

# PRINCIPAL COMPONENT ANALYSIS (PCA)

## WHAT IS PCA?

*Principal Component Analysis (PCA) is a statistical technique used to reduce the dimensionality of a data set while retaining as much of the original information as possible. This is a very popular preprocessing step for other analyses.*

*This is done by linearly transforming the initial data into a new coordinate system where most of the variation in the data can be described by fewer dimensions than the initial data.*

### When to use PCA?

PCA is used when analyzing data sets with many correlated variables. By reducing the dimensionality, PCA can:

- Make it easier to visualize and analyze data
- Decrease computation time in code
- Reduce noise and detect outliers in the dataset
- Help mitigate the problem of overfitting

**Popular places where PCA is used:** Computer Vision, bioinformatics, machine learning, speech processing, and many more!

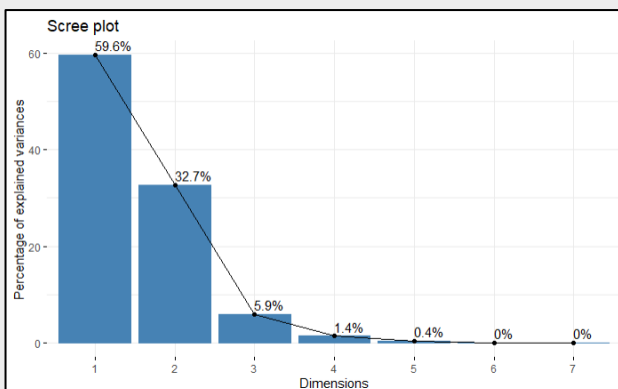
### What are the data requirements and assumptions for PCA?

- Data must be numeric
- Data must have at least three features / variables
- Data must be linear (assess visually with pairwise plots or matrix scatter plots)
- Data must be standardized and continuous
- Data with missing values must be removed
- Data set should be highly correlated (assess with Pearson correlation)

### How PCA works:

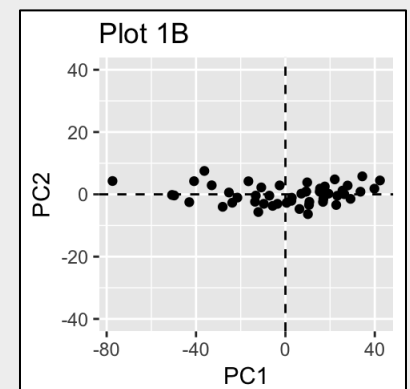
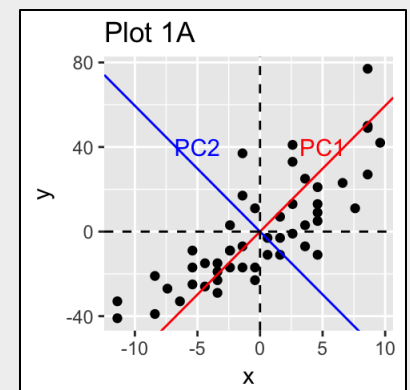
**NOTE: This is the Singular Value Decomposition (SVD) method of performing PCA.**

1. **Standardize the data!** Data must be on the same scale.
2. **Find the first principal component (PC1)!** Find the best fit multiple regression line through the data.
3. **Find the second principal component (PC2)!** Find the best line of fit that is perpendicular to PC1.
4. **Repeat for each variable!** Find the PCs for each variable.
5. **Interpret the results!** Analyze the relationship between variables using the PCs.



*Figure 2: Image of Scree plot to help determine how many PCs are necessary to explain a percentage of variance in the data.*

*Look for the "elbow" or point where the curve flattens for the optimal number of components to retain*



*Figure 1: Plot 1A is a scatterplot with identified principal components; Plot 1B is reduced data by projecting each sample onto the first PC*

*Resources:*

[StatQuest: Principal Component Analysis \(PCA\), Step-by-Step](#)

[Data Camp: Principal Component Analysis in R Tutorial](#)

[Toward Data Science: Principal Component Analysis \(PCA\) 101, using R](#)

[BuiltIn: Principal Component Analysis](#)

[UC business Analytics R Programming Guide: Principal Component Analysis](#)

[STHDA: Principal Component Methods in R: Practical Guide](#)

[Statology: Principal Component Analysis in R: Step-by-Step Example](#)

[Keboola: A Guide to Principal Component Analysis \(PCA\) for Machine Learning](#)

[Geeks for Geeks: Principal Component Analysis with R Programming](#)

[CRAN: Step-by-Step PCA](#)