

School of Computer Science

Linear and General Regression Models

in Fulfilment of

MATH9903

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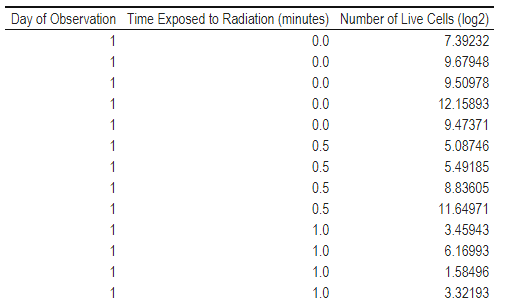
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**Date: 04/03/2022**

Introduction

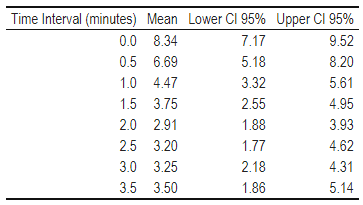
The skin cell data consists of 118 observations of human skin cells colonies. Samples where exposed to solar radiation from 0 (control) to 3.5 minutes in intervals of 0.5 minutes, this experiment was replicated over a ~28 days several times each day per interval of radiation exposure. Albeit some samples were contaminated and had to be discarded. The response variable for this analysis is the number of live cells remaining counted under a microscope after exposure to radiation documented in log2 form. The data are displayed in Table 1 in raw form.

Table 1: Skin Cell Data - first 13 observations



The question at hand is whether there is a statistically significant effect of solar radiation on the mortality of human skin cells. Table 2 presents the mean difference by cell survival count with upper and lower 95% confidence limits based on the student t distribution. There is a downward mean trend that levels off at the 2-minute interval showing an interesting rebound in both mean, lower and upper confidence intervals indicating variable survival rate of cells at longer exposure time. Further supported by Figure 1, group-wise confidence intervals overlap which results in the population mean difference in cell survivorship cannot be assumed statistically different.

Table 2: Cell survival count & 95% CI by time interval



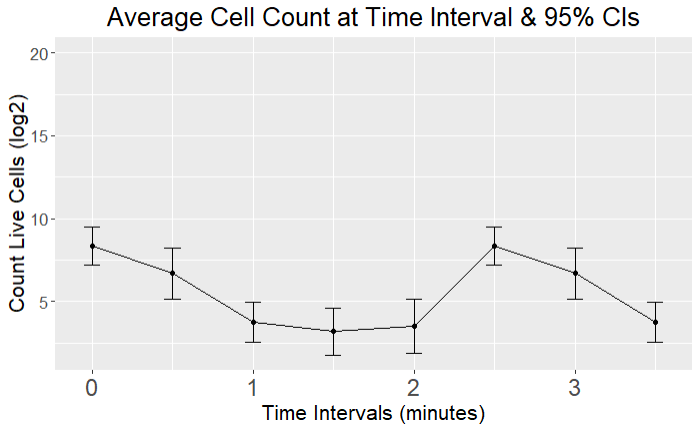
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Figure 1: Mean difference by cell survival count with upper and lower 95% confidence

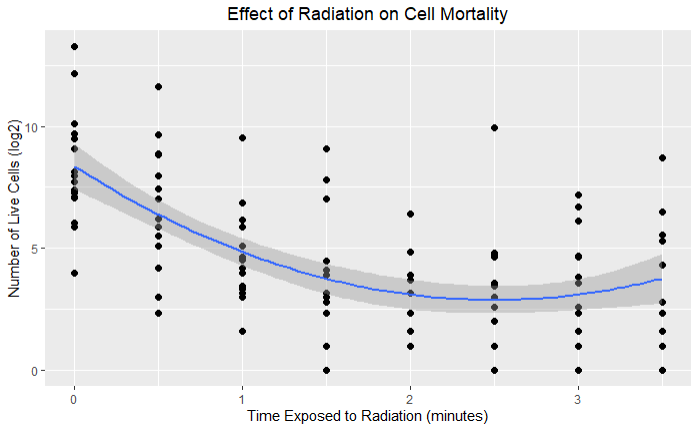


Figure 2: Scatterplot of Cell Survival Count

Model 1

A linear regression model was fitted to these data with *day of observation* as a categorical predictor and *time exposed to radiation (minutes)* as a continuous predictor. To assess interaction between *time exposed to radiation (minutes)*, *day of observation* was used as a factor to measure any variability imposed on *time exposed to radiation (minutes)*.

The model was: (fix the equation later)

Where is the live cell count for the sample , …….

