

Module 4: Z-scores, Normal Curve, and Simple Probability

Jon Starkweather, PhD

`jonathan.starkweather@unt.edu`

Consultant

Research and Statistical Support



Introduction to Statistics for the Social Sciences



The RSS short courses

The Research and Statistical Support (RSS) office at the University of North Texas hosts a number of “Short Courses”. A list of them is available at:

<http://www.unt.edu/rss/Instructional.htm>

Outline

1 Introduction

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- 1 Introduction
- 2 Z-scores

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- 4 Percentages and simple Probability

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- 2 Z-scores
- 3 Normal Curve
- 4 Percentages and simple Probability
- 5 Summary of Module 4

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- An individual scores 26 on a leadership test.
 - What does that mean? Are they a good leader or a bad leader?
 - If the mean of the test is 20, then we know that the person is above average in leadership (compared to others who have taken the test).

Example 1

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- The person with the score of 26 is two standard deviations above the mean.

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- Planning Ability

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- A person scores 84 on a test of planning ability.

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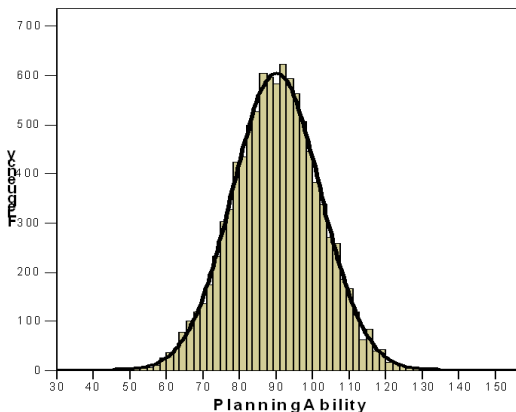
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- If $\bar{X} = 90$, $S = 12$, this person's score is 0.5 standard deviations below the mean.
- Thus the person is below average, but not by a lot.

Planning ability: $\bar{X} = 90, S = 12$

Histogram



We can visualize that the score of 84 would be slightly lower than the mean.

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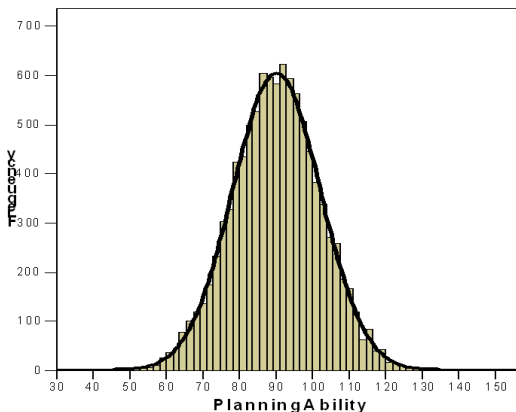
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- What if a person scored 78?
 - Then the person is below the average by about the average amount that scores deviated from the mean.
- Recall, standard deviation *is* the average amount the scores *deviated* from the mean.

Planning ability: $\bar{X} = 90, S = 12$

Histogram



Visualize the score of 114 as far above the mean, while the score of 78 would be slightly below the mean.

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 - So, if $S = 12$, then the average deviation for scores from our Planning Ability test mean is 12.
- The number of standard deviations a particular score is above or below the mean is called its **Z-score**.
 - If a score is below the mean, the Z-score is negative.
 - If a score is above the mean, the Z-score is positive.
- The standard deviation now becomes a kind of yardstick, a unit of measure in its own right.

Rulers Analogy

The two scales are something like a ruler with inches lined up on the one side and centimeters on the other.

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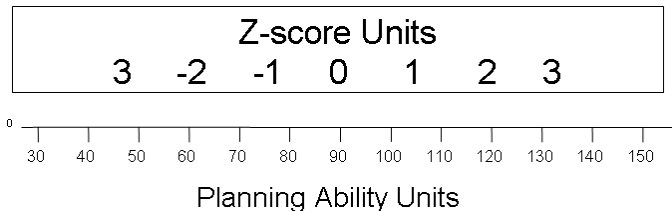
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The two scales are something like a ruler with inches lined up on the one side and centimeters on the other.

- A **raw** score of 78 is 1 standard deviation below the mean and so, it has a **Z-score** of -1.
- **Raw scores:** $\bar{X} = 90$, $S = 12$ and **Z-scores:** $\bar{X} = 0$, $S = 1$.



