
Uniform Distribution

ECE 341: Random Processes

Lecture #12

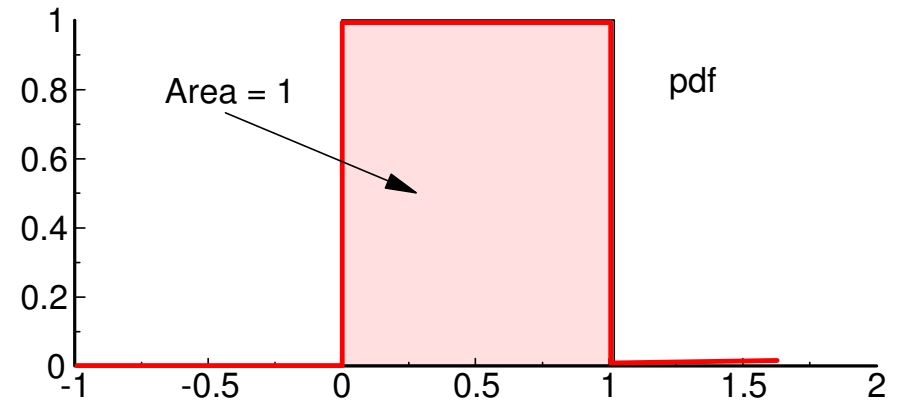
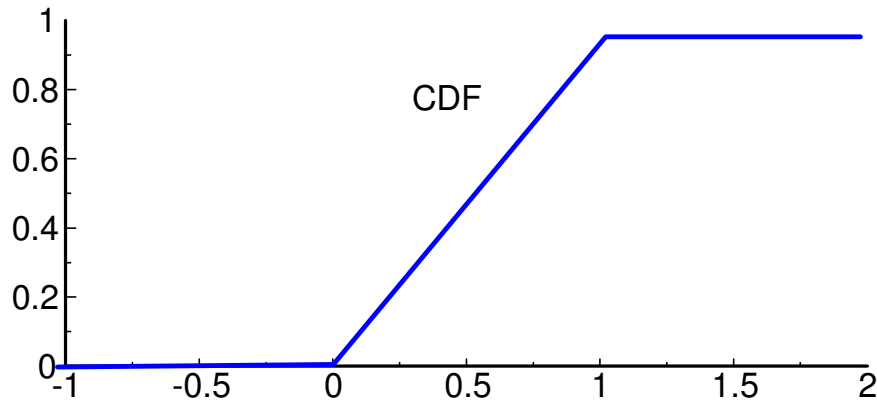
note: All lecture notes, homework sets, and solutions are posted on www.BisonAcademy.com

Uniform Distribution

A uniform distribution is one where the probability density function (pdf) is

- constant over a range (a, b) , and
- zero otherwise.

Example: Uniform distribution over the interval of $(1, 2)$:



Examples:

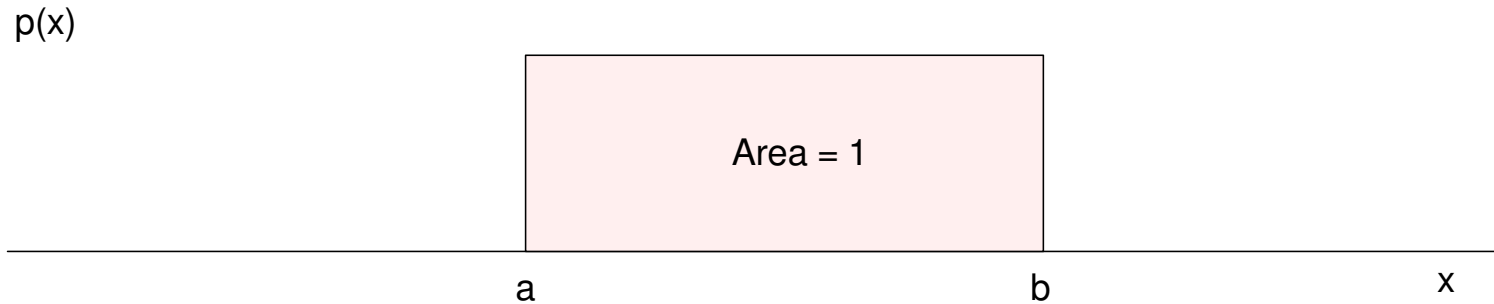
- A 1k resistor with a tolerance of 5% will have a value in the range of +/- 5%. This can be modeled as a uniform distribution over the range of $950 < R < 1050$
- A 3904 transistor has a nominal gain of 200 with a variation of +/- 100. This can be modeled as a uniform distribution over the range if $100 < \text{gain} < 300$

In general, if you know the limits but know nothing else about the distribution, a uniform distribution is a reasonable assumption to make about the random variable.

