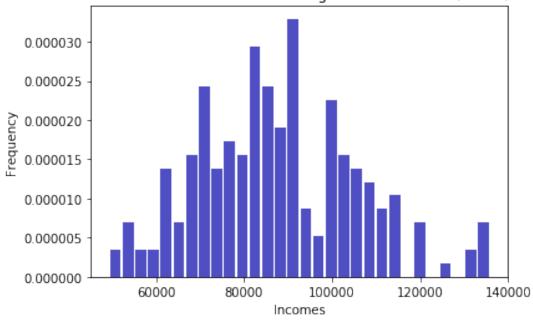
hw4

February 4, 2019

```
In [1]: import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    import scipy.stats as sts
    from scipy.integrate import quad

0.1 1
0.1.1 a
In [2]: Incomes = np.loadtxt('incomes.txt')
    fig,ax = plt.subplots()
    ax.set_xlabel('Incomes')
    ax.set_ylabel('Frequency')
    ax.set_title('Annual incomes of MACSS students graduated in 2018, 2019, 2020')
    ax.hist(x=Incomes, bins=30, color='#0504aa', alpha=0.7, rwidth=0.85, normed = True)
    plt.show()
```

Annual incomes of MACSS students graduated in 2018, 2019, 2020



```
0.1.2 b
```

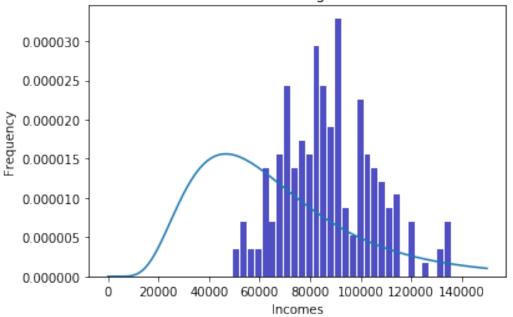
```
In [3]: def lognorm(x, mu=11, sigma=0.5):
            return 1/(x*sigma * np.sqrt(2 * np.pi))*np.e**(-(np.log(x) - mu)**2 / (2 * sigma**)
        def log_lik_truncnorm(x, mu=11, sigma=0.5, cut_lb=0, cut_ub=150000):
            pdf_vals = lognorm(x, mu, sigma)
            if cut_lb!=None and cut_ub!=None:
                total = quad(lambda x: lognorm(x, mu,sigma), cut_lb, cut_ub)[0]
            elif cut_lb!=None and cut_ub==None:
                total = 1-quad(lambda x: lognorm(x, mu,sigma), cut_lb, np.inf)[0]
            elif cut_lb==None and cut_ub!=None:
                total = quad(lambda x: lognorm(x, mu,sigma), -np.inf, cut_ub)[0]
            else:
                total = 1
            polishedpdf = pdf_vals/total
            ln_pdf_vals = np.log(polishedpdf)
            log_lik_val = ln_pdf_vals.sum()
            return log_lik_val
        total = quad(lognorm, 0, 150000)[0]
        X = np.linspace(0, 150000, 150001)
        fig,ax = plt.subplots()
        ax.set_xlabel('Incomes')
        ax.set_ylabel('Frequency')
        ax.set_title('Annual incomes of MACSS students graduated in 2018, 2019, 2020')
        ax.plot(X, lognorm(X)/total)
        ax.hist(x=Incomes, bins=30, color='#0504aa', alpha=0.7, rwidth=0.85, normed = True)
        plt.show()
        LLV = log_lik_truncnorm(Incomes)
        print('The value of the log likelihood value is', LLV)
```

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C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:2: RuntimeWarning: divide by

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:2: RuntimeWarning: invalid va

Annual incomes of MACSS students graduated in 2018, 2019, 2020



The value of the log likelihood value is -2379.120591931827

0.1.3 c

```
In [4]: def crit(params, *args):
           mu, sigma = params
            x, cut_lb, cut_ub = args
            log_lik_val = log_lik_truncnorm(x, mu, sigma, cut_lb, cut_ub)
           neg_log_lik_val = -log_lik_val
            return neg_log_lik_val
In [5]: import scipy.optimize as opt
       mu_init = 11
        sig_init = 0.5
       params_init = np.array([mu_init, sig_init])
       mle_args = (Incomes, 0, 150000)
       results = opt.minimize(crit, params_init, args=(mle_args))
       mu_MLE, sig_MLE = results.x
       LLV_MLE = log_lik_truncnorm(Incomes, mu_MLE, sig_MLE)
       VCV_MLE = results.hess_inv
       print('mu_MLE=', mu_MLE, ' sig_MLE=', sig_MLE)
       print('The value of the log likelihood value is', LLV_MLE)
        print('The variance-covariance matrix is', VCV_MLE)
```