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Z-scores' usefulness

- Z-scores provide a helpful way to compare scores on measures that are on completely different scales.
- If a person scored:
 - 26 on Leadership (example 1: $\bar{X} = 20, S = 3$) and
 - 84 on Planning Ability (example 2: $\bar{X} = 90, S = 12$)
- We can say that person's scores are much higher than average on Leadership and slightly lower than average on Planning Ability.

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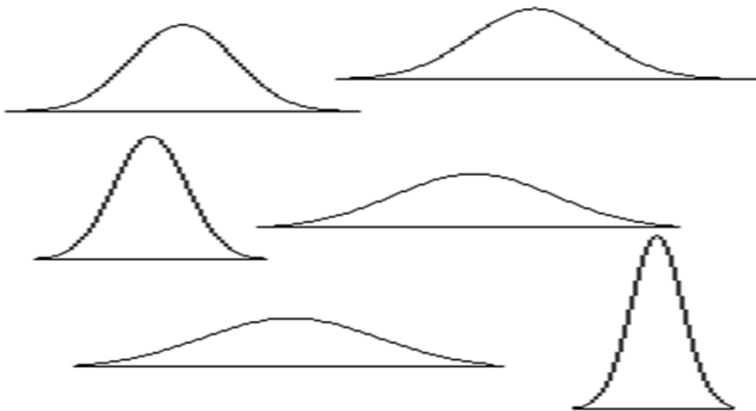
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- Example 3: $(2)(12) + 90 = 24 + 90 = 114$
- Example 4: $(-1)(12) + 90 = -12 + 90 = 78$

Normal Curves



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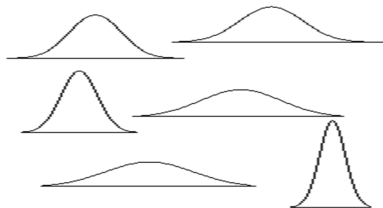
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- Dispersion can change.



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- Precisely defined by:
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- So, the shape of the Standard Normal Curve is always the same (i.e., standardized).

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- It is always a 'smooth' curve because as either a histogram or frequency polygon, the 'intervals' of the scale are infinitesimal.

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 - Symmetry; each extreme score in magnitude is as likely as one at the opposite end of the distribution.

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 - Assuming normality allows us to assign probabilities.

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 - If we find the assumption violated, there are often modern (robust) analysis alternatives.

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- The mean of that distribution (of means) will be **the** mean of the population.

http:

[//onlinestatbook.com/simulations/CLT/clt.html](http://onlinestatbook.com/simulations/CLT/clt.html)

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 - This has to do with probability, more on this later.

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 - If you know a particular Z-score, you can determine (often with a table) the percentage of scores above or below it.
 - If you know the percentages above and below, you can determine the Z-score.

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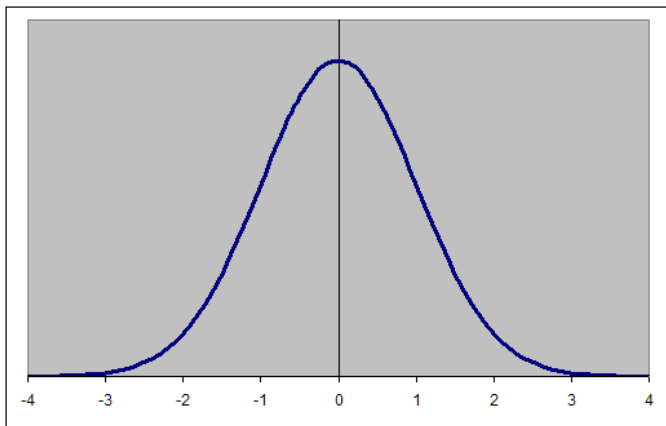
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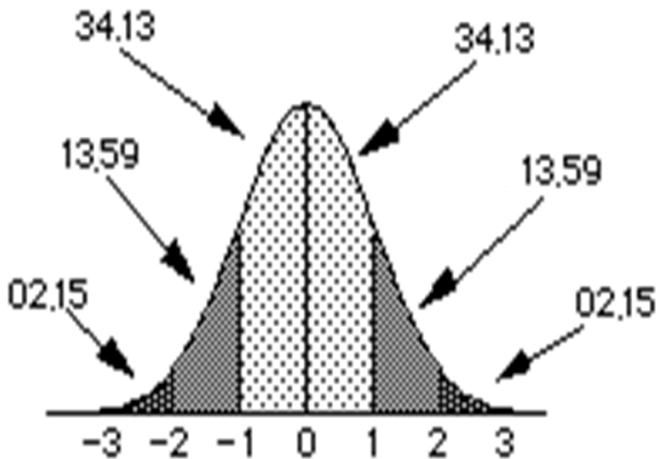
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- The score itself has no *area*, it is simply a boundary between, a marker if you prefer.
 - Like in geometry, point and line are identifiers.



