

## Applied Statistic – Final Project Part-1

Firstly, with samples in excel :

LASTGEL...  $\text{fx} = (E46 - (((((273 + E37) * E21 * E46) / (E33 - 2))^{(1/2)}) * 0,0238)) * (((273 + E37) * E21 * 0,0239) / ((E33 - 2) * E30)))$

Lab Temperature / Lab Sıcaklığı [°C]	21			Person in Charge
Sample No / Numune No	1	2	3	
Thickness of Sample-1 / Numunenin Kalınlığı-1, [mm]	51,71	50,41	52,42	
Thickness of Sample-2 / Numunenin Kalınlığı-2, [mm]	51,41	50,59	52,72	
Average of Thickness	51,41	50,50	52,57	
Start Time of Sample in Vacuum (50 mmHg, 3h) / Numunenin vakuma konulduğu saat (50 mmHg), [dd.mm.yyyy. hh:mm]				
Start Time of Sample Immersed in Solution (1h) / Numunenin çözeltiye konulduğu saat, [dd.mm.yyyy. hh:mm]				
Time of Sample out of Bucket (After 18±2 h) / Numunenin desikatörden çıkarıldığı saat, [dd.mm.yyyy. hh:mm]				
NaOH Temperature at Start / Başlangıçtaki NaOH Sıcaklığı, [°C]	23,5	23,5	23,5	51,4 50,5 52,6 0,0
Initial Voltage / İlk Voltaj, [V]	30	30	30	
Initial Current / İlk Akım, [mA]	43,50	42,80	40,10	
Starting Voltage / Başlangıçtaki Voltaj, [V]	30	30	30	
Starting Current / Başlangıçtaki Akım, [mA]	43,50	42,80	40,10	
Test Duration / Test Süresi, [h]	24	24	24	
Starting Time When New Voltage Is Applied / Yeni Voltaj Verildiği Andaki Başlangıç Saati, [dd.mm.yyyy. hh:mm]	8.08.2017	8.08.2017	8.08.2017	
Specimen Age, [days]	28	28	28	
Ending Voltage / Bitiş Voltajı, [V]	30	30	30	
Ending Current / Bitiş Akımı, [mA]	48,80	45,70	40,80	
Ending Time / Bitiş Zamanı, [dd.mm.yyyy. hh:mm]	9.08.2017	9.08.2017	9.08.2017	
Ending Temperature of NaOH / NaOH Son Sıcaklığı, [°C]	24,3	24,3	23,8	
Average of NaOH Temperature	23,9	23,9	23,65	
Chloride Ingress-1 / Klor Derinliği-1, [mm]	13,41	11,02	13,19	
Chloride Ingress-2 / Klor Derinliği-2, [mm]	12,21	9,72	11,35	
Chloride Ingress-3 / Klor Derinliği-3, [mm]	13,08	10,85	12,71	
Chloride Ingress-4 / Klor Derinliği-4, [mm]	13,23	11,71	13,85	
Chloride Ingress-5 / Klor Derinliği-5, [mm]	14,86	10,82	10,23	
Chloride Ingress-6 / Klor Derinliği-6, [mm]	23,55	15,20	17,14	
Chloride Ingress-7 / Klor Derinliği-7, [mm]	16,74	10,91	9,85	
Max. Depth of Chloride Ingress / Max. Klor Derinliği, [mm]	24,19	16,05	17,45	
Average Chloride Ingress, [mm] Xd	15,29	11,46	12,62	
Mass in Air / Havadaki Ağırlık, [g]	970	954	973	
Mass in Water / Sudaki Ağırlık, [g]	599	559	567	
Saturated Unit Weight / Doymuş Birim Ağırlık [g/cm³]	2419	2415	2397	
Diameter / Çap, [mm]	100	100	100	
Dnssm (non-steady-state migration coefficient, m²/s; $2 * E30$ )	5,1175413	5,89139413		

Independent variables in the formula forming the migration coefficient were found.