TYPE NO.	DESCRIPTION	LEAD CODE	V_{CB0} (V) MIN	V_{CEO} (V) $^{*}V_{CES}$ MIN	V_{EBO} (V) MIN	I_{CBO} @ V_{CB} (nA)		h_{FE}		@ V_{CE} (V)	@ I_C (mA)	$V_{CE(SAT)}$ @ I_C		C_{ob} (pF) $^{*}C_{rb}$ MAX	f_T (MHz) $^{*}TYP$ MIN	NF (dB) MAX	t_{off} MAX
						$^{*}I_{CES}$ $^{*}I_{CEV}$ MAX	V_{CB} (V)	MIN	$^{*}h_{FE}$ (1kHz) MAX			$V_{CE(SAT)}$ (V)	$V_{CE(SAT)}$ (mA)				
2N3860	NPN LOW NOISE	ECB	30	30	4.0	50	30	150	300	4.50	2.0	0.125	10	4.0	90	- -	- -
2N3903	NPN AMPL/SWITCH	EBC	60	40	6.0	50*	30	50	150	1.0	10	0.30	50	4.0	250	6.0	225
2N3904	NPN AMPL/SWITCH	EBC	60	40	6.0	50*	30	100	300	1.0	10	0.30	50	4.0	300	5.0	250
2N3905	PNP AMPL/SWITCH	EBC	40	40	5.0	50*	30	50	150	1.0	10	0.40	50	4.5	200	5.0	260
2N3906	PNP AMPL/SWITCH	EBC	40	40	5.0	50*	30	100	300	1.0	10	0.40	50	4.5	250	4.0	300

Properties of Uniform Distributions:

Assume a uniform distribution over the interval (a, b)



Since the area must be 1.000, the height must be

$$\text{Area} = \text{width} * \text{height} = 1$$

$$p(x) = \left(\frac{1}{b-a}\right) \quad a < x < b$$

Mean: The mean of the function (almost by inspection) is

$$\bar{x} = \left(\frac{a+b}{2} \right)$$

Variance:

$$\sigma^2 = \int_a^b p(x) (x - \mu)^2 dx$$

$$\sigma^2 = \left(\frac{(b-a)^2}{12} \right)$$

Moment Generating Function

$$\psi(s) = \left(\frac{1}{(b-a)s} \right) (e^{-as} - e^{-bs})$$



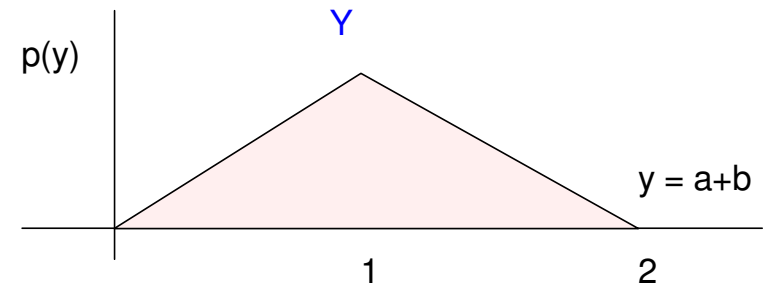
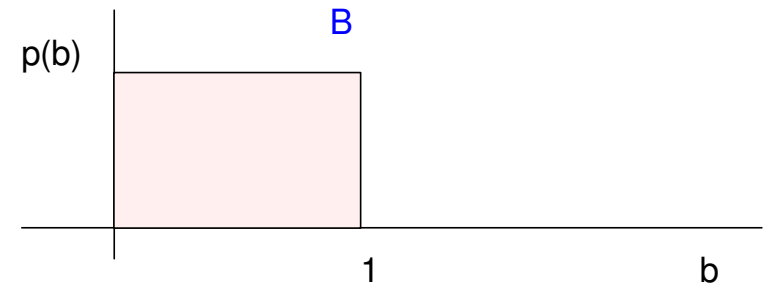
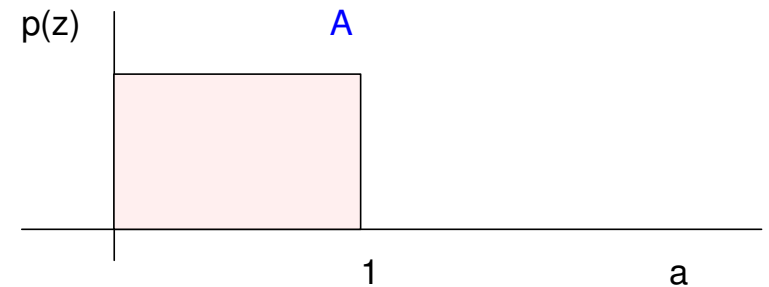
Combinations of Uniform Distributions

If you add two uniform distributions, the result is

- The convolution of their pdf's, or
- The product of their moment generating functions.

Example:

- Assume A and B are uniform distributions over the interval (0, 1).
- Find the pdf of the sum $Y = A + B$.



