

Emory University
QTM 220 Regression Analysis
Learning Notes

Jiuru Lyu

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Contents

1	Descriptive Statistics and Binary Covariates
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2

1 Descriptive Statistics and Binary Covariates

Definition 1.1 (Location). The *location* of the data is where it is. It is about approximating the data by a constant.

$$Y_i \approx \mu, \quad \text{for } i = 1, \dots, n$$

Example 1.2 Different ways to summarize location: mean, median

Definition 1.3 (Spread). The *spread* of the data is how far it tends to be from its location.

Definition 1.4 (Residuals). Spread summarizes the size of the *residuals* left over after constant approximation. We use $\hat{\varepsilon}$ to denote residuals.

$$\hat{\varepsilon}_i := Y_i - \hat{\mu}.$$

Definition 1.5 (Median Absolute Deviation and Standard Deviation).

- The *median absolute deviation (MAD)* is the median size of residuals.
- The *standard deviation (sd)* is the square root of the mean squared size of residuals.

Remark 1.1 The standard deviation is a sort of average in which big residuals count more than smaller ones.

Definition 1.6 (Distribution). We use *histograms* to summarize the *distribution* of the data.

Remark 1.2 Distribution of the data tells us more information than location and spread, but less than dot plot. For example, in this context, dot plot also includes the identities of the individuals in addition to the number of people having salary in the range.

Definition 1.7 (Binary Data). Binary data only have two options, and we usually denote those two options as 1's and 0's.

Corollary 1.8 : Hence, when drawing a dot plot, everyone falls into either of the two lines representing 1 and 0.

Theorem 1.9 Location of Binary Data

The median is whichever outcome is the most common, and the mean is the proportion of 1's in the data.

Remark 1.3 Hence, a histogram tells us no more information than $\hat{\mu}$.

Theorem 1.10 Spread of Binary Data

- Median absolute deviation will always be 0 in a binary case.
- The standard deviation is the square root of the mean squared distance from the mean, and

$$\text{sd} = \sqrt{\hat{\mu}(1 - \hat{\mu})}.$$

Proof 1. The claim concerning MAD is trivial. *Hint: there's only two possible values in the data, so median and MAD should always be the same.*

Now, let's consider the claim on standard deviation.

$$\begin{aligned}\text{sd}^2 &= \frac{1}{n} \sum_{i=1}^n (Y_i - \hat{\mu})^2 \\ &= \frac{1}{n} \sum_{y \in \{0,1\}} \sum_{i: Y_i=y} (Y_i - \hat{\mu})^2 \\ &= \frac{1}{n} \{N_1(1 - \hat{\mu}^2) + (n - N_1)(0 - \hat{\mu}^2)\} & [N_1 = \text{number of 1's}] \\ &= \frac{1}{n} \{N_1(1 - 2\hat{\mu} + \hat{\mu}^2) + (n - N_1)\hat{\mu}^2\} \\ &= \frac{1}{n} \{N_1 - 2N_1\hat{\mu} + n\hat{\mu}^2\} \\ &= \frac{1}{n} \{n\hat{\mu} - 2n\hat{\mu} \cdot \hat{\mu} + n\hat{\mu}^2\} & [N_1 = n\hat{\mu}] \\ &= \frac{1}{n} \{n\hat{\mu} - n\hat{\mu}^2\} \\ &= \hat{\mu} - \hat{\mu}^2 = \hat{\mu}(1 - \hat{\mu}).\end{aligned}$$

Therefore, we know

$$\text{sd} = \sqrt{\hat{\mu}(1 - \hat{\mu})}.$$

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Remark 1.4 *In binary data, knowing the mean \equiv knowing everything else.*