Z-scores: Takeaways 🖻

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Syntax

• Writing a function that converts a value to a z-score:

```
def z_score(value, array, bessel = 0):
    mean = sum(array) / len(array)

from numpy import std
    st_dev = std(array, ddof = bessel)

distance = value - mean
    z = distance / st_dev

return z
```

• Standardizing a Series :

• Transforming a standardized distribution to a different distribution, with a predefined mean and standard deviation:

Concepts

- A **z-score** is a number that describes the location of a value within a distribution. Non-zero z-scores (+1, -1.5, +2, -2, etc.) consist of two parts:
 - *A sign*, which indicates whether the value is above or below the mean.
 - *A value*, which indicates the number of standard deviations that a value is away from the mean.
- The z-score of the mean is o.
- To compute the z-score for a value coming from a population with mean and standard deviation , we can use this formula:
- To compute the z-score for a value coming from a sample with mean and standard deviation , we can use this formula:
- We can **standardize** any distribution by transforming all its values to z-scores. The resulting distribution will have a mean of 0 and a standard deviation of 1. Standardized distributions are often called **standard distributions**.
- Standardization is useful for **comparing values** coming from distributions with different means and standard deviations.
- We can transform any population of z-scores with mean and to a distribution with any mean and any standard deviation by converting each z-score to a value using this formula:

• We can transform any sample of z-scores with mean and to a distribution with any mean and any standard deviation by converting each z-score to a value using this formula:

Resources

• The z-score() <u>function from</u> scipy.stats.mstats — useful for standardizing distributions.

• The Wikipedia entry on z-scores.



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