

1.

a.

$$P = Zk^{-\alpha}$$

$$\int_{k_{\min}}^{\infty} P dk = Z \int_{k_{\min}}^{\infty} k^{-\alpha} dk = \frac{Z}{1-\alpha} [k^{1-\alpha}]_{k_{\min}}^{\infty} = 1, \text{ if } \alpha \neq 1$$

if  $\alpha > 1$ ,  $1 - \alpha$  is negative  $\therefore \infty^{1-\alpha} = 0$

$$\frac{Z}{1-\alpha} (-k_{\min}^{1-\alpha}) = 1$$

$$Z = (\alpha - 1) k_{\min}^{\alpha-1}$$

$$P(k) = (\alpha - 1) k_{\min}^{\alpha-1} k^{-\alpha}$$

b.

$$E(k) = \int_{k_{\min}}^{\infty} k P(k) dk$$

$$E(k) = \int_{k_{\min}}^{\infty} Z k^{1-\alpha} dk = \frac{Z}{2-\alpha} [k^{2-\alpha}]_{k_{\min}}^{\infty}, \text{ if } \alpha \neq 2$$

if  $\alpha > 2$ ,  $E(k) = \frac{Z}{2-\alpha} (-k_{\min}^{2-\alpha}) = \frac{Z}{\alpha-2} k_{\min}^{2-\alpha}$

if  $\alpha \leq 2$ ,  $E(k) = \infty$

c. The  $k$ th moment of the power law distribution is

$$= \int_{x_{\min}}^{\infty} x^k p(x) dx$$

$$= (\alpha - 1) / x_{\min}^{\alpha-1} \int_{x_{\min}}^{\infty} x^{-\alpha+k} dx$$

$$= x_{\min}^k \left( \frac{\alpha - 1}{\alpha - 1 - k} \right)$$

According to the definition of moments, the second moment is the variance.

When  $\alpha$  is between 2 and 3, the variance is infinite. Thus we can conclude that even though the expected value is finite, it is not a good estimator of the data since the variance is so large.

d.

$$P(X \geq k) = F(k) = \int_k^{\infty} Z x^{-\alpha} dx$$

$$F(k) = \frac{Z}{1-\alpha} [x^{1-\alpha}]_k^{\infty} = \frac{Z}{\alpha-1} k^{1-\alpha} = \frac{(\alpha-1)(k_{\min}^{\alpha-1})}{\alpha-1} k^{1-\alpha}$$

$$F(k) = \frac{k^{1-\alpha}}{k_{\min}^{1-\alpha}}$$

2.

a.