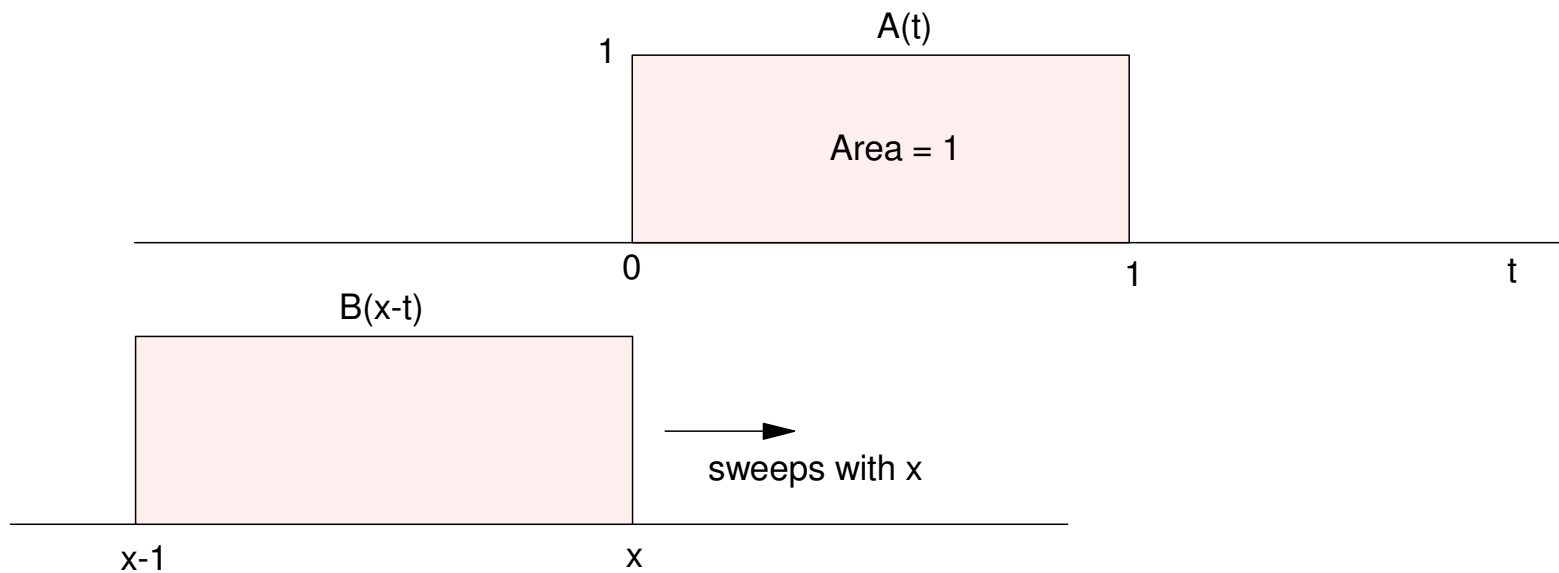
Solution using convolution:

$$A(x) = u(x) - u(x - 1)$$

$$B(x) = u(x) - u(x - 1)$$

$$Y(x) = A(x) * B(x)$$

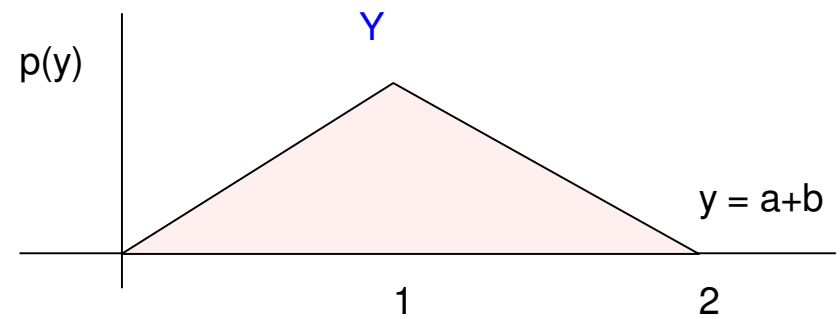
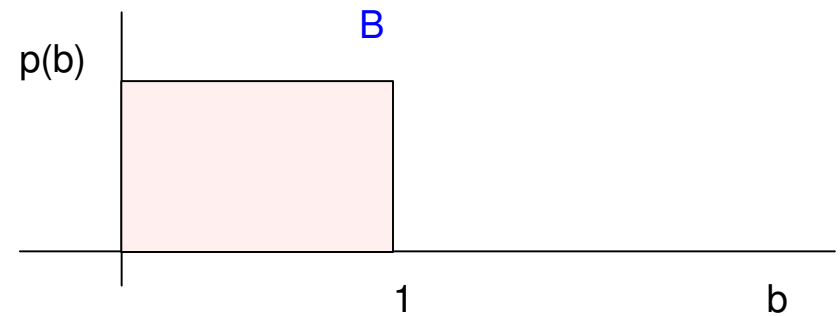
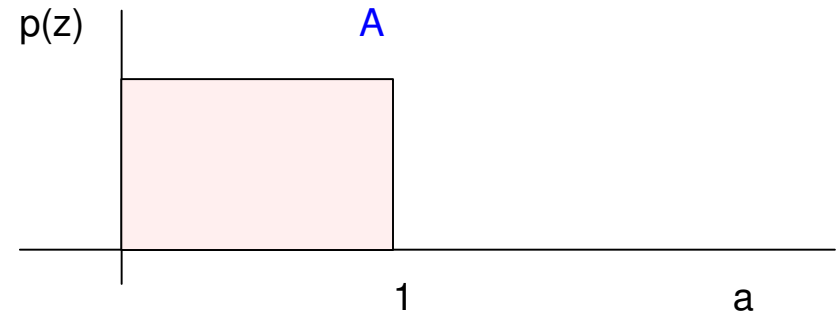
$$Y(x) = \int_{-\infty}^{\infty} A(t)B(x - t)dt$$



