

CONTINUOUS RANDOM VARIABLES

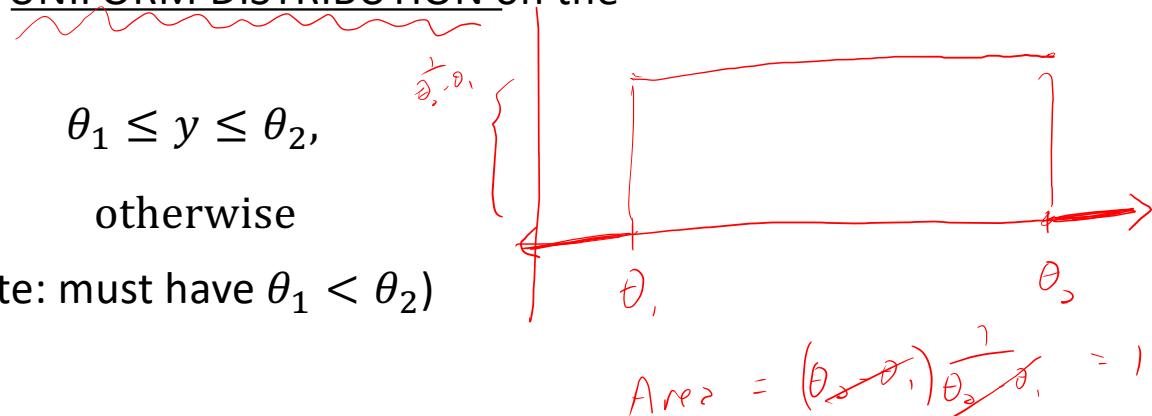
The uniform distribution

The uniform distribution

- Example 6: A bus is scheduled to arrive between 8:00am and 8:10am. The probability that it will arrive in any time interval is proportional to the length of the interval. If you arrive at the bus stop at 8:00am, let \underline{Y} be the time you wait until the bus arrives.
- This is an example of a continuous distribution.
- Definition 4: A random variable \underline{Y} has a UNIFORM DISTRIBUTION on the interval (θ_1, θ_2) if its pdf is

$$f(y) = \begin{cases} \frac{1}{\theta_2 - \theta_1}, & \theta_1 \leq y \leq \theta_2, \\ 0, & \text{otherwise} \end{cases}$$

- Shorthand notation: $Y \sim U(\theta_1, \theta_2)$. (Note: must have $\theta_1 < \theta_2$)



$$U(0, 10)$$

$$\theta_1 = 0$$

$$\theta_2 = 10$$

$$\frac{1}{10-0} = \frac{1}{10}$$

The uniform distribution

- Example 7: For the bus example, calculate the probability that the bus arrives (a) between 8:02 and 8:05; (b) after 8:06.

$$(a) P(2 < Y < 5) = \int_{2}^{5} \frac{1}{10} dy = \left[\frac{1}{10} y \right]_2^5 = \frac{1}{10} (5 - 2) = \frac{3}{10}$$

Rectangular $(5 - 2) \frac{1}{10} = \frac{3}{10}$

$$(b)$$

$$P(6 < Y < 10) = (10 - 6) \frac{1}{10} = \frac{4}{10}$$

$$f(y) = \begin{cases} \frac{1}{\theta_2 - \theta_1}, & \theta_1 < y < \theta_2 \\ 0, & \text{otherwise} \end{cases}$$

The uniform distribution

- Example 8: Find the cdf of Y if $Y \sim U(\theta_1, \theta_2)$.