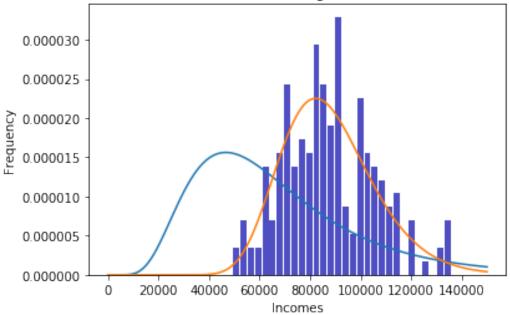
```
mu_MLE= 11.361699971261004 sig_MLE= 0.2117432623314924
The value of the log likelihood value is -2240.9343375116364
The variance-covariance matrix is [[8.67691939e-05 3.92325340e-05]
[3.92325340e-05 1.16938120e-04]]
```

```
In [6]: total1 = quad(lognorm, 0, 150000)[0]
    total2 = quad(lambda x: lognorm(x, mu=mu_MLE,sigma=sig_MLE), 0, 150000)[0]
    X = np.linspace(0,150000, 150001)
    fig,ax = plt.subplots()
    ax.set_xlabel('Incomes')
    ax.set_ylabel('Frequency')
    ax.set_title('Annual incomes of MACSS students graduated in 2018, 2019, 2020')
    ax.plot(X, lognorm(X)/total1, label = 'initial guess')
    ax.plot(X, lognorm(X, mu=mu_MLE,sigma=sig_MLE)/total2, label = 'MLE result')
    ax.hist(x=Incomes, bins=30, color='#0504aa', alpha=0.7, rwidth=0.85, normed = True)
    plt.show()
```

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```
0.1.4 d
```

```
In [7]: mu_new, sig_new = np.array([11, 0.5])
        log_lik_h0 = log_lik_truncnorm(Incomes, mu_new, sig_new, 0, 150000)
        print('hypothesis value log likelihood', log_lik_h0)
        log_lik_mle = log_lik_truncnorm(Incomes, mu_MLE, sig_MLE, 0, 150000)
        print('MLE log likelihood', log_lik_mle)
       LR_val = 2 * (log_lik_mle - log_lik_h0)
        print('likelihood ratio value', LR_val)
        pval_h0 = 1.0 - sts.chi2.cdf(LR_val, 2)
       print('chi squared of HO with 2 degrees of freedom p-value = ', pval hO)
hypothesis value log likelihood -2379.120591931827
MLE log likelihood -2240.9343375116364
likelihood ratio value 276.37250884038167
chi squared of HO with 2 degrees of freedom p-value = 0.0
0.1.5 e
In [8]: print("The probability that you will earn more than $100,000 is",
              quad(lambda x: lognorm(x, mu=mu MLE, sigma=sig MLE), 100000, 150000)[0]/total2)
       print("The probability that you will earn less than $75,000 is",
              quad(lambda x: lognorm(x, mu=mu MLE, sigma=sig MLE), 0, 75000)[0]/total2)
The probability that you will earn more than $100,000 is 0.23427612236731754
The probability that you will earn less than $75,000 is 0.26076018268903833
0.2 2
0.2.1 a
In [9]: def norm_pdf(x, sigma):
            sigma = np.abs(sigma)
            pdf_vals = (1/(sigma * np.sqrt(2 * np.pi)) *
                            np.exp(-x**2 / (2 * sigma**2)))
            return pdf_vals
        def log_lik_norm(x, beta0, beta1, beta2, beta3, sigma):
            error = x[:,0] - beta0 - beta1*x[:,1] - beta2*x[:,2] - beta3*x[:,3]
            pdf_vals = norm_pdf(error, sigma)
            ln_pdf_vals = np.log(pdf_vals)
            log_lik_val = ln_pdf_vals.sum()
            return log_lik_val
        def new_crit(params, *args):
           beta0, beta1, beta2, beta3, sigma = params
            x = args[0]
            log_lik_val = log_lik_norm(x,beta0, beta1, beta2, beta3, sigma)
            neg_log_lik_val = -log_lik_val
```