### R-DATA PREPARATION :

(When running rds → The path specified in the file must be changed)

#### ## Excel Operations

```
``{r pressure, echo=FALSE}
setwd("//Users//macbookair//Downloads//CM-RawData")
excel_data <- read_excel("YCMR-XXX 14.08.2017 G3HL Deneme 2.xlsx", sheet="Record_CMC")
```

Then, using R programming language :

- Received data on Record\_CMC sheet.
- Found rows with independent variables.(Renamed column names etc.)
- Respectively, merged data from 34 excel locations.
- $34 * 3 = 102$  data sample merging. (34 excel and 3 sample on each excel.)

```

excel_data <- read_excel("YCMR-XXX 14.08.2017 G3HL Deneme 2.xlsx",sheet="Record_CMC")
## Kullandığımız datayı almak için ölçüm yapılan yerler alınır
excel_data_1 <- excel_data %>% slice(17:51)
## Ölçüm sonucu ilgili kolonlar seçilerek yeni df ye aktarılır
dfNew <- excel_data_1[,c(4:6)]

dfNew_1_factor <- dfNew[,3,]
dfNew_2_factor <- dfNew[,12,]
dfNew_3_factor <- dfNew[,15,]
dfNew_4_factor <- dfNew[,19,]
dfNew_5_factor <- dfNew[,28,]
dfNew_ME_factor <- dfNew[,33,]

```

(Same process for every excel)

After data merging:

	y	x1	x2	x3	x4	x5
1	7.1229118394783226	51.41	24	30	23.9	15.294285714285712
2	5.1175412878446522	50.5	24	30	23.9	11.461428571428572
3	5.8913941287808891	52.57	24	30	23.65	23.65
4	3.5003630201665183	51.384999999999998	24	40	24.3	10.272857142857143
5	3.746300075618004	51.81	24	40	24.3	10.860000000000001
6	3.2008295168693022	52.125	24	40	24.3	9.3514285714285705
7	3.6343439856487789	51.31	24	50	24.9	12.991428571428571
8	2.673882714516469	51.17	24	50	24.9	9.7814285714285738
9	3.2910779969053054	52.144999999999996	24	50	24.75	11.680000000000001
10	3.778446584754291	51.034999999999997	24	50	24.950000000000003	13.535714285714283
11	2.4894839696572251	51.015000000000001	24	50	24.8	9.1814285714285724
12	2.1740865782447463	52.515000000000001	24	50	24.75	7.9099999999999993
13	2.8704098347666958	51.010000000000005	24	40	24.65	8.6100000000000012
14	3.3024425591158799	51.16	24	40	24.65	9.7642857142857142
15	3.6871085985161107	52.215000000000003	24	40	24.65	10.62142857142857
16	3.3174682847411674	50.53	24	40	24.6	9.9085714285714293
17	3.3159906045894947	51.489999999999995	24	40	24.65	9.7485714285714273
18	3.2434503791638019	51.36	24	40	24.55	9.5757142857142856
19	4.479989880319124	51.94	24	50	23.9	15.692857142857141
20	6.1380250675144845	51.480000000000004	24	50	24.35	21.288571428571426
21	5.1707634657610999	52.064999999999998	24	50	24.2	17.915714285714287
22	5.4048585381434098	51.445	24	40	23.7	15.382857142857144
23	4.7929995677438066	50.89	24	40	23.75	13.87857142857143
24	5.7755259296049903	50.989999999999995	24	40	23.8	16.491428571428571
25	4.7686340414800199	51.204999999999998	24	40	23.7	13.741428571428571
26	2.6261397217291531	51.314999999999998	24	40	23.7	7.9228571428571444
27	3.5942550332639183	50.989999999999995	24	40	23.65	10.614285714285714
28	7.1094067071323703	51.284999999999997	24	40	23.4	19.974285714285713
29	5.7538448994087865	51.884999999999998	24	40	23.4	16.204285714285714
30	3.3860543682374131	51.78	24	40	23.4	9.9257142857142835
31	2.6984340507556857	51.474999999999994	24	50	23.35	9.8585714285714285
32	4.3328221529192419	51.725000000000001	24	50	23.450000000000003	15.282857142857141
33	2.588847341172011	51.43	24	50	23.55	9.4885714285714293
34	4.0785771013760339	50.93	24	40	23.1	11.959999999999999
35	3.742321400010501	50.754000000000005	24	40	23.700000000000003	11.074285714285717