$$F(y) = P(Y \leq y)$$

CASE I

$$y < \theta_1$$

$$P(Y \leq y) = 0$$

CASE III

$$y > \theta_2$$

$$P(Y \leq y) = 1$$

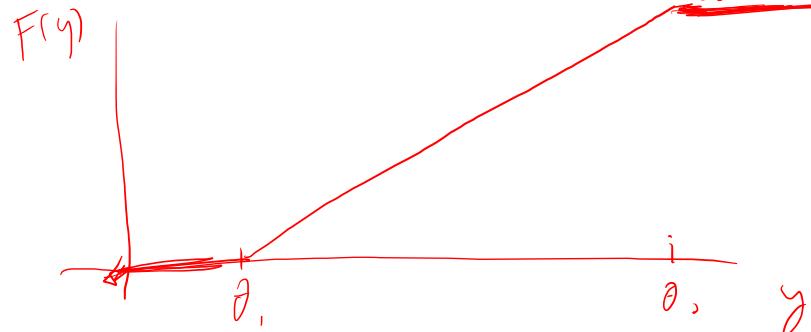
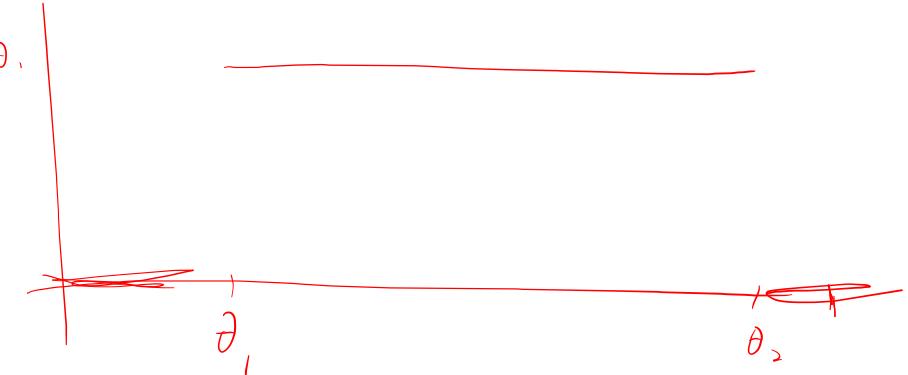
CASE II

$$\theta_1 < y < \theta_2$$

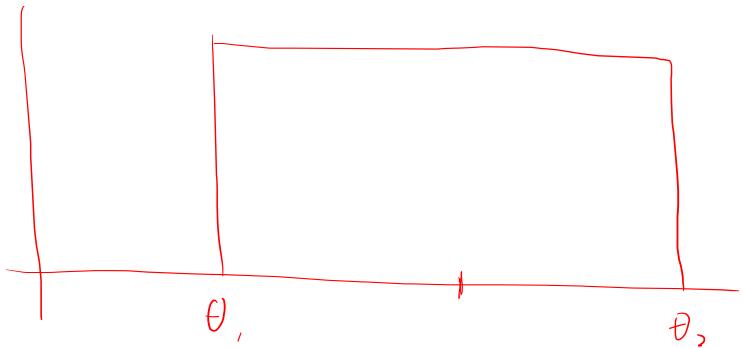
$$P(Y \leq y) = \int_{-\infty}^y f(t) dt$$

$$= \int_{\theta_1}^y \frac{1}{\theta_2 - \theta_1} dt = \left[\frac{1}{\theta_2 - \theta_1} t \right]_{\theta_1}^y = \frac{1}{\theta_2 - \theta_1} (y - \theta_1)$$

$$F(y) = \begin{cases} 0, & y < \theta_1 \\ \frac{y - \theta_1}{\theta_2 - \theta_1}, & \theta_1 \leq y \leq \theta_2 \\ 1, & y > \theta_2 \end{cases}$$



