## BLG 354E SIGNALS & SYSTEMS FOR COMPUTER ENGINEERING

HOMEWORK 4 REPORT

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## Question 1

Multiple convolutions of a signal produce an output that has the shape of a Gaussian curve. Proof:

CLT coplane the bell shaped curve

let 
$$X_i$$
 be our random variable, and real value of  $X \to \hat{x}$ 

Prob  $(a \le X \le b) = \int_{p}^{p} p(x) dx$  and  $\int_{-\infty}^{\infty} p(x) dx = 1$ 

Suppose we have 2 independent variables  $X_i$   $Y_{X_i}$  with  $p_i(x_i)$   $p_2(x_2)$ 

How will sim  $X_i + X_2$  be obstituted?

For any value of  $t$ 
 $p(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i dx_2$  with and  $p_i(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i dx_2$  with and  $p_i(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i dx_2$  with and  $p_i(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i dx_2$  with random variable  $p_i(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i dx_i$ 

We have extracted convolution

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From parabolity distribution.

The probability distribution.

Since  $p(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i$ 
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They have some distributions)

From  $p(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i$ 
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Since they sell have the one of  $p_i(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i$ 

Distribution of  $p(x) = \sum_{i=1}^{n} p_i(x) p_i(x) dx_i$ 

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The first  $p_i(x) = \sum_{i$ 

Take FFT to enolyze convolutes were easily:

$$\begin{aligned}
& \mp T \left( \sqrt{n} \, p^{kn} \left( \sqrt{n} \, x \right) \right) = \left( \mp T \left( p \right) \right)^n \left( \frac{S}{n} \right) = \left( \mp T \, p \left( \frac{S}{n} \right) \right) \\
& = \int_{-\infty}^{\infty} e^{-2\pi j} \left( \frac{S}{n} \right)^n x \rho(x) \, dx \quad \text{Using Toylor (area)} \\
& = \int_{-\infty}^{\infty} \left( 1 - \frac{2\pi S}{n} \right)^n - \left( \frac{2\pi S}{n} \right)^n \frac{1}{2} + \dots - \int_{-\infty}^{\infty} p(x) \, dx \\
& = \int_{-\infty}^{\infty} p(x) \, dx - \frac{2\pi S}{n} \int_{-\infty}^{\infty} x p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{-\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx - \frac{2\pi S^2}{n} \int_{\infty}^{\infty} x^2 p(x) \, dx + \dots - \int_{\infty}^{\infty} p(x) \, dx + \dots - \int_{\infty}^{\infty}$$

See code for graph plots.

From the plots it can be deduced that longer square waves obtain Gaussian form faster than smaller ones. (Low pass filter with higher bandwidth approaches to Gaussian plot much faster, within less steps of convolution). Since there are more points to evaluate in each multiplication.

## Question 2

$$\frac{A_{h}}{1} = \frac{1}{1} \int f(t) e^{-jt} \frac{dt}{dt} \qquad X(t) = \sum_{n=1}^{\infty} a_{n} e^{-jnt} \frac{dt}{dt}$$

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For first graph
$$a_{1} = \frac{1}{2j\pi} \left(1 - 3e^{-j\frac{4\pi}{3}} + 2e^{-j2\pi}\right) \quad 42 = \frac{1}{4j\pi} \left(1 - 3e^{-j\frac{4\pi}{3}} + 2e^{-j2\pi}\right)$$

$$a_{3} = \frac{1}{6\pi i} \left(1 - 3e^{-j\frac{4\pi}{3}} + 2e^{-j2\pi}\right)$$
See code for father