

Self Notes - Week2

Simple Linear Regression

Important Notes:

1. Assuming estimation is most important for this course in terms of examination point of view.
2. Difference between deterministic relationship and statistical relationship.
3. **Least squares regression** - we sum the square of the errors and with that we can minimise the errors.
4. For MCQs - we can get a question like give the assumptions or list the assumptions from the given plot (slide 26)
5. Hypothesis testing - <https://www.section.io/engineering-education/hypothesis-testing-data-science/>

Extra Notes:

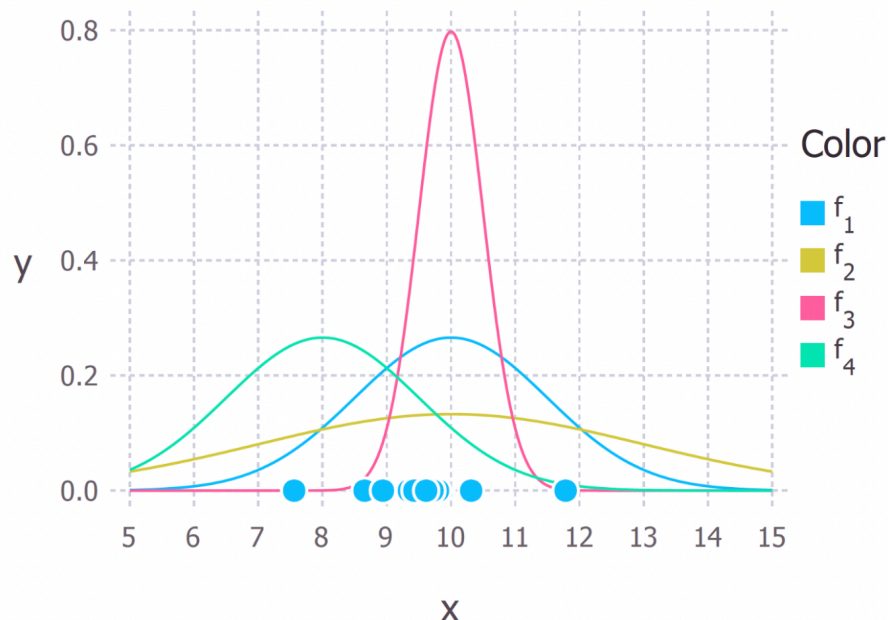
1. In Linear Models we have 1 target variable and 1 predictor variable
2. For estimating the population variable we generally use Alpha, Beta for linear models (equation of a line) or Beta0, Beta1 etc.
3. we generally use least square estimates method for linear regression.
4. ANOVA - Analysis of variance table to test the significance of the parameters.

5. The more the scatter, the less the accuracy of the model.
 6. Model checking - testing the assumptions.
 7. **Slope** and **intercept** are the two measures that describes the strength of linear model.
 8. **Intercept(alpha)** is the value of "y" when "x" equals 0.
 9. **Slope(beta)** is how much the value for "y" changes when we change the value for "x".
 10. if slope increases or if slope is positive then when x increases, y increases. If slope is negative then when x increases, y decreases.
 11. sd and variance is a spread in a plot.
 12. R code - `lm(Sales~Type)` to check the significance.
 13. R code - `confint(model)` is confident interval
 14. R code - `anova(model)` is used to determine the coefficient of determination(R^2) - Residuals are not captured from regression (can be say it as residuals = errors [need to check])
 15. Residual mean square is the estimate of the variance of Y variable.
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Paragraph for more explanation or useful links:

- Maximum Likelihood estimates is a method that determines the parameters of a

model. The parameter values are found such that they maximise the likelihood or let's say maximise the correctness of data points for a function. Good read: <https://towardsdatascience.com/probability-concepts-explained-maximum-likelihood-estimation-c7b4342fdbb1>



The 10 data points and possible Gaussian distributions from which the data were drawn. f_1 is normally distributed with mean 10 and variance 2.25 (variance is equal to the square of the standard deviation), this is also denoted $f_1 \sim N(10, 2.25)$. $f_2 \sim N(10, 9)$, $f_3 \sim N(10, 0.25)$ and $f_4 \sim N(8, 2.25)$. The goal of maximum likelihood is to find the parameter values that give the distribution that maximise the probability of observing the data.

- Least square estimates is a method of estimating the values of alpha, beta such that it minimises the error. For this we can think of finding the **loss function** first for the error and then **using square of partial derivative of this loss function** we can find the least square estimates.
- **Deterministic relationship** is when all the points lie on the straight-line or the curve which we used as a function.
- **Statistical relationship** is when we get a line which is in between those scattered data points and for which we can get our

population parameters.

- **Normal Distribution** - <https://www.analyticsvidhya.com/blog/2021/05/normal-distribution-an-ultimate-guide/>