From the graph, this integral is

$Y(x) =$

- 0 $x < 0$
- x $0 < x < 1$
- $1 - x$ $1 < x < 2$
- 0 $2 < x$



$y = a+b$

Solution using Calculus:

$$Y(x) = \int_{-\infty}^{\infty} A(t) B(x-t) dt$$

$$Y(x) = \int_{-\infty}^{\infty} (u(t) - u(t-1)) (u(x-t) - u(x-t-1)) dt$$

Multiply out giving four terms

$$\begin{aligned} &= \int_{-\infty}^{\infty} u(t)u(x-t) dt + \int_{-\infty}^{\infty} -u(t)u(x-t-1) dt \\ &+ \int_{-\infty}^{\infty} -u(t-1)u(x-t) dt + \int_{-\infty}^{\infty} u(t-1)u(x-t-1) dt \end{aligned}$$

$$\begin{aligned} &= \int_0^x 1 dt - \int_0^{x-1} 1 dt - \int_1^x 1 dt + \int_1^{x-1} 1 dt \\ &= xu(x) - (x-1)u(x-1) - (x-1)u(x-1) + (x-2)u(x-2) \\ &= xu(x) - 2(x-1)u(x-1) + (x-2)u(x-2) \end{aligned}$$

Solve using moment generating functions

$$A(s) = \left(\frac{1}{s}\right)(1 - e^{-s})$$

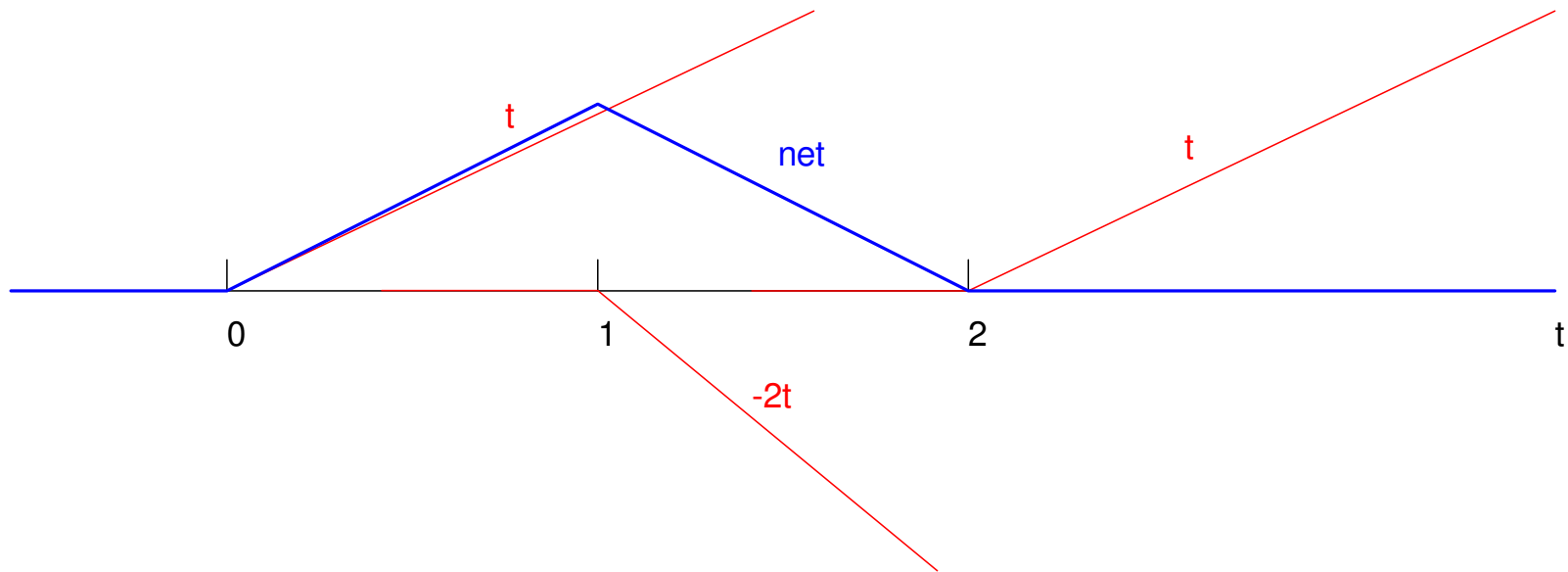
$$B(s) = \left(\frac{1}{s}\right)(1 - e^{-s})$$

