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Section 4.2: Normal Distribution

Characteristics of a data set:

- mean
- median
- standard deviation
- five-number summary

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- median
- standard deviation
- five-number summary

These don't always tell you everything though.

Bar graph for scores of Math 109 quiz:





It's somewhat close to a bell-shaped curve.

Definition

We call that bell-shaped curve a normal curve.

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 A data set whose distribution is a normal curve has a normal distribution.

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- Note the outliers
- the spike at 15
- the whole picture is a little skewed to the right

Non-normal distributions:



- The first one has an inverted bell shape.
- The second one is called a bimodal distribution.

Lots of data sets do have normal or near-normal distributions though, so we're interested in them.



This is a normal distribution. (SAT scores)

Section 4.2.2: Properties of Normal Distributions

Normal distributions approximate real life distributions

Normal distributions approximate real life distributions Knowing the properties of normal curves, we can approximate the properties of near-normal data sets.



but they all have common properties.

Symmetry A normal curve is symmetric about the vertical line at the mean μ . This is called its **axis of symmetry**.

Concavity A normal curve has a concave up region on the left, a concave down region in the middle, and a concave up region on the right.

Inflection points The points where the curve changes concavity are called inflection points.

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 There are two inflection points, each one standard deviation from the center.

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- 68% of the data set is located under the concave down part of the curve. That's within σ of μ.
- 95% of the data points are within 2 σ 's of μ .
- 99.7% of the data points are within 3 σ's of μ.

Quartiles

$${\cal Q}1=\mu-$$
 0.675 σ and

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So 50% of data points are within 0.675 σ of μ .



For near-normal distributions, the above properties should be taken as approximations.

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score	4	5	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	25
freq.	1	1	2	2	3	5	9	12	11	13	9	8	7	5	3	2	1	1
cum fr.	1	2	4	6	9	14	23	35	46	59	68	76	83	88	91	93	94	95

We found earlier: • $\mu = 14.64$

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- *σ* = 3.35
- \blacktriangleright 62 out of 95 scores are within one σ from μ

$$\frac{62}{95} \cdot 100\% = 65.3\%$$

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$$\mu = 14.64$$

- ► *σ* = 3.35
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which is close to the ideal 68%

• number of scores within 2 σ 's of μ :

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- Finally, 93 out of 95 (i.e., 97.9%) points are within 3 σ's of μ.

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Finally, 93 out of 95 (i.e., 97.9%) points are within 3 σ's of μ.
 All of this is close to the ideal 68-95-99.7 principle.

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- $Q1 \approx \mu 0.675\sigma = 12.38$
- $\sim Q3 pprox \mu + 0.675 \sigma = 16.90$

(Faster than actually calculating them.)

Conclusion: If you know that the distribution of data points is normal, it is enough to know just 2 numbers to describe the data set completely:

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Conclusion: If you know that the distribution of data points is normal, it is enough to know just 2 numbers to describe the data set completely:

- Mean μ
- Standard deviation σ

Section 4.2.3: Using the 68-95-99.7 Principle

• Each tail outside one σ contains $\frac{100\% - 68\%}{2} = 16\%$





This means that a random point has:

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This means that a random point has: • 16% chance of being below $\mu - \sigma$ • 84% chance of being above $\mu - \sigma$ $pr(x < \mu - \sigma) = 0.16$ and $pr(x \ge \mu - \sigma) = 0.84.$

Similarly, each tail outside 2σ contains



Example

The weights of 6-month-old baby boys in the U.S. have a near-normal distribution with

•
$$\mu = 17.25$$
 lbs

•
$$\sigma = 2$$
 lbs

Example

Questions:

(a) What can we say about babies whose weight is 19.25 lb.?

Example

Questions:

- (a) What can we say about babies whose weight is 19.25 lb.?
- (b) If we pick a random baby, what's the probability he's less than 13.25 lb.?

(a) The baby whose weight is 19.25 lb.

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is certainly above the average
 Moreover, his weight

$$19.25 = 17.25 + 2 = \mu + \sigma$$

is precisely one standard deviation above the average.

Solution Thus • 84% of babies weigh less than him

Thus

84% of babies weigh less than him
16% of babies weigh more than him

We say that his weight is in the 84th percentile.

(b) A weight of 13.25 lb. or less places a baby at least 2 standard deviations below the average.

$13.25 = 17.25 - 2 \cdot 2 = \mu - 2\sigma$

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$$13.25 = 17.25 - 2 \cdot 2 = \mu - 2\sigma$$

The probability for that is 2.5%.

Next Time:

Section 4.3: Data Collection Section 4.3.1: Population v. Sample