```
Data = []
        with open('sick.txt') as f:
            for line in f:
                line = line.strip()
                data = line.split(',')
                Data.append(data)
        Data.pop(0)
        for i in range(len(Data)):
            Data[i] = [float(x) for x in Data[i]]
        Data = np.array(Data)
        new_Data = np.ones_like(Data)
        new_Data[:,1:] = Data[:,1:]
        beta0_init, beta1_init, beta2_init, beta3_init = np.linalg.lstsq(new_Data,Data[:,0])[0]
        sigma_init = 0.01
        parameters_init = np.array([beta0_init, beta1_init, beta2_init, beta3_init, sigma_init,
        results2 = opt.minimize(new_crit, parameters_init, args=(Data))
        beta0_MLE, beta1_MLE, beta2_MLE, beta3_MLE, sigma_MLE = results2.x
        new_LLV_MLE = log_lik_norm(Data, beta0_MLE, beta1_MLE, beta2_MLE, beta3_MLE, sigma_MLE
        new_VCV_MLE = results2.hess_inv
        print('beta0_MLE =',beta0_MLE,
              'beta1_MLE =',beta1_MLE,
              'beta2_MLE =',beta2_MLE,
              'beta3_MLE =',beta3_MLE,
              'sigma_MLE =',sigma_MLE)
        print('The value of the log likelihood value is', new_LLV_MLE)
        print('The variance-covariance matrix is', new_VCV_MLE)
        Error = Data[:,0] - beta0_MLE - beta1_MLE*Data[:,1] - beta2_MLE*Data[:,2] - beta3_MLE*
        X2 = np.linspace(-0.01, 0.01, 10001)
        fig,ax = plt.subplots()
        ax.set_xlabel('Errors')
        ax.set_ylabel('Frequency')
        ax.set_title('Estimate Errors')
        ax.plot(X2, norm_pdf(X2,sigma_MLE))
        ax.hist(x=Error, bins='auto', color='#0504aa', alpha=0.7, rwidth=0.85, normed = True)
        plt.show()
beta0_MLE = 0.2516415006612955 beta1_MLE = 0.012928701541639099 beta2_MLE = 0.4004996036681293
The value of the log likelihood value is 871.487646107643
The variance-covariance matrix is [[ 9.99749814e-01 -1.07472308e-02 -4.38873773e-04 -1.1594833
  -1.55722702e-04]
 [-1.07472308e-02 5.38385787e-01 -1.88499775e-02 -4.98050664e-01
   1.13266069e-02]
 [-4.38873773e-04 -1.88499775e-02 9.99230140e-01 -2.03364503e-02
   1.57898527e-03]
 [-1.15948335e-02 -4.98050664e-01 -2.03364503e-02 4.62631712e-01
  -1.05543181e-02]
 [-1.55722702e-04 1.13266069e-02 1.57898527e-03 -1.05543181e-02
```

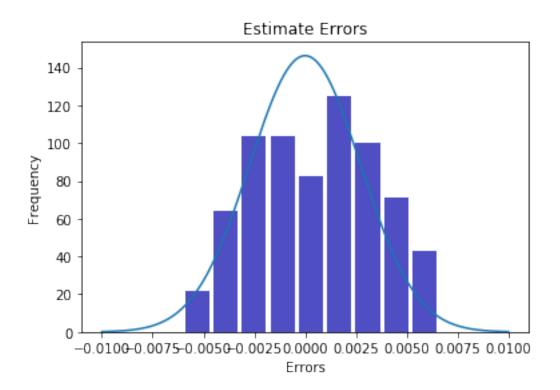
return neg_log_lik_val

2.42229668e-04]]

```
C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:30: FutureWarning: `rcond` packages\ipykernel_launcher.py:30: FutureWarning: `rcond` packages\ipykernel_launcher.py:9: RuntimeWarning: divide by if __name__ == '__main__':
C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:9: RuntimeWarning: divide by if __name__ == '__main__':
```

C:\ProgramData\Anaconda3\lib\site-packages\scipy\optimize\optimize.py:643: RuntimeWarning: invegrad[k] = (f(*((xk + d,) + args)) - f0) / d[k]

C:\ProgramData\Anaconda3\lib\site-packages\ipykernel_launcher.py:9: RuntimeWarning: divide by :
 if __name__ == '__main__':



0.2.2 b

hypothesis value log likelihood -2253.700688042125
MLE log likelihood 871.487646107643
likelihood ratio value 6250.376668299536
chi squared of HO with 5 degrees of freedom p-value = 0.0

The results show that it's not likely that age, number of children, and average winter temperature have no effect on the number of sick days.