This hypothesis was tested using the standard formulation for g

The intercept p = 2e-16 and time exposed to radiation p = 3.17e-05 were found to be statistically significant. Furthermore, time exposed to radiation had a t value of 10.798 indicating a large difference between samples. An interesting observation was made for day of observation factor 2 where it had a p value of 0.034 meaning there was a statistically significant interaction between differences in the lab environment. However, this was not the case for any other factor nor did it have an effect when combined with time exposed to radiation, meaning that this factor can be discarded as it is not statistically significant. This was further tested with the f-test statistic which was 0.445 on 3 and 7 degreed of freedom, yielding a p-value of 0.72. The null hypothesis in this case in not rejected.

Model 2

Common slope model:

Xxxxx

Leads to the following regression line

After removing day of observation as a factor from the model and rerunning the model on time exposed to radiation as a factor all time intervals have been found to be statistically significant. The effect of time exposed to radiation was statistically significant with an F-test statistic of 49.78 on 1 and 116 degrees of freedom, yielding a p-value of 1.339e-10. The slope for x was 6.13 at 2.5% and

Comparison among time intervals

Code sample

library(pastecs) #For creating descriptive statistic summaries

library(ggplot2) #For creating histograms with more detail than plot

library(psych) # Some useful descriptive functions

library(semTools) #For skewness and kurtosis

library(FSA) # For percentage

library(car) # For Levene's test for homogeneity of variance

library(effectsize) # To calculate effect size for t-test

library(kableExtra) # Used to generate report ready tables

# A copy paste of libraries I use in R projects

# Load data and setup project

setwd("~/GitHub/Masters-Classes-L-O/Applied Statistics/data\_stats")

getwd()

data = read.csv("skincells.csv")

anovatab <-

function(mod){

tab=as.matrix(anova(mod))

rows=dim(tab)[1]

moddf=sum(tab[,1])-tab[rows,1]

ssmodel=sum(tab[,2])-tab[rows,2]

msmodel=ssmodel/moddf

f=msmodel/tab[rows,3]

p=1-pf(f,moddf,tab[rows,1])

tab2=tab[(rows-1):rows,]

tab2[1,1:5]=c(moddf,ssmodel,msmodel,f,p)

tab2=rbind(tab2,c(moddf+tab2[2,1],ssmodel+tab2[2,2],rep(NA,3)))

rownames(tab2)=c('Model','Error','Total')

colnames(tab2)[1]='df'

return(print(tab2,na.print = "" , quote = FALSE,digits=3))

}

colnames(data)[1] = "Day of Observation"

colnames(data)[2] = "Time Exposed to Radiation (minutes)"

colnames(data)[3] = "Number of Live Cells (log2)"

tmp\_df = data # store reference

data$`Day of Observation` = as.factor(data$`Day of Observation`)

# Plots and table generation

scatter\_plot = ggplot(data, aes(x=`Time Exposed to Radiation (minutes)`,

y=`Number of Live Cells (log2)`,

show.legend = T)) +

geom\_point(size=2) +

ggtitle("Effect of Radiation on Cell Mortality") +

stat\_smooth(method = "lm", formula = y ~ x + I(x^2), size = 1) +

theme(plot.title = element\_text(hjust = 0.5))

scatter\_plot

data = tmp\_df # reset reference

kbl(data) %>%

kable\_classic(full\_width = F)

group\_means=by(data$`Number of Live Cells (log2)`, data$`Time Exposed to Radiation (minutes)`, t.test)

group\_means=matrix(c(unlist(group\_means[['0']][5:4]),

unlist(group\_means[['0.5']][5:4]),

unlist(group\_means[['1.0']][5:4]),

unlist(group\_means[['1.5']][5:4]),

unlist(group\_means[['2.0']][5:4]),

unlist(group\_means[['2.5']][5:4]),

unlist(group\_means[['3.0']][5:4]),

unlist(group\_means[['3.5']][5:4])),

nrow=8, ncol=3, byrow=T)

group\_means = data.frame(cbind(group\_means,c(0, 0.5, 1.0, 1.5, 2.0, 2.5, 3.0, 3.5)))

group\_means$X1 = round(group\_means$X1, digit=3)

group\_means$X2 = round(group\_means$X2, digit=3)

group\_means$X3 = round(group\_means$X3, digit=3)

colnames(group\_means)=c('Mean','Lower CL 95%','Upper CL 95%','Time Interval (minutes)')

ggplot(group\_means, aes(x=`Time Interval (minutes)`, y=group\_means[,1])) +

geom\_errorbar(aes(ymin=`Lower CL 95%`, ymax=`Upper CL 95%`), width=.1) +

geom\_line() +

geom\_point() +

expand\_limits(y=c(5, 20)) +

ylab("Count Live Cells (log2)") +

xlab('Time Intervals (minutes)') +

labs(title="Average Cell Count at Time Interval & 95% CIs") +

theme(text = element\_text(size=16), axis.text.x=element\_text(size=18), plot.title = element\_text(hjust = 0.5))

group\_means = group\_means %>% relocate(`Time Interval (minutes)`, .before = `Mean`)

kbl(group\_means) %>%

kable\_classic(full\_width = F)