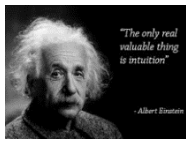


Variance, Covariance, Correlation and Analysis of Variance





Explaining Variance: Intuition

- Understanding **central tendencies** (e.g., means, median) and **dispersion** around these tendencies (e.g., variance, standard deviation) is central to statistics, machine learning and predictive modeling.
- Variation of single attributes → e.g., **variance**
- Comparing variance of 2 attributes → e.g., **ANOVA**
- Analyzing how 2 or more attributes vary together → **covariance, correlation**
- **Correlation** → measure of statistical association
- **Prediction** → estimating outcomes based on observed associations – e.g., regression, decision trees, etc.
- **Causation** → harder to prove; prediction does not imply causation; need more rigorous methods to prove causation

Variance and Covariance Refresher

- A lot of what we do in analytics is understanding and explaining variance and covariance. **Knowing these concepts is key.**
- **Variance** is how much values vary relative to the mean. The value is squared so that values, both below and above the mean contribute positively to the variance statistic.

$$Var = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1}$$

- **Covariance** indicates how two variables are related – i.e., how do they “co-vary” or how they move together. If when x is above (or below) the mean y is generally above (or below) the mean, then the covariance is positive. Otherwise is negative.

$$Cov(x, y) = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n - 1}$$

- Covariance is a useful concept, but it has limited practical use because the covariance value is dependent on the **scale** used to measure x and y. **Correlation** takes care of this scale problem.



Correlation Refresher

- Correlation is like covariance, but the deviation from the mean of each variable are **divided by the standard deviation** of the variable → it can be shown mathematically that the correlation of two variables will not change with re-scaling.
- Mathematically, correlation statistics ranges from **-1.0** to **1.0**
- This is really a **descriptive analytics** method, but it is a necessary first step before predictive analytics
- Provides an indication for whether two variables vary in the same or opposite direction, or if they are **independent** from each other

$$\rho(x, y) = \frac{1}{n-1} \sum_{i=1}^n \left(\frac{x_i - \bar{x}}{\sigma_x} \right) \left(\frac{y_i - \bar{y}}{\sigma_y} \right) = \frac{Cov(x, y)}{\sigma_x \sigma_y}$$

- Analyzing the **descriptive statistics** (i.e., mean and standard deviation) and the **correlation** among all variables is a necessary **first step** (i.e., descriptive analytics) before building predictive models. This is often called **“eye-balling the data”**



Correlation Analysis – 2 Key Values

1. **Magnitude** – how large is the association

$\rho = 1.0$ Perfectly positively correlated

$\rho = +$ Positively correlated

$\rho =$ Around 0 Uncorrelated (i.e., independent)

$\rho = -$ Negatively correlated

$\rho = - 1.0$ Perfectly negatively correlated

2. **Significance** – probability that the observed correlation happened by chance – i.e., **$p \rightarrow \text{prob}(p=0)$**

$p > 0.10 \rightarrow$ Not significantly $\neq 0$ – i.e., independent

$p < 0.10 \rightarrow$ Moderately significant

$p < 0.05 \rightarrow$ Significant

$p < 0.01 \rightarrow$ Very significant

$p < 0.001 \rightarrow$ Highly significant

It is useful to look at the **correlation matrix**

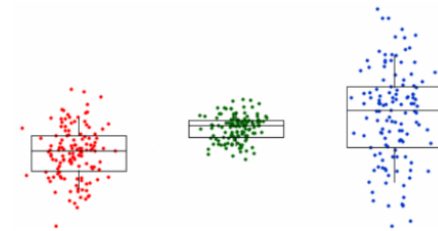


Correlation Matrix (Automobile Data)



ANOVA Refresher

- **Analysis of variance** (ANOVA) provides a statistical test of whether the mean of a given variable is equal among two or more groups.
- It is called analysis of “**variance**” and not analysis of “**means**” because it compares the variance within each group against the variance of the means between groups.
- For example, if we want to test if the mileage is different between foreign and domestic cars, or if the price of a diamond is different for various color classifications, we can do an ANOVA test.
- The **intuition** is that if the variance **between group** means is significantly larger than the variance within groups, then the means are significantly different. Otherwise they are not.
- **ANOVA** and **regression** are tightly related because ANOVA tests whether the **variance explained** by the regression line (or the various variables in the model) is significantly different than the variance of the dependent variable alone.
- As we will see later, ANOVA is very useful when **comparing** whether **one regression model** explains more variance than **another**, so it is a key test when evaluating predictive models.



Tips

`ggplot2{ggplot}` → `ggplot2` library in the `ggplot` package is very popular for statistical plots and graphs

`cor()`, `var()`, and `cov()` → in the `{stats}` library provide the correlation, variance and covariance for a matrix or data frame

`rcorr{Hmisc}` → provides more complete correlation data, like p-values

`ggpairs{GGally}`, `pairs{GGally}` → correlation matrix with visual scatterplots

`corrplot{corrplot}` → correlation matrix with visual scatterplots

`aov{stats}` → Traditional ANOVA to test differences in means; yields the same results as `lm()`, except with “repeated measures” (e.g., one person provides multiple observations – for example: recovery time with and without medicine) – `aov()` is preferred in such cases

`anova{stats}` → ANOVA → Used primarily to compare 2 linear models

`boxplot{graphics}` → Graphical contrast of means



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