



## Central Limit Theorem and Different Estimators

College of Engineering and Computing  
ECE 345: Engineering Statistics Final Project  
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### A) Approximate Theoretical PDF

First we solved for the random variable  $X$  in the equation  $Y = X^2$ . This allows us to find the pdf of  $x$  with respect to  $y$ . Solving the equation gives us:

$$Y = X^2$$
$$X = \pm\sqrt{Y}$$

Then we solved for the expectation of  $Y$ :

$$E(A_n) = E(Y) = \int_0^\infty y f_Y(y) dy$$
$$E(Y) = \int_0^\infty y \frac{1}{\sigma\sqrt{2\pi y}} e^{-\frac{1}{2}(\frac{y}{\sigma^2})} dy = \sigma_x^2$$
$$V(Y) = \int_0^\infty (y - \sigma^2)^2 \frac{1}{\sigma\sqrt{2\pi y}} e^{-\frac{1}{2}(\frac{y}{\sigma^2})} dy = 2\sigma_x^4$$

The pdf of  $x$  with respect to  $y$  will result in the following:

$$f_x(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2}, \mu = 0$$
$$f_x(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{x^2}{2\sigma^2}}, x^2 = y$$

The complete derived equation:

$$f_x(y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{y}{2\sigma^2}}, y > 0$$

According to the central limit theorem the approximate theoretical pdf of  $A_n$  is shown as follows:

$$f(A_n) = \frac{1}{\sqrt{2\pi}\sigma_{A_n}} e^{-\frac{(A_n - \mu_{A_n})^2}{2\sigma_{A_n}^2}}, x^2 = y$$

Using the following values of  $n = 2, 3, 16, 144$ :

$$f(A_2) = N(\sigma_X^2, \sigma_X^4), y > 0$$

$$f(A_3) = N(\sigma_X^2, \frac{2\sigma_X^4}{3}), y > 0$$

$$f(A_{16}) = N(\sigma_X^2, \frac{\sigma_X^4}{8}), y > 0$$

$$f(A_{144}) = N(\sigma_X^2, \frac{\sigma_X^4}{72}), y > 0$$

## B) True PDF of $A_n$ for $n = 2$ without Approximation

$$f(A_n) = \frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{((\frac{1}{2}(x_1+x_2))-\sigma^2)^2}{2\sigma^4}}, y > 0$$

Calculating the bounds so we can solve for the pdf:

$$y > 0$$

$$z - y > 0$$

$$y < z \text{ and } y > 0, \text{ resulting in } 0 > y > z$$

$$A_2 = \frac{1}{2}(y_1 + y_2), \text{ transforming into } Z = X + Y$$

$$f_z(z) = \int \int_{x+y=z} f_{xy}(x, y) dx dy$$

Since  $x$  and  $y$  are independent:

$$f_z(z) = \int \int_{x+y=z} f_x(x) f_y(y) dx dy$$

Resulting in a convolution integral of:

$$f_z(z) = \int f_y(y) f_y(z - y) dy$$

Implementing the bounds:

$$\int_0^z f_y(y) f_y(z - y) dy$$

$$f_y(y) = \frac{1}{\sqrt{2\pi}y\sigma} e^{-\frac{y}{2\sigma^2}}, y > 0$$

$$\begin{aligned}
f(A_2) = f(z) &= \int_0^z \frac{1}{\sqrt{2\pi y}\sigma} e^{-\frac{y}{2\sigma^2}} \frac{1}{\sqrt{2\pi(z-y)}\sigma} e^{-\frac{(z-y)}{2\sigma^2}} dy \\
&= \int_0^z \frac{1}{\sqrt{2\pi y}\sigma} \frac{1}{\sqrt{2\pi(z-y)}\sigma} e^{-\frac{z}{2\sigma^2}} dy \\
&= e^{-\frac{z}{2\sigma^2}} \int_0^z \frac{1}{\sqrt{2\pi y}\sigma} \frac{1}{\sqrt{2\pi(z-y)}\sigma} dy \\
&= \boxed{\frac{e^{-\frac{z}{2\sigma^2}}}{2\sigma^2}}
\end{aligned}$$