$$Y(s) = A(s)B(s)$$

$$Y(s) = \left(\frac{1}{s^2}\right)(e^{-2s} - 2e^{-s} + 1)$$

Take the inverse LaPlace transform

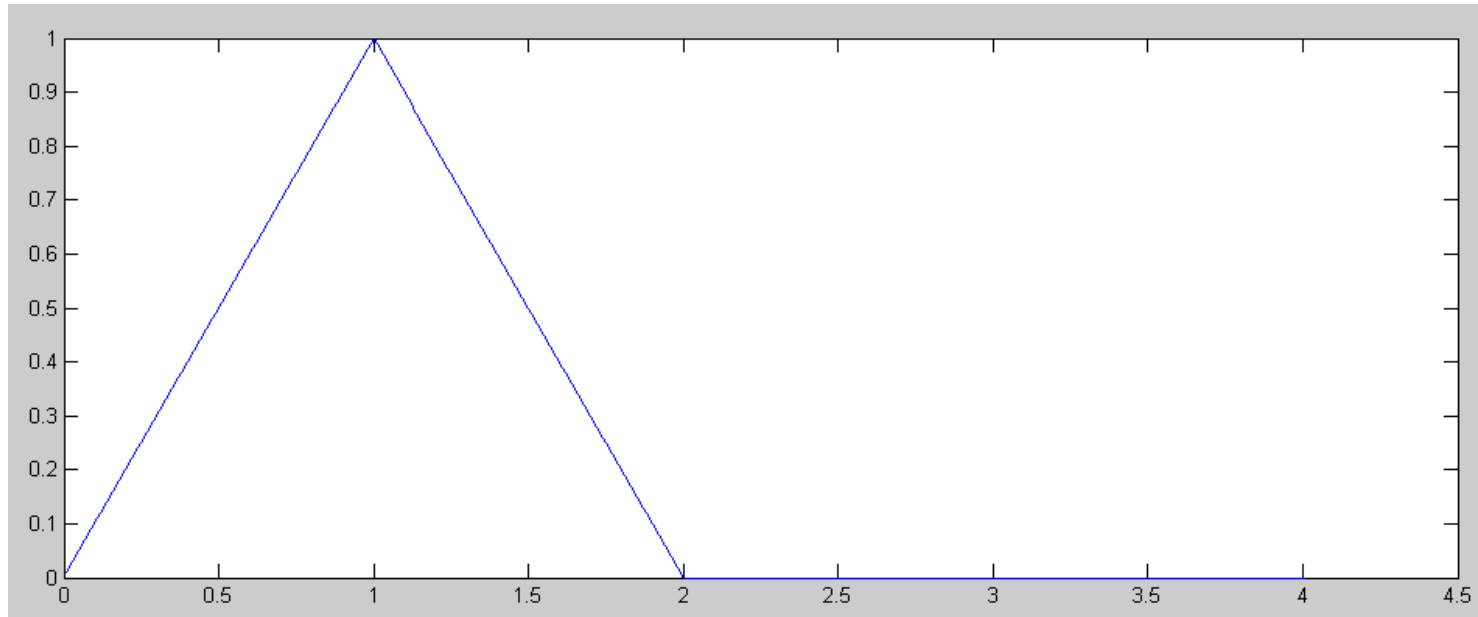
$$y(x) = (x - 2) u(x - 2) - 2(x - 1) u(x - 1) + x u(x)$$



Solve in Matlab

- Approximate a uniform distribution with 100 points over the interval (0, 1)

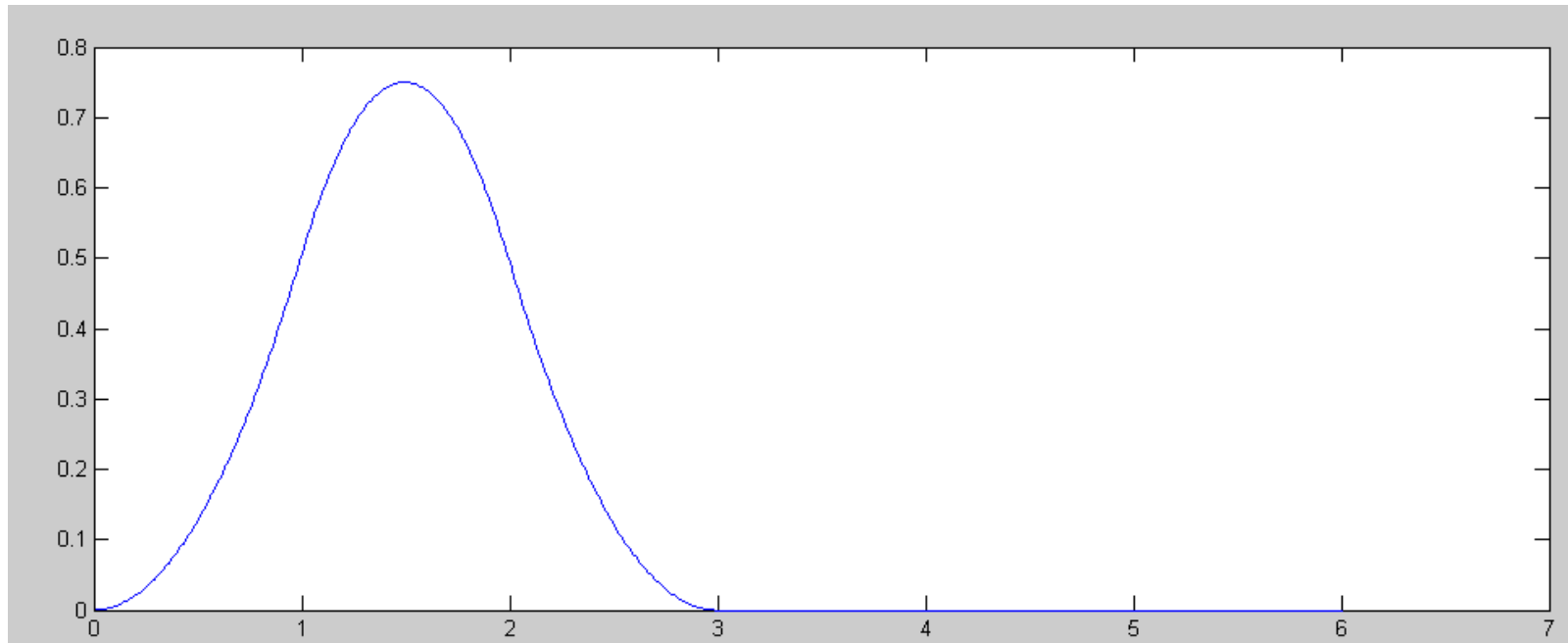
```
dx = 0.01;  
x = [0:dx:2]';  
A = 1*(x<1);  
B = 1*(x<1);  
Y = conv(A, B) * dx;  
plot([1:length(Y)]*dx, Y)
```