(34\*3 = 102 data)

Y -> Migration Coefficient(dnssm)

X1 -> Average Of Thickness

X2 -> Test Duration

X3 -> Ending Voltage

X4 -> Average of NaOH temperature

X5 -> Average Chloride Ingress



a correlation test was performed for each factor :

```
res_x4 <- cor.test(migration_data$y, migration_data$x4,  
  method = "pearson")
```

Pearson's product-moment correlation

```
data: migration_data$y and migration_data$x4  
t = -1.3594, df = 100, p-value = 0.1771  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 -0.32077428 0.06138036  
sample estimates:  
 cor  
-0.1347025
```

Pearson's product-moment correlation

```
data: migration_data$y and migration_data$x1  
t = 1.7185, df = 100, p-value = 0.0888  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 -0.02596046 0.35224223  
sample estimates:  
 cor  
0.1693695
```

the standard deviation is zero

Pearson's product-moment correlation

```
data: migration_data$y and migration_data$x2  
t = NA, df = 100, p-value = NA  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 NA NA  
sample estimates:  
 cor  
 NA
```

Pearson's product-moment correlation

```
data: migration_data$y and migration_data$x3  
t = -3.7927, df = 100, p-value = 0.0002552  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 -0.5136722 -0.1720107  
sample estimates:  
 cor  
-0.3546228
```

Pearson's product-moment correlation

```
data: migration_data$y and migration_data$x5  
t = 22.328, df = 100, p-value < 2.2e-16  
alternative hypothesis: true correlation is not equal to 0  
95 percent confidence interval:  
 0.8731433 0.9402412  
sample estimates:  
 cor  
0.9126467
```

After the correlation testing :

P-value was compared with the significant value(0.05) :

```
res_x4 <- cor.test(migration_data$y, migration_data$x4,
                  method = "pearson")
res_x4#p-value = 0.1771 p_Value larger than 0.05
#Correlation coefficient -> -0.1347025

res_x1 <- cor.test(migration_data$y, migration_data$x1,
                  method = "pearson")
res_x1 #p-value = p-value = 0.0888 p_Value larger than 0.05.
#Correlation coefficient -> 0.1693695
|
res_x2 <- cor.test(migration_data$y, migration_data$x2,
                  method = "pearson")
res_x2#p-value = NA

res_x3 <- cor.test(migration_data$y, migration_data$x3,
                  method = "pearson")
res_x3#p-value = 0.0002552 p_Value less than 0.05
# We can conclude that y and x3 are significantly correlated with a correlation coefficient of -0.3546228 and p-value of 0.0002552 .

res_x5 <- cor.test(migration_data$y, migration_data$x5,
                  method = "pearson")
res_x5#p-value = p-value < 2.2e-16 p_Value larger than 0.05
# We can conclude that y and x3 are significantly correlated with a correlation coefficient of 0.9126467 and p-value of 2.2e-16 .
```

Found :

X3 and x5 → significantly correlated

Then :

### Multiple Regression applied

First , applied for all factors(5 factors):