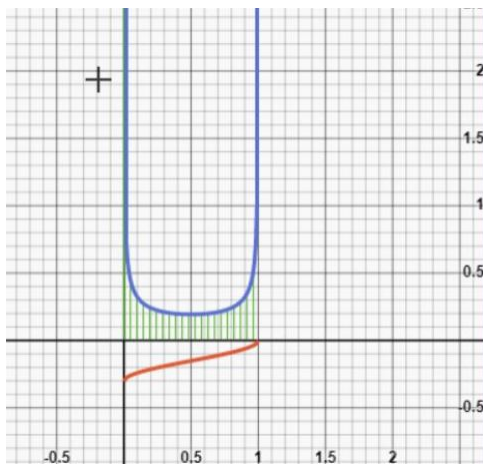


Figure 1: Graph of  $f(A_2) = f(z)$

# Appendix

## Matlab part C using Y

```
1  %%% Matlab part C using Y. %%%
2  clear all
3  x=-3:0.03:3;
4
5  % theoretical pdf of standard norm.
6  y1=normpdf(x, 0, 1);
7  % generate n pts - normal distribution
8  bins=100000;
9  rx=randn(1, bins);
10 [y2]=hist(rx, x)/bins/(x(2)-x(1));
11 %{
12 plot(x, y1, 'o', x, y2, '');
13 grid; legend('theory', 'simulation')
14 xlabel('x'),
15 ylabel('pdf of stand. normal r.v.')
16 %}
17 %{
18 n = 2;
19 y = x.^2;
20 ry = rx.^2;
21 for i = 1:n
22     index = randi([1, bins]);
23     ysamp(i) = rx(index);
24 end
25
26
27 %fy = 1/(sqrt(2pi))*exp(-y/2);
28 %An = sum(ysamp)/n;
29 z = sum(ysamp)/n;
30 fA2 = exp(-z/2)/2;
31 plot(y,fA2)
32 %}
33 y = [0.01:(.03/2):3.01];
34 n = 144;
35 hold on
36 fA = normpdf(y, 1, 2/n);
37 plot(y,fA);
38
39
40 fy = 1./(sqrt(2pi)).*exp(-y./2);
41 bins=1000000;
42
43 samp=100000;
44 for samples = 1:samp
45     for index = 1:n
46
47         fy = fy/sum(fy); % Make sure probabilities add up to 1.
48         cp = [0, cumsum(fy)];
49         r = rand;
50         ind = find(r>cp, 1, 'last');
51         randy = y(ind); % https://www.mathworks.com/matlabcentral/
52             % answers/506654-select-random-number-from-an-array-with-probabilities
53         ysamp(index)=randy;
54     end
55     An(samples)=1/nsum(ysamp);
56
57 end
58 [y2]=hist(An,y)/(samp/(n*10));
59 bar(y,y2)
60 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
```

## Matlab part C using $X^2$

```
1  %% Matlab part C using X^2. %%
2  clear all
3  sz = 0.01;
4  x=-3:sz:3;
5  y = [0:(.01/2):3];
6  n = 16;
7
8  % theoretical pdf of standard norm.
9  y1=normpdf(x, 0, 1);
10
11 % theoretical pdf of Y
12 fA = normpdf(y, 1, 2/n);
13 plot(y,fA);
14 hold on
15
16 %fy = 1./(sqrt(2*pi)).exp(-y./2);
17
18 samp=100000;
19 for samples = 1:samp
20     for index = 1:n
21
22         y1 = y1/sum(y1); % Make sure probabilities add up to 1.
23         cp = [0, cumsum(y1)];
24         r = rand;
25         ind = find(r>cp, 1, 'last');
26         randx = x(ind); % ...
27         % https://www.mathworks.com/matlabcentral/answers/506654
28         % -select-random-number-from-an-array-with-probabilities
29         xsamp(index)=randx;
30     end
31     An(samples)=(1/n) sum(xsamp.^2);
32
33 end
34
35 % simulated pdf of Y
36 [y3]=hist(An,y)/(samp*(sz/2));
37 bar(y,y3)
38 %%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
```