

## 625.661 - Homework Seven

Eric Niblock

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1. **The data included represents the fraction of active chlorine in a chemical product as a function of time after manufacturing.**

- (a) **Construct a scatterplot of the data.**

Results for this problem are provided in the attached PDF. The scatter plot was constructed and includes both the data as well as the fitted regression.

- (b) **Fit the Mitcherlich law (see Problem 12.10) to these data. Discuss how you obtained the starting values.**

Results for this problem are provided in the attached PDF. The coefficients were determined to have the following values:  $\theta_1 = 0.3829$ ,  $\theta_2 = -0.1986$ ,  $\theta_3 = 0.0802$ . Thus, the model can be represented by the following,

$$y = 0.3829 + 0.1986e^{-0.0802x} \quad (1)$$

The starting values for the fitting function were obtained by analyzing the scatter plot and expectation function.

- (c) **Test for significance of regression.**

Results for this problem are provided in the attached PDF. An  $F$ -value of 120.635 was calculated, which has a corresponding  $p$ -value of

approximately zero. Therefore, the null hypothesis can be rejected, and we note that the regression is significant.

- (d) **Find approximate 95% confidence intervals on the parameters  $\theta_1$ ,  $\theta_2$ , and  $\theta_3$ . Is there evidence to support the claim that all three parameters are different from zero?**

Results for this problem are provided in the attached PDF. Observe the 95% confidence intervals constructed for each of the parameters,  $\theta_1$ ,  $\theta_2$ , and  $\theta_3$ . There is evidence to support that all three parameters differ from zero, since none of the confidence intervals contain zero.

- (e) **Analyze the residuals and comment on model adequacy.**

Results for this problem are provided in the attached PDF. A normal probability plot of the residuals was constructed. Though the residuals tend to oscillate around the normal line, there does appear to be a distinct pattern around this oscillation. This calls the normality assumption into question, and the residuals may not be entirely normal. Furthermore, the residuals were plotted as a function of  $y$ , and the magnitude of the residuals does appear to be correlated with the value of  $y$ . Therefore, the data does display some heteroscedasticity.

## Data Cleaning and Selection

```
In [141]: import numpy as np
import matplotlib.pyplot as plt
```

```
In [142]: import matplotlib.pyplot as plt
from scipy.optimize import curve_fit
```

```
In [143]: data = [[0.49, 0.49, 8] ,
[0.48, 0.47, 0.48, 0.47, 10 ] ,
[0.46, 0.46, 0.45, 0.43, 12 ] ,
[0.45, 0.43, 0.43, 14 ] ,
[0.44, 0.43, 0.43, 16 ] ,
[0.46, 0.45, 18 ] ,
[0.42, 0.42, 0.43, 20 ] ,
[0.41, 0.41, 0.40 , 22 ] ,
[0.42, 0.40, 0.40, 24 ] ,
[0.41, 0.40, 0.41, 26 ] ,
[0.41, 0.40, 28 ] ,
[0.40, 0.40, 0.38, 30 ] ,
[0.41, 0.40, 32 ] ,
[0.40, 34 ] ,
[0.41, 0.38, 36 ] ,
[0.40, 0.40, 38 ] ,
[0.39, 40 ] ,
[0.39, 42]]

choices = np.random.choice(len(data),14,replace=False)
print('Sample rows from provided table:')
sample = [data[e] for e in choices]
sample
```

Sample rows from provided table:

```
Out[143]: [[0.41, 0.38, 36],
[0.41, 0.4, 28],
[0.4, 0.4, 0.38, 30],
[0.48, 0.47, 0.48, 0.47, 10],
[0.39, 40],
[0.49, 0.49, 8],
[0.42, 0.42, 0.43, 20],
[0.41, 0.4, 32],
[0.41, 0.4, 0.41, 26],
[0.46, 0.45, 18],
[0.4, 34],
[0.44, 0.43, 0.43, 16],
[0.42, 0.4, 0.4, 24],
[0.46, 0.46, 0.45, 0.43, 12]]
```

```
In [144]: fulldata = []
for l in sample:
    for i in l[:-1]:
        fulldata.append([i,l[-1]])
fulldata = np.array(fulldata)
print('Data organized into points:')
fulldata
```

Data organized into points:

```
Out[144]: array([[ 0.41, 36. ],
 [ 0.38, 36. ],
 [ 0.41, 28. ],
 [ 0.4 , 28. ],
 [ 0.4 , 30. ],
 [ 0.4 , 30. ],
 [ 0.38, 30. ],
 [ 0.48, 10. ],
 [ 0.47, 10. ],
 [ 0.48, 10. ],
 [ 0.47, 10. ],
 [ 0.39, 40. ],
 [ 0.49,  8. ],
 [ 0.49,  8. ],
 [ 0.42, 20. ],
 [ 0.42, 20. ],
 [ 0.43, 20. ],
 [ 0.41, 32. ],
 [ 0.4 , 32. ],
 [ 0.41, 26. ],
 [ 0.4 , 26. ],
 [ 0.41, 26. ],
 [ 0.46, 18. ],
 [ 0.45, 18. ],
 [ 0.4 , 34. ],
 [ 0.44, 16. ],
 [ 0.43, 16. ],
 [ 0.43, 16. ],
 [ 0.42, 24. ],
 [ 0.4 , 24. ],
 [ 0.4 , 24. ],
 [ 0.46, 12. ],
 [ 0.46, 12. ],
 [ 0.45, 12. ],
 [ 0.43, 12. ]])
```