Example 2: Find the pdf of summing three uniform distributions

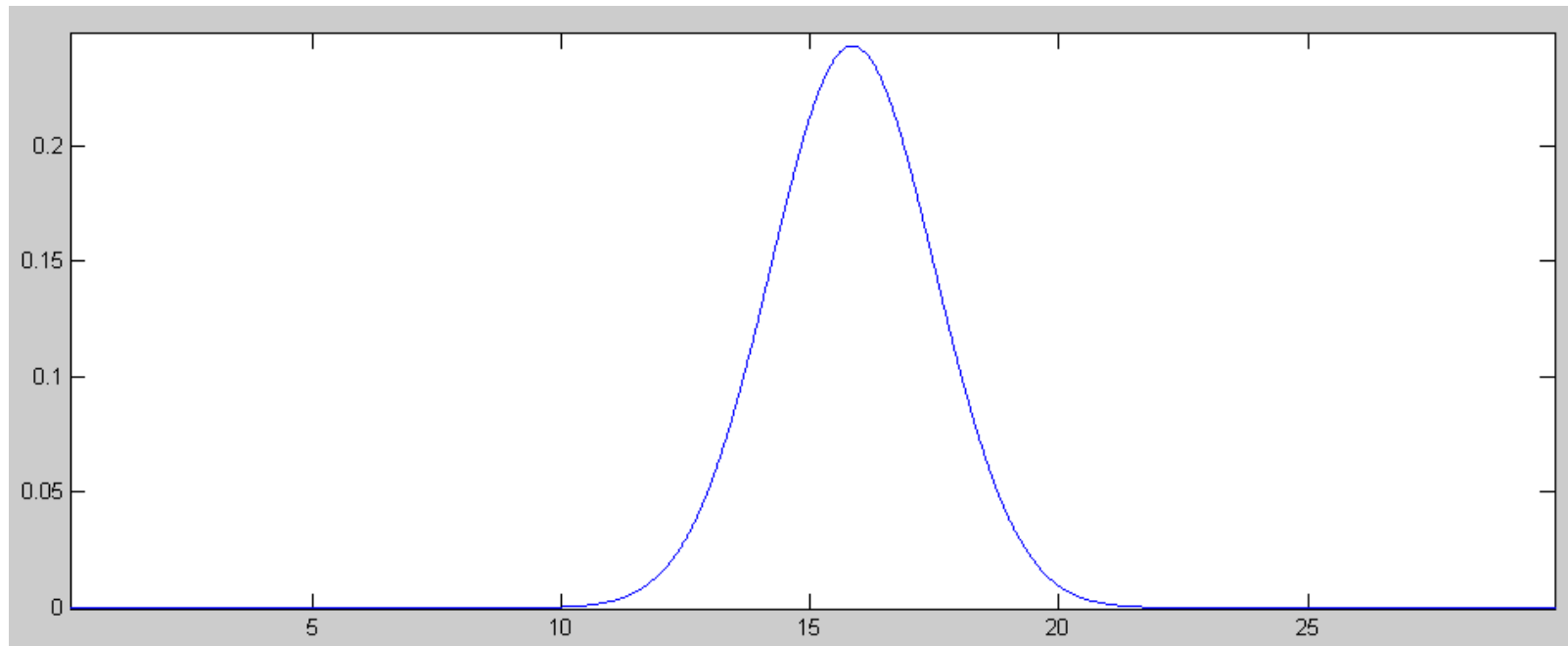
```
x = [0:dx:2]';  
A = 1*(x<1);  
B = 1*(x<1);  
C = 1*(x<1);
```

```
Y = conv(A, B) * dx;  
Y = conv(Y, C) * dx;  
plot([1:length(Y)]*dx, Y)
```



Example 3: Find the pdf for the sum of 32 uniform distributions:

```
x = [0:dx:2]';  
A = 1*(x<1);  
Y2 = conv(A, A) * dx;  
Y4 = conv(Y2, Y2) * dx;  
Y8 = conv(Y4, Y4) * dx;  
Y16 = conv(Y8, Y8) * dx;  
Y32 = conv(Y16, Y16) * dx;  
plot([1:length(Y32)]*dx, Y32)
```



Uniform Distribution in Circuit Analysis:

Determine the pdf for V_{ce} . Assume

- Resistors have a 1% tolerance
- The transistor has a gain in the range of 100 to 300