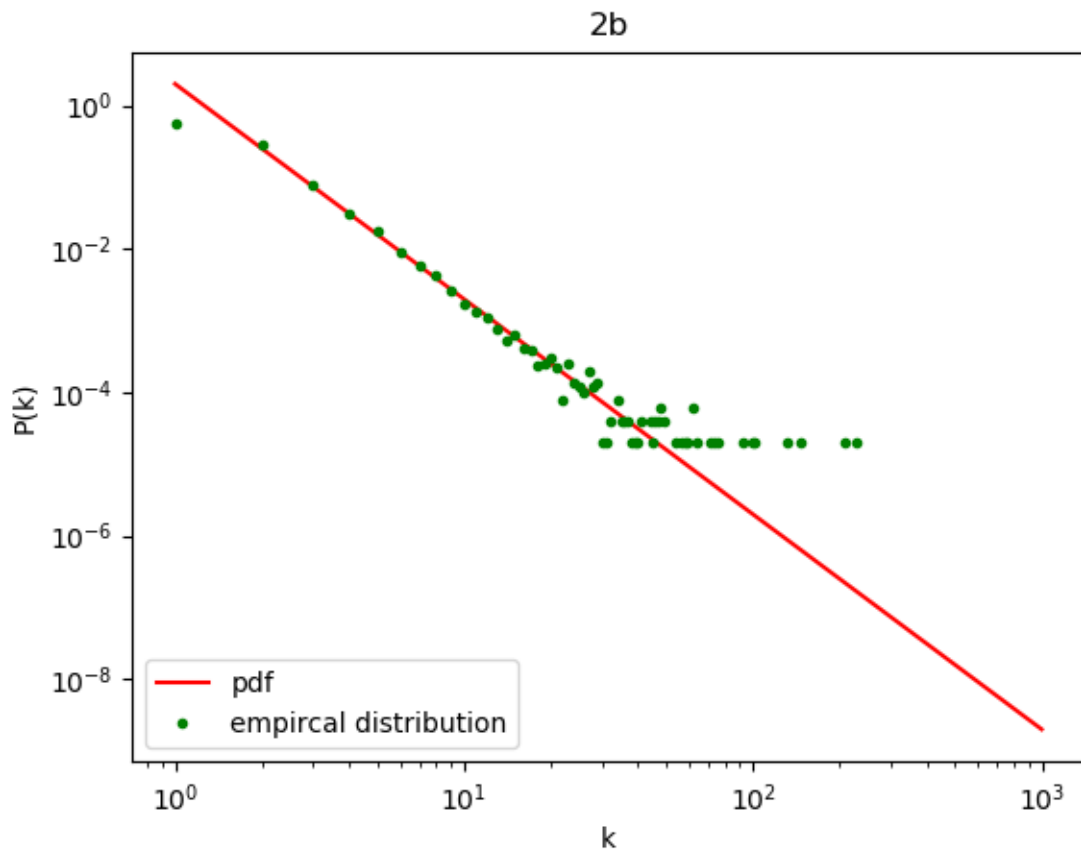
$$F(k) = u$$

$$u = \frac{k^{1-\alpha}}{k_{min}^{1-\alpha}}$$

$$u^{1/(1-\alpha)} = \frac{k}{k_{min}}$$

$$k = k_{min} u^{1/(1-\alpha)}$$

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c.

i.

$$P = Zk^{-\alpha}$$

$$\ln P = \ln Z - \alpha \ln k$$

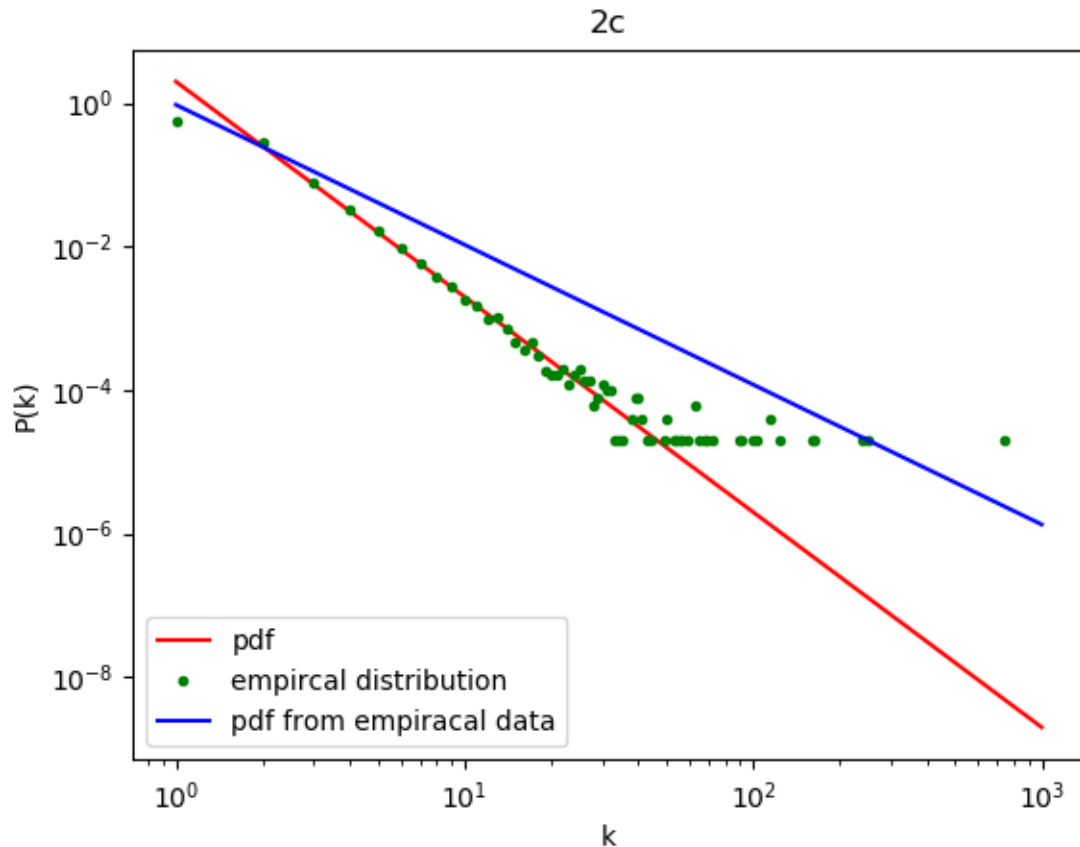
Let  $\beta$  be the result of a linear regression of  $\ln P$  on  $\ln k$