The uniform distribution



- Theorem 3: If $Y \sim U(\theta_1, \theta_2)$ then

$$\mu = \boxed{E[Y] = \frac{\theta_1 + \theta_2}{2}} \text{ and } \sigma^2 = \boxed{Var(Y) = \frac{(\theta_2 - \theta_1)^2}{12}}.$$

$$E[Y] = \int_{-\infty}^{\infty} y f(y) dy = \int_{\theta_1}^{\theta_2} y \frac{1}{\theta_2 - \theta_1} dy \xrightarrow{\text{factoring}} \frac{1}{\theta_2 - \theta_1} \left(\frac{1}{2} y^2 \Big|_{\theta_1}^{\theta_2} \right) = \frac{1}{\theta_2 - \theta_1} \cdot \frac{1}{2} (\theta_2^2 - \theta_1^2)$$

$$E[Y^2] = \frac{1}{\theta_2 - \theta_1} \int_{\theta_1}^{\theta_2} y^2 dy = \frac{1}{\theta_2 - \theta_1} \left[\frac{1}{3} y^3 \right]_{\theta_1}^{\theta_2} \stackrel{\checkmark}{=} \frac{1}{3} (\theta_2 - \theta_1)^2$$

$$Var(Y) = E[Y^2] - (E[Y])^2 = \frac{1}{3} (\theta_2 - \theta_1)^2 - \frac{1}{4} (\theta_1 + \theta_2)^2 = \dots = \frac{1}{12} (\theta_2 - \theta_1)^2$$

$\boxed{\quad}$

The uniform distribution

- Any probability distribution is determined by its *parameters*.
- The *parameters* of the $U(\theta_1, \theta_2)$ distribution are θ_1 and θ_2 .
- If a uniform random variable U has $\theta_1 = 0$ and $\theta_2 = 1$ (i.e., $U \sim U(0,1)$) then we say that U has the standard uniform distribution.

$$f(y) = \begin{cases} 1 & , 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

Moment generating functions for continuous random variables

- Moments and mgfs for continuous random variables are defined just as for discrete random variables.
- The k th moment [taken about the origin] of a random variable is $\mu'_k = E[Y^k]$.
- The k th central moment (or k th moment about the mean) of a random variable is $\mu_k = E[(Y - \mu)^2]$.
- The moment generating function for a random variable Y is $m(t) = E[e^{tY}]$. (For this to exist there must be some constant $b > 0$ such that $m(t)$ is finite for $-b < t < b$.)

Moment generating function for the uniform distribution

$$Y \sim U(\theta_1, \theta_2)$$

- Example 9: Find the mgf for the ~~standard~~ uniform random variable $Y \sim U(\theta_1, \theta_2)$ and determine whether it exists.

$$\begin{aligned} m(t) &= E[e^{tY}] = \int_{\theta_1}^{\theta_2} \frac{1}{\theta_2 - \theta_1} e^{ty} dy \\ &= \frac{1}{\theta_2 - \theta_1} \left(\frac{1}{t} e^{ty} \Big|_{\theta_1}^{\theta_2} \right) = \frac{e^{\theta_2 t} - e^{\theta_1 t}}{(\theta_2 - \theta_1)t} \end{aligned}$$

+ + +
- b +

~~Any $t \neq 0$~~ finite
 ~~$t = 0?$~~

$$\frac{e^0 - e^0}{(\theta_2 - \theta_1)^0}$$

$$\frac{1-1}{0} = \frac{0}{0} \quad |||$$

indeterminate

$$\frac{\theta_2 e^{\theta_2 t} - \theta_1 e^{\theta_1 t}}{\theta_2 - \theta_1}$$

$$\frac{\theta_2 e^0 - \theta_1 e^0}{\theta_2 - \theta_1} = \frac{\theta_2 - \theta_1}{\theta_2 - \theta_1} = 1$$

$$\lim_{t \rightarrow 0} L'Hopital's \quad \frac{e^{\theta_2 t} - e^{\theta_1 t}}{(\theta_2 - \theta_1)t} = \lim_{t \rightarrow 0} \frac{\frac{d}{dt}(e^{\theta_2 t} - e^{\theta_1 t})}{\frac{d}{dt}(\theta_2 - \theta_1)t} = \lim_{t \rightarrow 0}$$

$\lim_{t \rightarrow 0}$ finite $\Rightarrow Mgf$ exists