The equations for this circuit (from ECE 321)

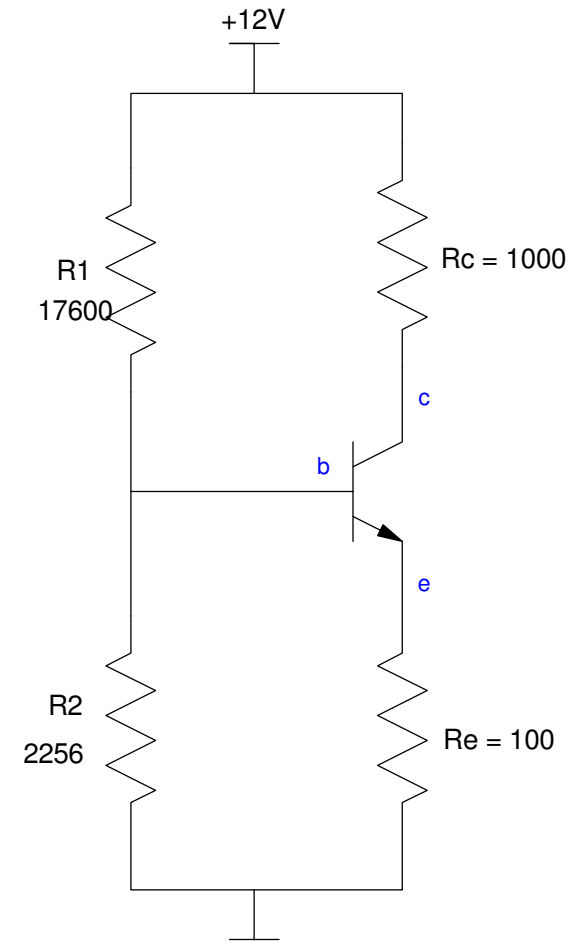
$$V_{th} = \left(\frac{R_2}{R_2 + R_1} \right) 12V$$