$$Z = (\alpha - 1)k_{min}^{\alpha-1} = e^{\beta_1}$$

$$\alpha = \beta_2$$

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iii.



iv. This method produces a bad fit for  $\alpha$  because there is a minimum probability due to the number of points, and this causes points that occur that have large  $k$  values to have a probability much higher than what would be anticipated

d.

i.

$$P = \frac{1}{k_{min}^{1-\alpha}} k^{1-\alpha}$$

$$\ln P = \ln \frac{1}{k_{min}^{1-\alpha}} + (1 - \alpha) \ln k$$

Let  $\beta$  be the result of a linear regression of  $\ln P$  on  $\ln k$

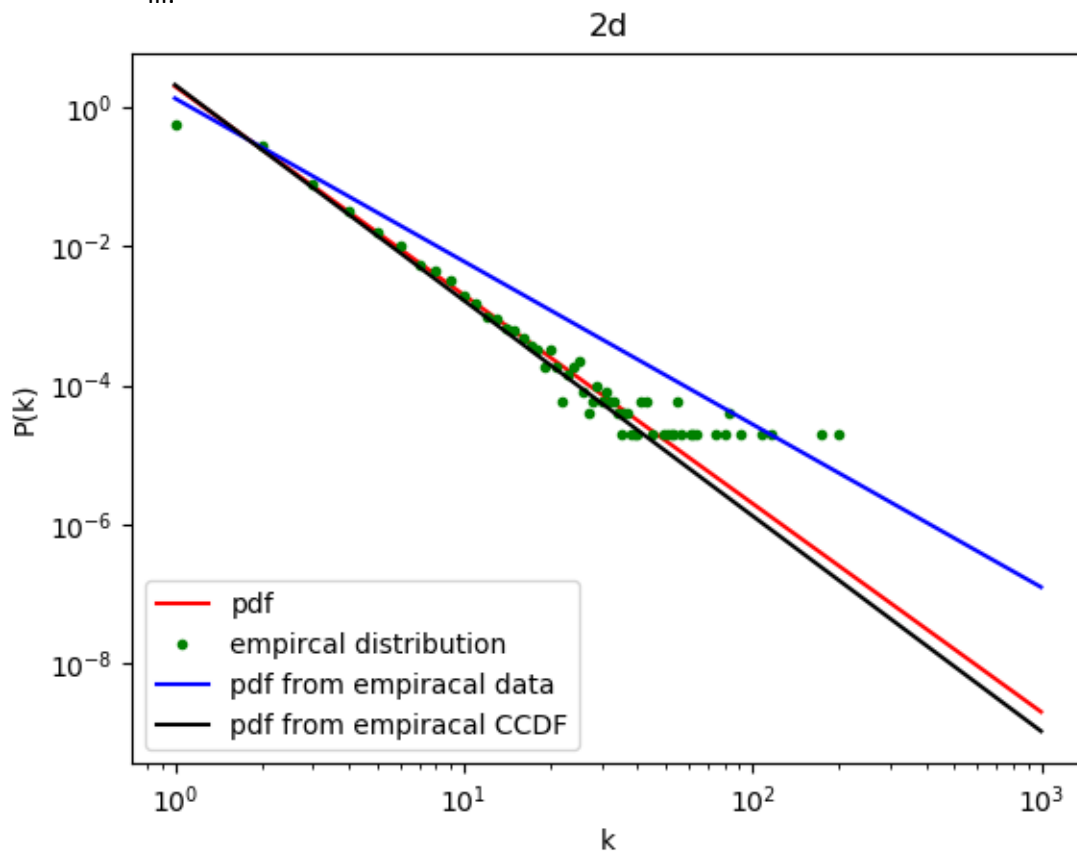
$$(\alpha - 1) \ln k_{min} = \beta_1$$

$$1 - \alpha = \beta_2$$

$$\alpha = 1 - \beta_2$$

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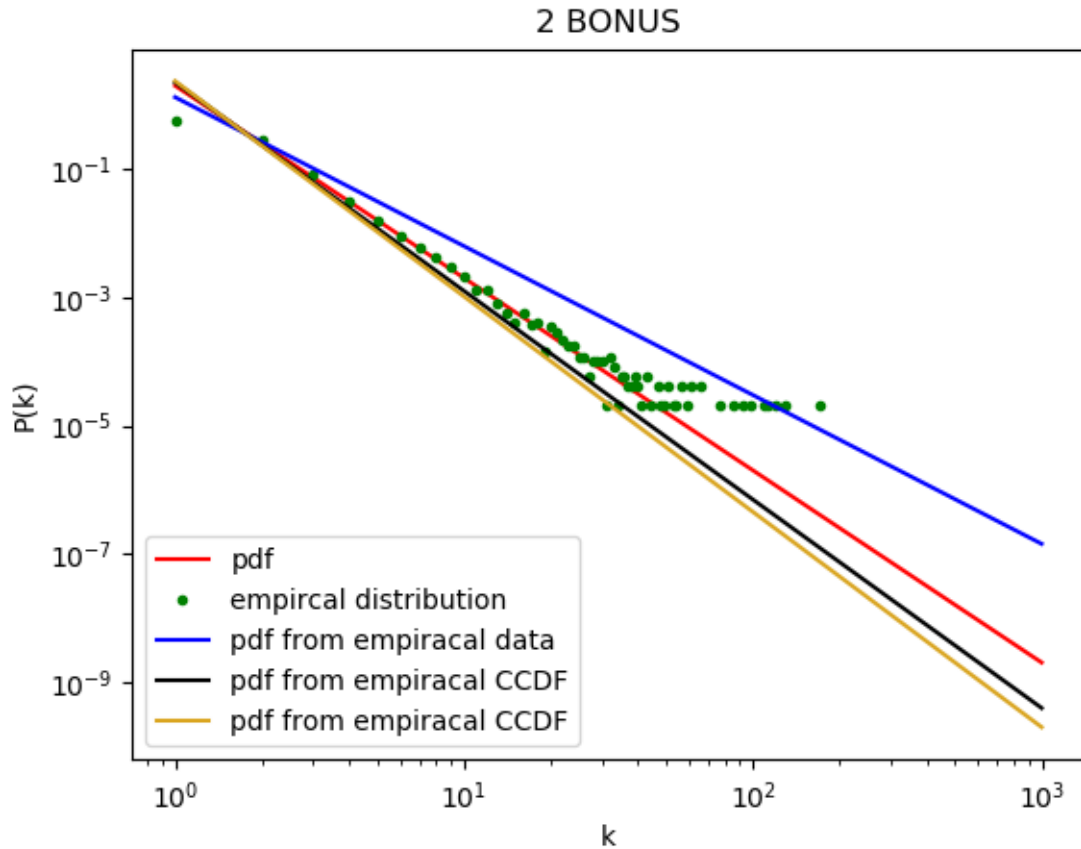
iii.



