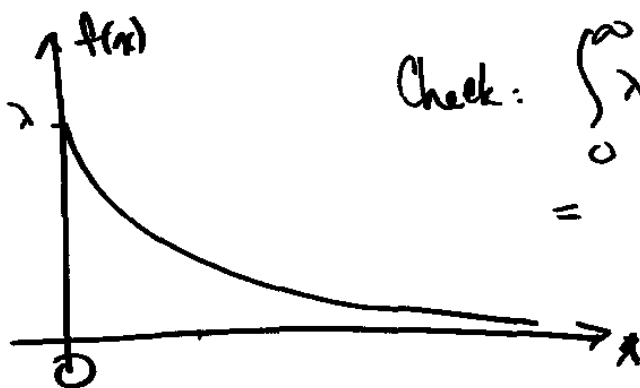
$$= \frac{1}{3} (b^2 + ab + a^2) - \frac{1}{4} (a^2 + 2ab + b^2)$$

$$= \frac{b^2 - 2ab + a^2}{12} = \frac{(b-a)^2}{12}$$

(9)

The exponential distribution

$$f(x) = \lambda e^{-\lambda x} \quad x \geq 0$$



$$\begin{aligned} \text{Check: } \int_0^{\infty} \lambda e^{-\lambda x} dx &= \lambda \left. \frac{e^{-\lambda x}}{-\lambda} \right|_0^{\infty} \\ &= -(0 - 1) \\ &= 1 \quad \checkmark \end{aligned}$$

(10)

Find $F(x)$, μ , σ^2

$$\begin{aligned} F(x) &= \begin{cases} 0 & x < 0 \\ \int_0^x \lambda e^{-\lambda t} dt & x \geq 0 \end{cases} \\ &= \left. \lambda \frac{e^{-\lambda t}}{-\lambda} \right|_{t=0}^x = -(e^{-\lambda x} - 1) \\ &= 1 - e^{-\lambda x} \end{aligned}$$

$$F(x) = \begin{cases} 0 & x < 0 \\ 1 - e^{-\lambda x} & x \geq 0 \end{cases}$$

$$\mu = \int_0^{\infty} x \lambda e^{-\lambda x} dx = \lambda \int_0^{\infty} \underline{x e^{-\lambda x}} dx \quad (11)$$

$$= \lambda \left[\frac{x e^{-\lambda x}}{-\lambda} \Big|_0^{\infty} - \int_0^{\infty} \frac{e^{-\lambda x}}{-\lambda} dx \right] \quad \begin{array}{l} u = x \\ dv = e^{-\lambda x} dx \\ \int_a^b u dv = uv \Big|_a^b - \int_a^b v du \end{array}$$

$$= -\left(\frac{x}{e^{\lambda x}}\right) \Big|_0^{\infty} + \frac{e^{-\lambda x}}{-\lambda} \Big|_0^{\infty} \quad \begin{array}{l} du = dx \\ v = \frac{e^{-\lambda x}}{-\lambda} \end{array}$$

$$= 0 + (-\frac{1}{\lambda})(0 - 1) \quad \begin{array}{l} \text{same limit as } \frac{1}{\lambda e^{\lambda x}} \\ \therefore \mu = \frac{1}{\lambda} \end{array}$$

$$E[X^2] = \int_0^{\infty} x^2 \lambda e^{-\lambda x} dx = \lambda \int_0^{\infty} \underline{x^2 e^{-\lambda x}} dx \quad (12)$$

$$= \lambda \left[\frac{x^2 e^{-\lambda x}}{-\lambda} \Big|_0^{\infty} - \int_0^{\infty} \frac{2x e^{-\lambda x}}{-\lambda} dx \right] \quad \begin{array}{l} u = x^2 \quad du = 2x dx \\ dv = e^{-\lambda x} dx \\ v = \frac{e^{-\lambda x}}{-\lambda} \end{array}$$

$$= -\frac{x^2}{e^{\lambda x}} \Big|_0^{\infty} + 2 \frac{1}{\lambda} \underbrace{\int_0^{\infty} \lambda x e^{-\lambda x} dx}_{\mu}$$

$$= 0 + 2 \cdot \frac{1}{\lambda} \cdot \frac{1}{\lambda} = \frac{2}{\lambda^2} \quad \left| \begin{array}{l} \therefore \sigma^2 = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} \\ = \frac{1}{\lambda^2} \end{array} \right.$$

(13)

Relationship between the Poisson and
Exponential distributions

↑ continuous ↓ discrete

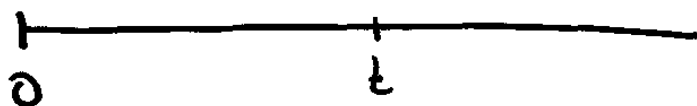
Assume $X \sim \text{Poisson}(\alpha)$ α expected # occurrences per unit time

X counts the # of occurrences of a certain type in a time interval.

Let T be time until the 1st occurrence.

(14)

Find $P(T > t) = P(X = 0)$
in the $(0, t)$ interval



$$P(X=0) = \frac{(\alpha t)^0 e^{-\alpha t}}{0!} = e^{-\alpha t}$$

α was the expected # of occurrences per unit time, so the expected # in time t is αt

$$P(T > t) = e^{-\alpha t}$$

$$\therefore F(t) = P(T \leq t) = 1 - e^{-\alpha t}$$

(15)

$$\therefore f(t) = -e^{-\alpha t}(-\alpha)$$

$= \alpha e^{-\alpha t}$, which is an
exponential pdf.