$$R_{th} = \left(\frac{R_1 R_2}{R_1 + R_2} \right)$$

$$I_b = \left(\frac{V_{th} - 0.07}{R_{th} + (1 + \beta)R_e} \right)$$

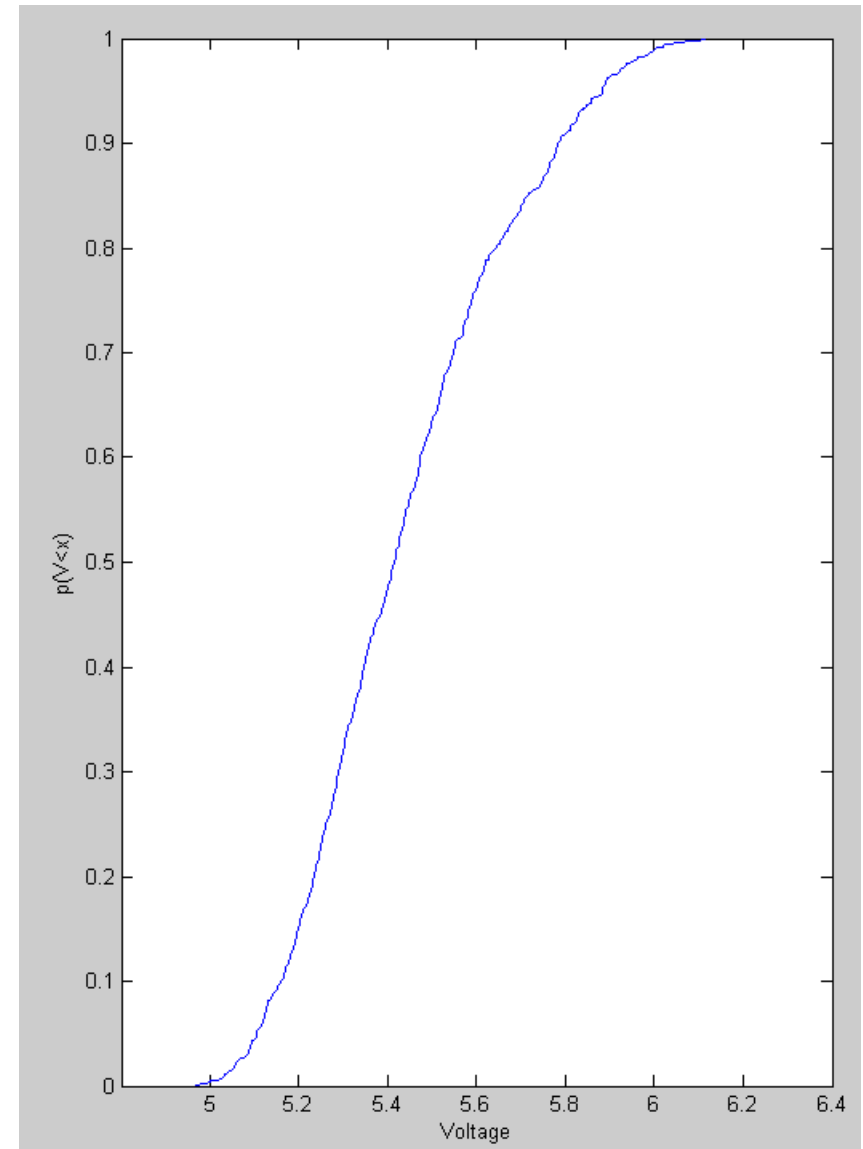
$$I_c = \beta I_b$$

$$V_{ce} = 12 - R_c I_c - R_e (I_b + I_c)$$



Using a Monte-Carlo simulation

```
DATA = [];  
  
for i=1:1000  
    R1 = 17600 * (1 + (rand()*2-1)*0.01);  
    R2 = 2256 * (1 + (rand()*2-1)*0.01);  
    Rc = 1000 * (1 + (rand()*2-1)*0.01);  
    Re = 100 * (1 + (rand()*2-1)*0.01);  
    Beta = 200 + 100*(rand()*2-1);  
    Vb = 12*(R2 / (R1+R2));  
    Rb = 1/(1/R1 + 1/R2);  
    Ib = (Vb-0.7) / (Rb + (1+Beta)*Re);  
    Ic = Beta*Ib;  
    Vce = 12 - Rc*Ic - Re*(Ic+Ib);  
    DATA = [DATA; Vce];  
end  
  
DATA = sort(DATA);  
  
p = [1:length(DATA)]' / length(DATA);  
plot(DATA, p)  
xlabel('Voltage');  
ylabel('p(V<x)')
```

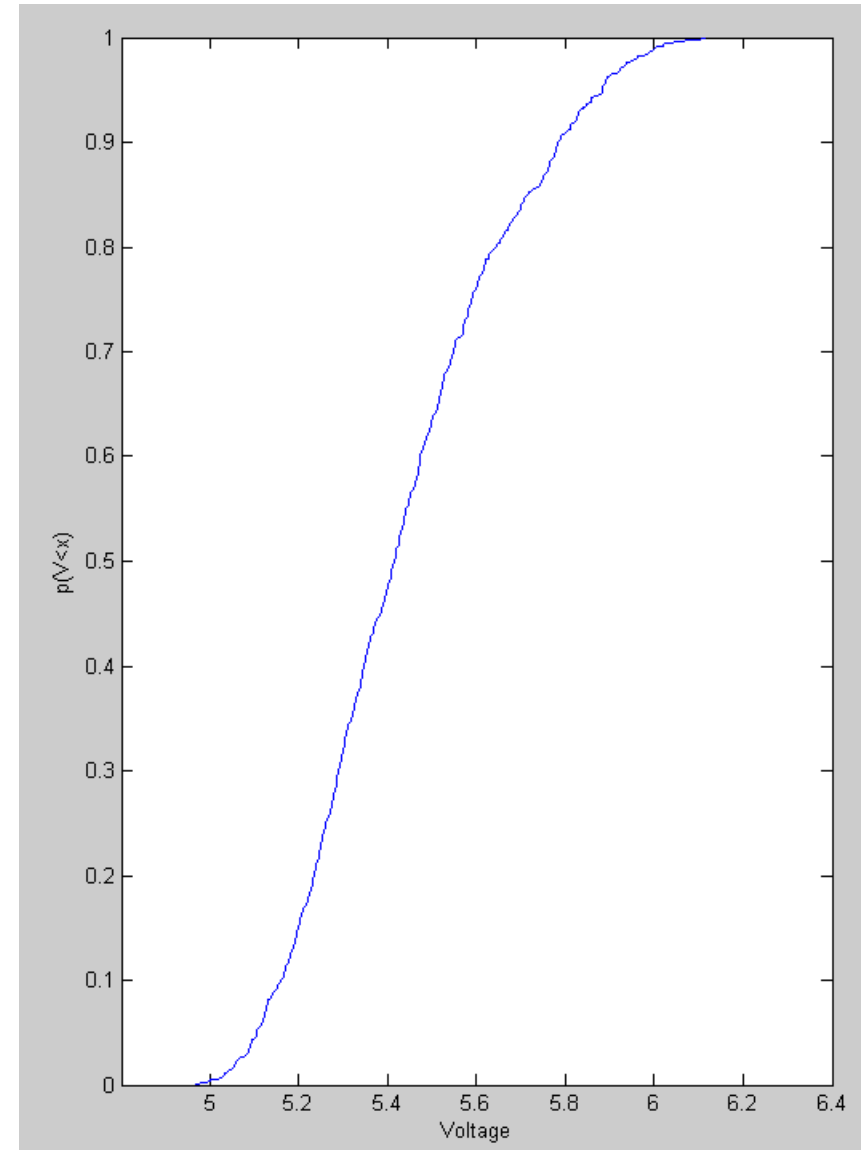


Notes:

- The resulting plot is the cdf of the voltage
- Differentiate it and (in theory) you get the pdf

Due to addition, division, multiplication, etc. it isn't clear what this distribution is.

- Later, we'll fit a Weibull distribution to this data



Summary

Uniform distributions are fairly common and easy to model

- `rand()` function in Matlab

The sum of uniform distributions converges to a Normal distribution

- Central Limit Theorem (coming soon)

Convolution is required to sum uniform distributions

- `conv()` in Matlab
- Or you can use multiplication if using Laplace transforms
 - a.k.a. moment generating functions in statistics