Call:

```
lm(formula = y ~ x1 + x2 + x3 + x4 + x5, data = x)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-2.86135	-0.09795	-0.01359	0.06551	1.06431

Coefficients: (1 not defined because of singularities)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.142910	3.090331	1.341	0.183
x1	0.013639	0.063968	0.213	0.832
x2	NA	NA	NA	NA
x3	-0.085218	0.006254	-13.626	<2e-16 ***
x4	-0.046446	0.058588	-0.793	0.430
x5	0.319148	0.008843	36.089	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.3643 on 97 degrees of freedom

Multiple R-squared: 0.9438, Adjusted R-squared: 0.9415

F-statistic: 407.4 on 4 and 97 DF, p-value: < 2.2e-16



(Coefficients find by --coefficients(fit) # model coefficients)

When the equation is created :

$Y <- 4.142910 + 0.013639 * 51.41 + -0.085218 * 30 + -0.046446 * 23.9 + 0.319148 * 15.29$

(values taken by "YCMR-XXX 14.08.2017 G3HL Deneme 2.xlsx")

Result : [1] 6.057265

Then, I removed some variables from the equation based on the p values resulting from the correlation test :

First – I removed Second variable :

Call:

```
lm(formula = y ~ x1 + x3 + x4 + x5, data = x)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.86135	-0.09795	-0.01359	0.06551	1.06431

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	4.142910	3.090331	1.341	0.183
x1	0.013639	0.063968	0.213	0.832
x3	-0.085218	0.006254	-13.626	<2e-16 ***
x4	-0.046446	0.058588	-0.793	0.430
x5	0.319148	0.008843	36.089	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.3643 on 97 degrees of freedom

Multiple R-squared: 0.9438, Adjusted R-squared: 0.9415

F-statistic: 407.4 on 4 and 97 DF, p-value: < 2.2e-16

But I realized that the same with the first equation. Because SD is 0.

Result : [1] 6.057265(Result is the same with first.)

Then I removed 2 and 4 factor.

Call:

```
lm(formula = y ~ x1 + x3 + x5, data = x)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.86786	-0.09820	-0.01228	0.07224	1.03648

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	3.899548	3.069211	1.271	0.207
x1	-0.002733	0.060428	-0.045	0.964
x3	-0.085691	0.006214	-13.791	<2e-16 ***
x5	0.320258	0.008715	36.747	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.3636 on 98 degrees of freedom

Multiple R-squared: 0.9435, Adjusted R-squared: 0.9417

F-statistic: 545.1 on 3 and 98 DF, p-value: < 2.2e-16

Result :  $3.899548 + (-0.002733 * 51.41) + (-0.085691 * 30) + (0.320258 * 15.29) = 6.085059$

Then I removed 1,2 and 4 factor.

Call:

```
lm(formula = y ~ x3 + x5, data = x)
```

Residuals:

Min	1Q	Median	3Q	Max
-2.87068	-0.09774	-0.01032	0.07140	1.03610

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	3.761437	0.305642	12.31	<2e-16 ***
x3	-0.085714	0.006161	-13.91	<2e-16 ***
x5	0.320172	0.008462	37.84	<2e-16 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.3617 on 99 degrees of freedom

Multiple R-squared: 0.9435, Adjusted R-squared: 0.9423

F-statistic: 825.9 on 2 and 99 DF, p-value: < 2.2e-16

When I looked at the improvement in the results, I found that the best improvement was made by removing the factors 1,2 and 4.

**Last Equation :**

```
y_last <- 3.761437 + (-0.085714*30) + (0.320172*15.29) = 6.085447
```

**y = 3.761437 + (-0.085714)\*x3 + (0.320172)\*x5**

### Comparing factors

- Results (Equation results compared -- closest to real value is successful)
- Adjusted R-Squared(used to compare the accuracy of the models – The largest value is true)
- Residual standard error (Small is success)

	Results	Adjusted-R-Squared	RSE
First Version – With Five Factors	6.057265	0.9415	0.3643
Second Factor Removed	6.057265	0.9415	0.3643
Second and fourth Factor Removed	6.085059	0.9417	0.3636
First,Second and Fourth Factor Removed	6.085447	0.9423	0.3617