Z-scores along the bottom, the percentage of scores between each can be figured mathematically but, we use tables instead:

<http://www.mathsisfun.com/data/standard-normal-distribution-table.html>



Z-scores along the bottom, percentages around the top
Memorize this image...burn it into your brain!!

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- The proportion of scores between any two Z-scores is the same as the probability of selecting another score between those two scores.
- Keep in mind, the complete table of z-scores and their associated probabilities is available at multiple sites online.

<http://www.mathsisfun.com/data/standard-normal-distribution-table.html>

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Knowing the formula for converting a raw score to a z-score and knowing the percentages of the standard normal curve (or at least having access to a table):

- We can calculate the percentage of scores between any two points of the normal distribution.
- For instance, between a raw score and the mean or between any two Z-scores.

More on Percentages.

- When figuring percentage of scores between any two z-scores or between the mean and a Z-score, pay close attention to the column in the table you need.

Z	% Mean to Z	% in Tail
.00	.00	50.00
.01	.40	49.60
.	.	.
.	.	.
.	.	.
.50	19.15	30.85
1.00	34.13	15.87

http://davidmlane.com/hyperstat/z_table.html

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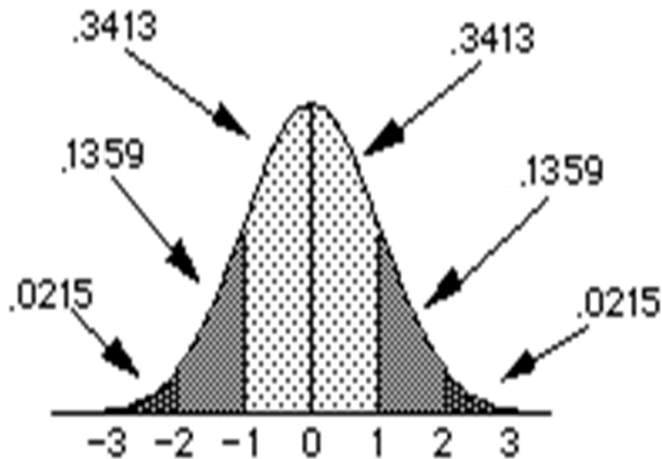
- Remember, when finding a percentile rank, the area under the normal curve becomes ranked (from 1 to 100 percentiles).
- For instance, we know that a Z-score of +2 represents 2 standard deviations above the mean. This same location, when converted to a percentile would be the 98th percentile.
 - 50% of the scores are below the mean, (+) 34% of the scores between the mean and a z-score of +1, (+) the other 14% of scores between a Z-score of +1 and a Z-score of +2 = 98% of the scores, or the 98th percentile.

Percentiles

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 - 2% of the scores are beyond 2 standard deviations below the mean, (+) 14% of the scores between 2 standard deviations below the mean and 1 standard deviation below the mean = 16% of the scores are below our Z-score of -1; a raw score with the Z-score of -1 is the 16th percentile.



Z-scores along the bottom, probabilities around the top
Memorize this image...burn it into your brain!!

Percentages & simple Probabilities

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- The proportion of scores between any two Z-scores is the same as the probability of selecting a case between those two Z-scores.

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 - The number of possible successful or desirable outcomes divided by the number of **all** possible outcomes.

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 - There is a 16.66% chance I'll roll a 5 on one independent roll of the die.

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 - $p = 0$
- Probability of getting a head on one toss of the coin after 2500 tosses of the coin?
 - $p = .50$
 - Remember the old saying: "The coin has no memory."

Practice Examples

- The 'Bear Cubs Problem'
- The 'Balls of Two Colors' problem

Available toward the bottom of the page at:

<http://www.cut-the-knot.org/probability.shtml>

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All of these topics will be re-visited consistently in future modules.

This concludes Module 4

Next time Module 5.

- Next time we'll begin covering Hypothesis Testing.
- Until next time; have a nice day.

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