

# Statistical Inference and Multivariate Analysis (MA324)

## LECTURE SLIDES Lecture 25

### Linear Regression



Indian Institute of Technology Guwahati

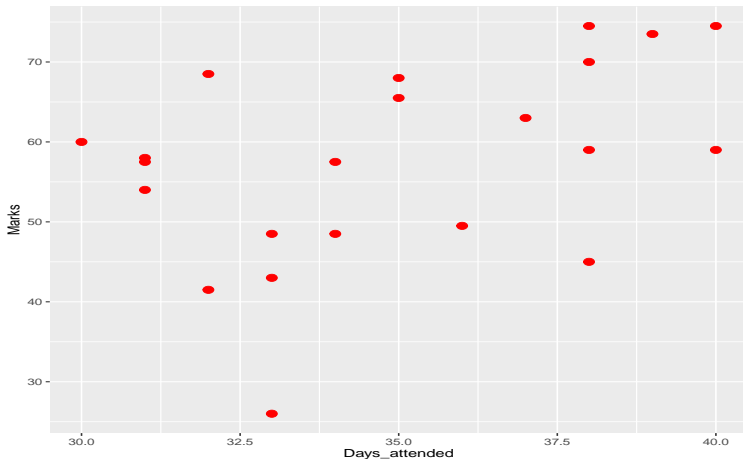
Jan-May 2023

# Research Question: What is the impact of attending classes on students' final marks

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Let's start with a real data from IITG which you can feel about it!!

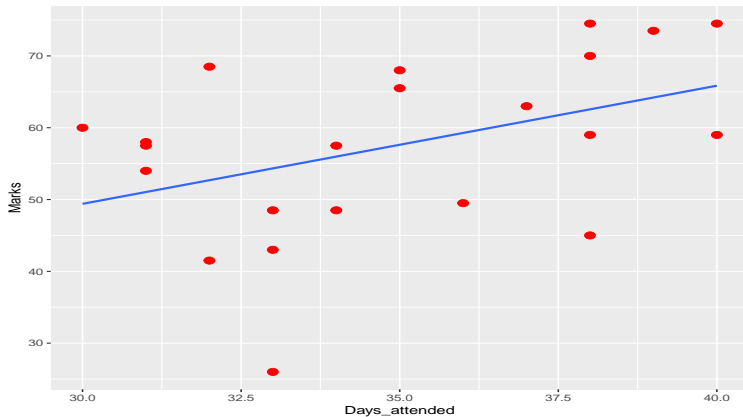
Scatter plot of number of days attended and marks by the students in MTech Data Science course (MA589)



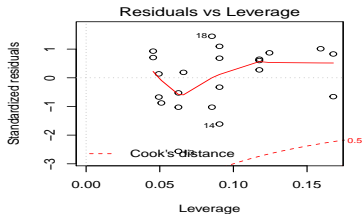
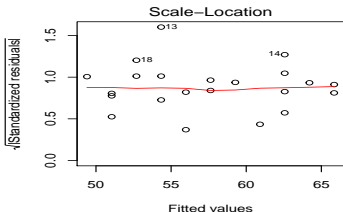
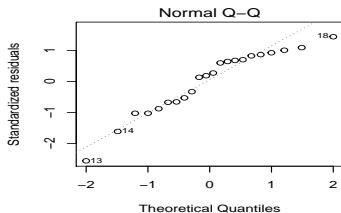
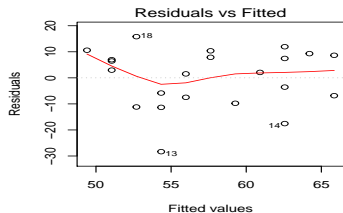
- We will do a regression analysis to the above data to find the answer. Is using regression appropriate here?
- **Strong Advice:** As a Data Scientist/Statistician, you have to **spent a lot of your time** and effort to **pre-process/clean the raw data** to make it **analysis ready**. It is part of your job to clean the data: so learn it quickly!!
- Here is the output after fitting a regression line in R:

# Fitting Regression line

Scatter plot of number of days attended and marks by the students in MTech Data Science course (MA589):  
with fitted regression line



# Residual Analysis



# Linear Models: Simple and Multiple Linear Regressions

The regression framework can be characterized in the following way<sup>1</sup>:

- We have one particular variable that we are **interested in understanding or modeling**, such as sales of a particular product, sale price of a home, or voting preference of a particular voter. This variable is called the **target, response, or dependent variable, and is usually represented** by  $y$ .
- We have a set of  $p$  **other variables** that we think might be **useful in predicting or modeling the target variable** (the price of the product, the competitor's price, and so on; or the lot size, number of bedrooms, number of bathrooms of the home, and so on; or the gender, age, income, party membership of the voter, and so on). These are called the **predicting, or independent variables, and are usually represented** by  $x_1, x_2, \dots, x_p$ .