## Parts (a) (b)

```
In [145]: def func(x, b1, b2, b3):
return b1 - b2*np.exp(-1*b3*x)
```

```
In [146]: popt, pcov = curve_fit(func, fulldata[:,1], fulldata[:,0], p0=[0,-0.1,0])
```

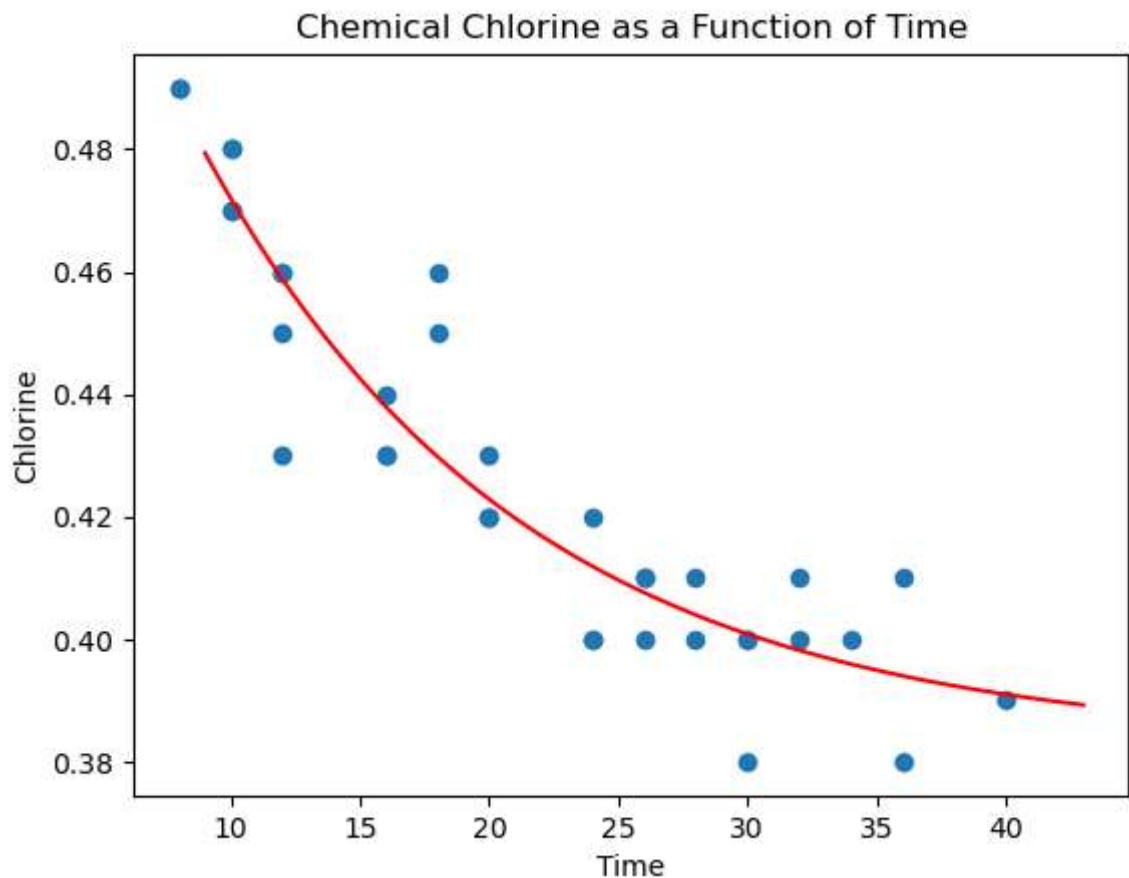
```
In [147]: print('Fitted coefficients: ')
popt
```

Fitted coefficients:

```
Out[147]: array([ 0.38295219, -0.19862392,  0.08025721])
```

```
In [148]: xs = np.linspace(9,43,100)
ys = [popt[0] - popt[1]*np.exp(-1*popt[2]*x) for x in xs]
```

```
In [149]: plt.scatter(fulldata[:,1], fulldata[:,0])
plt.plot(xs,ys, c='r')
plt.title('Chemical Chlorine as a Function of Time')
plt.xlabel('Time')
plt.ylabel('Chlorine')
plt.show()
```



## Part (c)

```
In [157]: ypred = [popt[0] - popt[1]*np.exp(-1*popt[2]*x) for x in fulldata[:,1]]
```

```
In [158]: ss_tot = sum((fulldata[:,0] - np.mean(fulldata[:,0]))**2)
ss_res = sum((fulldata[:,0] - ypred)**2)
ms_res = ss_res/(len(fulldata)-3)
```

```
In [163]: F = (ss_tot-ss_res)/(2*ms_res)
print('F-statistic: ', F)
```

```
F-statistic:      120.63526928616666
```

