





Linear Regression Day 2



Learning Objectives

- ☐ How to load data from a text file
- ☐ How to visualize data via a scatter plot
- □Describe a linear model for data
 - ☐ Identify the target variable and predictor
- □Compute optimal parameters for the model using the regression formula
- □Fit parameters for related models by minimizing the residual sum of squares
- □Compute the measure of the fit



- ☐ Motivating Example: Virus inactivation in wetland waters due to sunlight
- □Linear Model
- □Least Squares Fit Problem
- □Sample Mean and Variance
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- □ Assessing Goodness of Fit



- ☐ Getting the data:
- ☐ Data can be found here.





- ☐ Reading & Visualizing the Data:
- ☐ Using Python packages Pandas, Numpy, Matplotlib.
- Pandas:
 - Used for reading and writing data files
 - Loads data into dataframes
- Numpy:
 - Used for numerical operations, including linear algebra
 - Data is stored in ndarray structure
 - We convert from dataframes to ndarray
- ☐Matplotlib:
 - ☐ Used for MATLAB-like plotting and visualization

import pandas as pd
import io

import numpy as np

import matplotlib
import matplotlib.pyplot as plt
%matplotlib inline



- ☐ Reading the data using python's pandas library:
- □ <u>pd.read_csv</u> converts a comma-separated values file into a 2D data structure with labeled axes.

	Time (h)	Ct1 Clear	Ct1 5cm	Ct1 20cm	Ct2 Clear	Ct2 5cm	Ct2 20cm
0	0.00	1000000.000000	1000000.000000	1000000.000000	1000000.000000	1000000.000000	1000000.000000
1	0.75	NaN	547323.127588	680310.384129	NaN	427843.265172	724747.957348
2	1.00	124572.849877	NaN	NaN	124242.476640	NaN	NaN
3	1.50	NaN	191599.336146	434275.771006	NaN	163016.250572	491827.603978
4	2.00	23632.953475	NaN	NaN	28479.804254	NaN	NaN
5	2.25	NaN	66699.257600	267168.775510	NaN	60215.597490	224766.301820
6	3.00	2364.275054	21358.069727	123072.776978	3703.735805	15289.278797	112632.545682
7	4.00	472.992223	4481.980885	69507.072936	569.997564	3151.417918	33667.665216

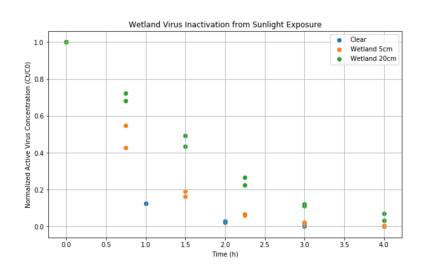
```
import pandas as pd
import io

df = pd.read_csv(io.BytesIO(uploaded[filename]))
df
```



- ☐ Visualizing the Data:
- ☐ It's always a good idea to visualize the data before performing any operations on it.
- ☐ Using python's Matplotlib:
- \square plt.scatter(x,y) plots a scatter plot of y vs x.

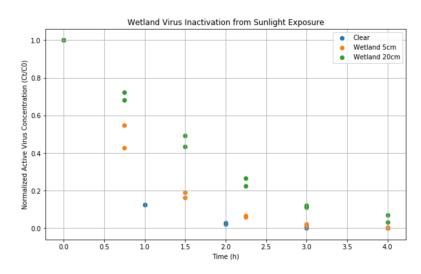
```
plt.scatter(cat_time[-nanloc_clear], ct_clear[-nanloc_clear])
plt.scatter(cat_time[-nanloc_wetla], ct_5cm[-nanloc_wetla])
plt.scatter(cat_time[-nanloc_wetla], ct_20cm[-nanloc_wetla])
plt.grid(True)
plt.legend(['Clear', 'Wetland 5cm', 'Wetland 20cm'])
plt.title('Wetland Virus Inactivation from Sunlight Exposure')
plt.xlabel('Time (h)')
plt.ylabel('Normalized Active Virus Concentration (Ct/C0)');
```





Exercise: Postulate a Model

- ☐ Try to find a mathematical model to predict the active virus concentration from time:
 - Try to make a reasonable/eyeball guess, without using a program.