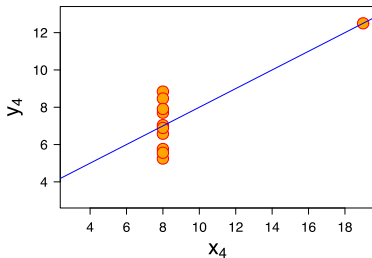
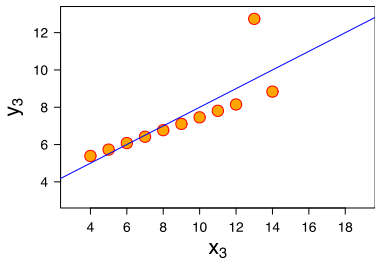
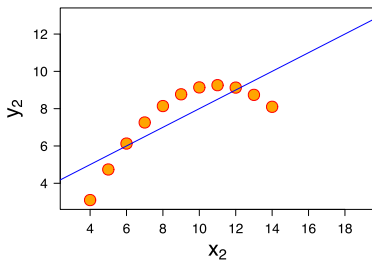
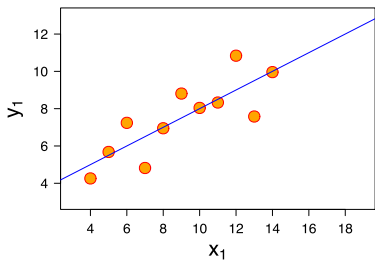
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<sup>1</sup>Handbook of Regression Analysis. By Samprit Chatterjee and Jeffrey S. Simonoff.

- Typically, a regression analysis is used for one (or more) of three purposes:
  - 1 **modeling the relationship** between  $x$  and  $y$ ;
  - 2 **prediction of the target** variable (forecasting);
  - 3 and **testing of hypotheses**.



# Importance of graphing data before analyzing it



- Which one of the above do you think has **highest correlation and ideal for linear regression?**

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<sup>2</sup>The graph is taken from the Wikipedia

- Which one of the above do you think has **highest correlation and ideal for linear regression?**
- In **all the four graphs**: mean of  $x = 9$  (with variance 11); mean of  $y = 7.50$  (with variance 4.1) ; correlation between  $x$  and  $y = 0.816$
- Fitted linear regression in each cases:  $y = 3 + 0.5x$
- In 1973, **Anscombe** demonstrated the **importance** of **graphing data**<sup>2</sup> before analyzing it and the **effect of outliers** on statistical properties

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<sup>2</sup>The graph is taken from the Wikipedia

# Linear Models: Meaning of Linearity and Transformation

## What does “Linear” mean?

- A linear model is **linear in the parameters  $\beta$** , but not necessarily in the  $x$ 's, e.g.
  - \*  $y = \beta_0 + \beta_1 x + \beta_2 x^2$  is a **linear model**, because it is linear in  $\beta$  (even though not in  $x$ ).
  - \*  $y = \beta_0 + \beta_1 x^{\beta_2}$  is a **non-linear model**, because it is not linear in  $\beta$ .

## Transformation

- Clearly  $y = f(x) = \beta_0 x^{\beta_1}$  is **not a linear model**, but

$$\ln f(x) = \ln \beta_0 + \beta_1 \ln x$$

If we let  $f^* = \ln f$ ,  $\beta_0^* = \ln \beta_0$ ,  $\beta_1^* = \beta_1$ ,  $x^* = \ln x$ , we have

$$f^* = \beta_0^* + \beta_1^* x^*,$$

which is **a linear model**.

# Transformation

- Thus, although linear models seem to be simple and restrictive, they **can actually be quite flexible by transformation of the response and the predictors.**
- Linear models are **not just straight lines, they can be curved.** Can you give an examples?

Transform the following **non-linear models to linear models:**

- $y = \frac{e^{\beta x}}{1+e^{\beta x}}$ , where  $y \in (0, 1)$ .
- $y = \frac{1}{\beta_0 + \beta_1 x_1 + \beta_2 x_2}$ .
- $y = e^{e^{\beta x}} - 1$ , where  $y > 0$ .