





- Motivation of Dimension Reduction:
 - Imagine a dataset with 30+ features, how would you understand the key features?
 - Visualization and Data Analysis have limitations when the number of feature dimensions increases.





- Important Note:
 - Dimensionality Reduction algorithms such as PCA <u>do not</u> simply choose a subset of the existing features.
 - They create <u>new</u> dimensional components that are combinations of proportions of the existing features.





- Dimension Reduction
 - Helps visualize and understand complex data sets.
 - Can also act as a simpler data set for training data for machine learning algorithms.
 - Reduce dimensions then train ML Algorithm on smaller data set.





- Variance Explained
 - We've often seen that certain features are more important or have more explanatory power than other features.
 - For example, size of a house is probably much more important than the color of a house when explaining the price of a house for sale.





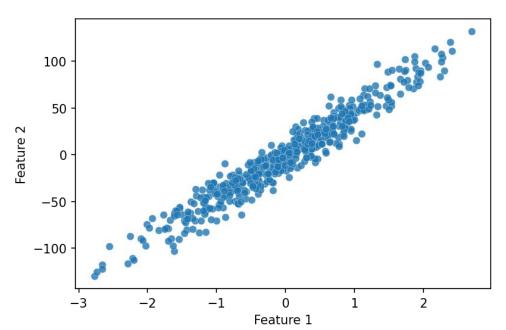
- Variance Explained
 - This idea of more important features is easy to understand when we can directly correlate features to a known label. But what about unlabeled data?
 - What measurement can we use to determine feature "importance"?





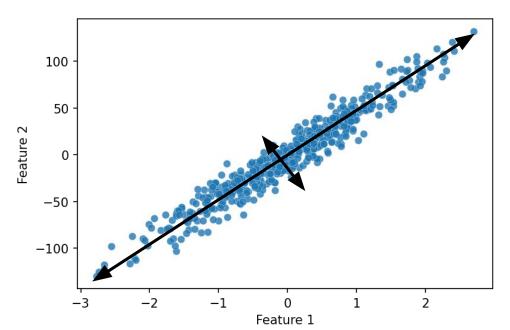
- Variance Explained
 - Measure the proportion to which each feature accounts for dispersion in the data set.





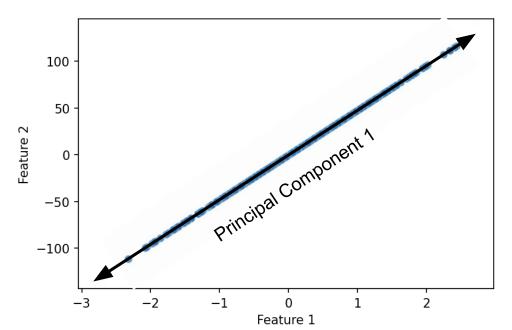






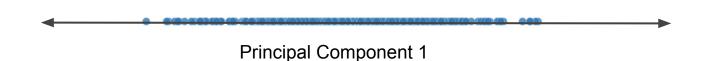














- Variance Explained
 - Principal Component is a linear combination of original features.
 - The more variance the original feature accounts for, the more influence it has over the principal components.





- Variance Explained
 - Here we went from 2 features down to 1 principal component.
 - This single principal component can "explain" some percentage of the original data, for example 90% of variance explained by principal component.





- Variance Explained
 - 100% of the variance in the data is explained by all the original features.
 - We trade off some of the explained variance for less dimensions.
 - This can be significant savings for data sets with many dimensions, but only a few strong features.





