## 1 <br> Today's plan:

- 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

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

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- it's close, but not super close to the normal curve
- 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
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- Knowing the properties of normal curves, we can approximate the properties of near-normal data sets.

Normal curves can vary:

but they all have common properties.

Symmetry A normal curve is symmetric about the vertical line at the mean $\mu$. 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.

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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 $\sigma$ of $\mu$.
- $95 \%$ of the data points are within $2 \sigma$ 's of $\mu$.
- $99.7 \%$ of the data points are within $3 \sigma$ 's of $\mu$.


## Quartiles

$Q 1=\mu-0.675 \sigma$ and

$$
Q 3=\mu+0.675 \sigma
$$

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$$
\text { Q3 }=\mu+0.675 \sigma
$$

So $50 \%$ of data points are within $0.675 \sigma$ of $\mu$.


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

## Example

## Recall Math 109 quiz:



| 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:

- $\mu=14.64$

| score | 4 | 5 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 | 17 | 18 | 19 | 20 | 21 | 22 | 25 |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
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We found earlier:

- $\mu=14.64$
- $\sigma=3.35$

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| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| 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:

- $\mu=14.64$
- $\sigma=3.35$
- 62 out of 95 scores are within one $\sigma$ from $\mu$
$\frac{62}{95} \cdot 100 \%=65.3 \%$

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| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| freq. | 1 | 1 | 2 | 2 | 3 | 5 | 9 | 12 | 11 | 13 | 9 | 8 | 7 | 5 | 3 | 2 | 1 | 1 |
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We found earlier:

- $\mu=14.64$
- $\sigma=3.35$
- 62 out of 95 scores are within one $\sigma$ from $\mu$

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

which is close to the ideal $68 \%$

- number of scores within $2 \sigma$ 's of $\mu$ :
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- number of scores within $2 \sigma$ 's of $\mu$ : $\mu-2 \sigma=7.94$ and $\mu+2 \sigma=21.34$, so 91 out of 95 (i.e., 95.8\%)
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- number of scores within $2 \sigma$ 's of $\mu$ :
$\mu-2 \sigma=7.94$ and $\mu+2 \sigma=21.34$,
so 91 out of 95 (i.e., $95.8 \%$ ), close to the ideal $95 \%$.
- Finally, 93 out of 95 (i.e., 97.9\%) points are within $3 \sigma$ 's of $\mu$.
- number of scores within $2 \sigma$ 's of $\mu$ :
$\mu-2 \sigma=7.94$ and $\mu+2 \sigma=21.34$,
so 91 out of 95 (i.e., $95.8 \%$ ), close to the ideal $95 \%$.
- Finally, 93 out of 95 (i.e., 97.9\%) points are within $3 \sigma$ 's of $\mu$. All of this is close to the ideal 68-95-99.7 principle.

We can also estimate the quartiles for the Math 109 quizzes:

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- Q1 $\approx \mu-0.675 \sigma=12.38$
- Q3 $\approx \mu+0.675 \sigma=16.90$
(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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- Mean $\mu$
- Standard deviation $\sigma$


# Section 4.2.3: Using the 68-95-99.7 Principle 

- Each tail outside one $\sigma$ contains

$$
\frac{100 \%-68 \%}{2}=16 \%
$$



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- $16 \%$ chance of being below $\mu-\sigma$
- $84 \%$ chance of being above $\mu-\sigma$

This means that a random point has:

- $16 \%$ chance of being below $\mu-\sigma$
$-84 \%$ chance of being above $\mu-\sigma$

$$
\begin{aligned}
& \operatorname{pr}(x<\mu-\sigma)=0.16 \text { and } \\
& \quad \operatorname{pr}(x \geq \mu-\sigma)=0.84 .
\end{aligned}
$$

## Similarly, each tail outside $2 \sigma$ contains

$$
\begin{gathered}
\frac{100 \%-95 \%}{2}=2.5 \% \\
\frac{\downarrow_{\mu-2 \sigma}}{2.5 \%} \\
\hline
\end{gathered}
$$

## Example

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

- $\mu=17.25 \mathrm{lbs}$
- $\sigma=2$ lbs


## Example

Questions:

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

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- (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 .?

Solution
(a) The baby whose weight is 19.25 lb.

## Solution

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(a) The baby whose weight is 19.25 lb.

- 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

Solution
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.

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

$$
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$$

The probability for that is $2.5 \%$.

Next Time:

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