## Part (d)

```
In [170]: print('Confidence Interval for B_1: ')
print(popt[0]-(2.037*pcov[0,0]**0.5), ',', popt[0]+(2.037*pcov[0,0]**0.5))
```

```
Confidence Interval for B_1:
0.3618467068905411 , 0.4040576757902321
```

```
In [171]: print('Confidence Interval for B_2: ')
print(popt[1]-(2.037*pcov[1,1]**0.5), ',', popt[1]+(2.037*pcov[1,1]**0.5))
```

```
Confidence Interval for B_2:
-0.24806890917051916 , -0.149178930091885
```

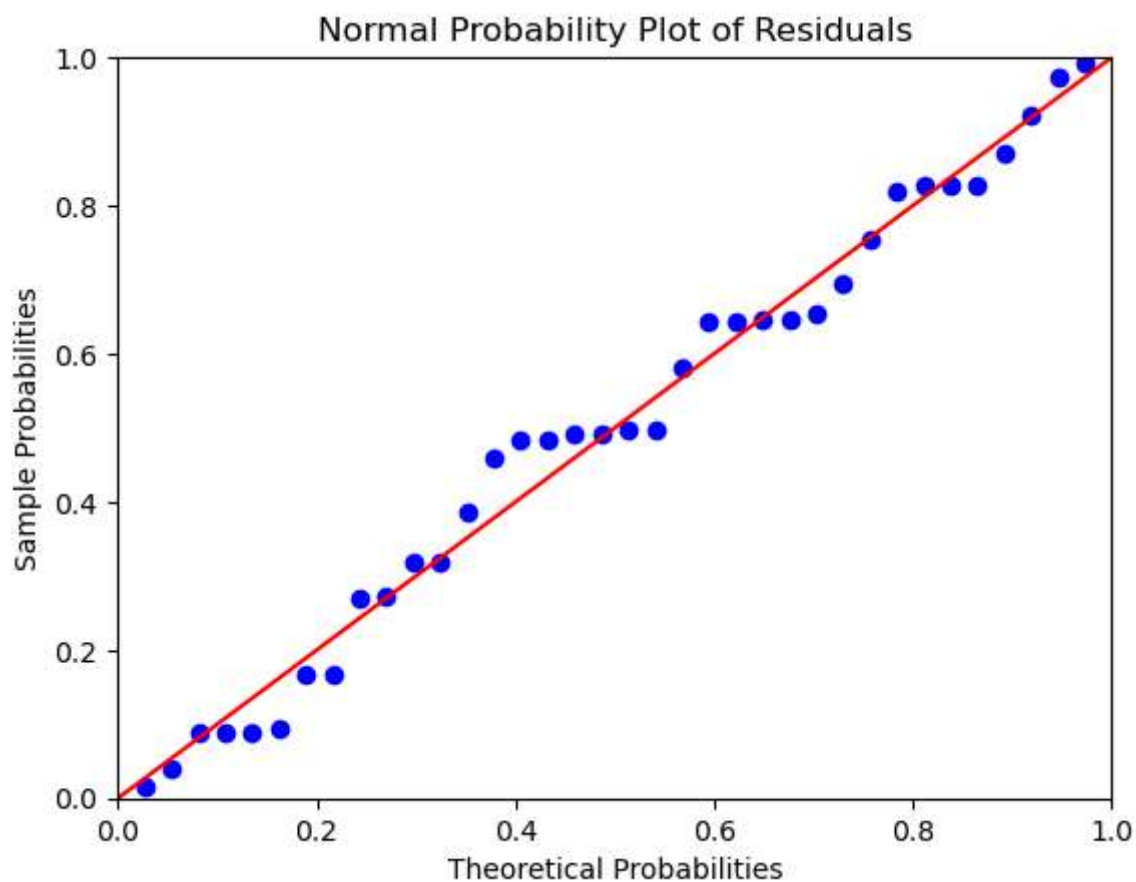
```
In [172]: print('Confidence Interval for B_3: ')
print(popt[2]-(2.037*pcov[2,2]**0.5), ',', popt[2]+(2.037*pcov[2,2]**0.5))
```

```
Confidence Interval for B_3:
0.03893443450520916 , 0.12157999019601332
```

## Part (e)

```
In [128]: res = fulldata[:,0] - ypred
```

```
In [132]: import statsmodels.api as sm
import scipy.stats as stats
pplot = sm.ProbPlot(res, stats.t, fit=True)
fig = pplot.ppplot(line="45")
h = plt.title("Normal Probability Plot of Residuals")
plt.show()
```



```
In [135]: plt.scatter(fulldata[:,0],res, c='b')
plt.plot([0.375,0.485],[0,0],c='r')
plt.xlim(0.375,0.485)
plt.xlabel('y')
plt.ylabel('Residual')
h = plt.title("Residuals as a Function of y")
plt.show()
```