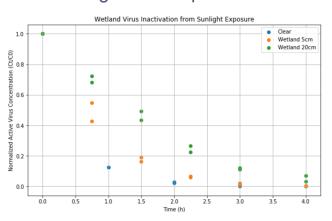


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Understanding the Data:

- □ Our y axis: The variable we are trying to predict. Can be called: Dependent variable, response variable, target, regressand, ...
- ☐ Our x axis: The variable we are using to predict. Can be called: Predictor, attribute, covariate, regressor ...
- □ Each data point is called a sample. In this example, we are using a scatter plot to view the samples.





Linear Model

☐ Assume a linear relationship among the samples - using the intercept (beta0) and slope (beta1).

$$\beta_0 = c_a - \beta_1 t_a \qquad | \beta_1 = \frac{c_b - c_a}{t_b - t_a} |$$

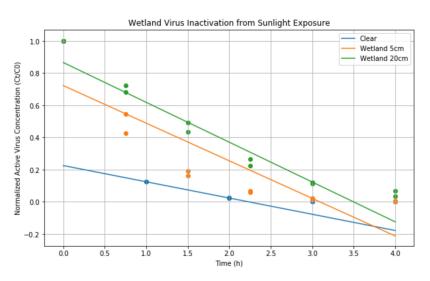
□Why do we use a Linear Model?

- ☐ Generally, most natural phenomena have a linear relationship
- □ Simple computation, and easy to interpret.



Linear Model

- ☐ Plotting the Linear Model using python's Matplotlib:
- □ *plt.plot* is used to plot y versus x as lines or markers.





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Linear Model Residual

- ☐ As we can see, a linear model does not fit all the sample, which means it is not a good fit for our data.
- ☐ We add a residual term to our lineal model (e):

$$y = \beta_0 + \beta_1 x + \epsilon$$

- \square For a $\beta = (\beta_0, \beta_1)$
- We define a residual sum of squares (RSS): $RSS(\beta_0, \beta_1) := \sum_{i=1}^{\infty} (y_i \hat{y}_i)^2$
- $\square \text{A Least Squares Solution is to find } (\beta_0, \beta_1) \text{ to minimise RSS.} \sum_{i=0}^{N-1} (y_i \hat{y}_i)^2 = \sum_{i=0}^{N-1} (y_i \beta_0 \beta_1 x_i)^2$



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Least-Squares Fit Solution: Sample Mean and Standard Deviations

Given data: $(x_i, y_i), i = 1, ..., N$

Sample mean
$$\bar{x} = \frac{1}{N} \sum_{i=1}^{N} x_i$$
, $\bar{y} = \frac{1}{N} \sum_{i=1}^{N} y_i$

Sample variances

$$s_x^2 = \frac{1}{N} \sum_{i=1}^{N} (x_i - \bar{x})^2$$
, $s_y^2 = \frac{1}{N} \sum_{i=1}^{N} (y_i - \bar{y})^2$

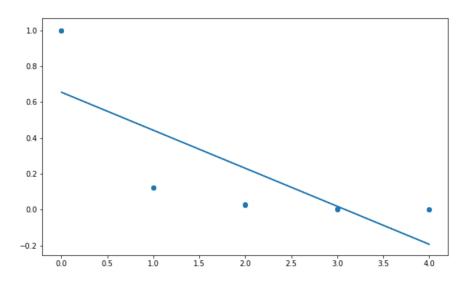


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Least-Squares Fit Solution

- ☐ Using python to find the Least Squares solutions
- □ <u>np.mean</u> uses python's numpy library to compute the arithmetic mean along the specified axis.



```
# Calculate the mean of x and y
xm = np.mean(x)
ym = np.mean(y)

syy = np.mean((y-ym)**2)  # Variance of y
syx = np.mean((y-ym)*(x-xm))  # Covariance of x and y
sxx = np.mean((x-xm)**2)  # Variance of x

beta1 = syx/sxx
beta0 = ym - beta1*xm
print('beta0 = {:.2f}, beta1 = {:.2f}'.format(beta0,beta1))

beta0 = 0.65, beta1 = -0.21
```



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Assessing the goodness of the fit

- ☐ We can use the R2 score to estimate the goodness of our fit.
- ☐ The best (and maximum) R2 score is 1. It can be negative if the fit is really bad.
- ☐ It can be calculated on python by:

```
RSS = np.sum((y - y_hat)**2)
N = y.size  # Number of samples in the data set
R2 = 1 - (RSS/N)/syy
print('R^2 = {:.2f}'.format(R2))
```

```
R^2 = 0.60
```