- e. using pdf, average alpha = 2.21909439331  
 using pdf, standard deviation alpha = 0.166870329052  
 using ccdf, average alpha = 2.87928313183  
 using ccdf, standard deviation alpha = 0.382480866581

please refer to our code for detailed computation of the values. It is clear that ccdf is a better estimate of  $\alpha$ . The average of pdf method is way too far away from 3.

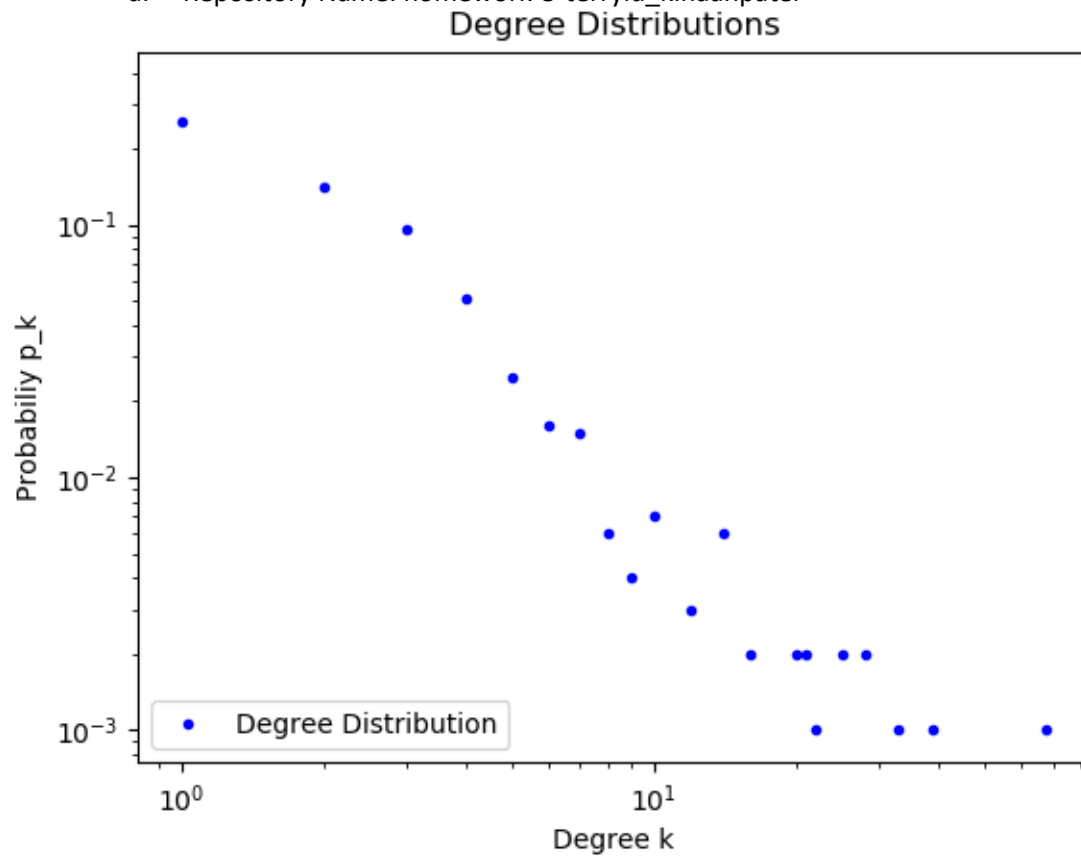
- f. BONUS:  
i. Repository Name: homework-3-terrylu\_kinaanpatel  
ii.



- iii.
- The mean estimate using MLE, 3.34178463954, is larger than the actual value 3, and is also larger than the average using ccdf method, 2.87928313183, and using pdf method, 2.21909439331. The standard deviation using MLE, 0.014035993224 is much lower than the standard deviation using Pdf, 0.166870329052, and it is also lower than the standard deviation using ccdf 0.382480866581.

3.

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b. average  $\alpha$  = 2.47375682356  
standard deviation  $\alpha$  = 0.0700854604155