

# **Principal Component Analysis**

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## **What is PCA?**

**PCA** is a technique that reduces the number of variables in a dataset while preserving as much important information as possible.

## Cognate/Professional Electives

Think of it as summarizing a long book into a short summary that **still captures** the main ideas.

## **Why do we need PCA?**

**Simplification:** Dealing with too many variables can be confusing. PCA helps by reducing the number.

## **Why do we need PCA?**

**Visualization:** It's hard to visualize data with many dimensions. PCA allows us to plot data in 2D or 3D.

# **Why do we need PCA?**

**Noise Reduction:** It filters out insignificant details, focusing on what's truly important.

# How does PCA work?

# 1. Collecting Data

Suppose you have information about different fruits:

**Apple:** Red color, round shape, sweet taste.

**Banana:** Yellow color, long shape, sweet taste.

**Lemon:** Yellow color, oval shape, sour taste.

You have multiple characteristics (variables) for each fruit.



## **2. Standardizing the Data**

Before comparing, we need to make sure each characteristic is on the same scale. It's like adjusting the units so everything is measured similarly.

### 3. Finding Patterns

PCA looks for patterns in the data. It identifies characteristics that **vary** the most among the fruits.

In our example, taste might vary more than color.

## **4. Creating Principal Components**

These patterns are turned into new variables called **principal components**. Each principal component is a combination of the original characteristics.

**Principal Component 1 (PC1):** Might represent overall sweetness.

**Principal Component 2 (PC2):** Could represent shape differences.

## 5. Reducing Dimensions

We can now describe each fruit using these principal components instead of the original characteristics.

Often, the **first few principal components** capture most of the important information.

### An Everyday Analogy

Think about taking a photo of a 3D object.

The photo is 2D, but it **still shows** the **essential features** of the object.

PCA does something similar by projecting **high-dimensional** data onto **fewer dimensions**.

**Thank you very much for listening.**