

The Normal Distribution!

STAT 120

Day 15

- We've covered the core ideas of intro stats:
 - EDA: using pictures and numbers to make sense of data
 - Estimation: estimating unknown parameters with confidence
 - Testing: assessing hypotheses with p-values
- Rest of the course covers more types of inference methods
 - Instead of using computer simulations to generate bootstrap/randomization distributions
 - Use probability models to do this

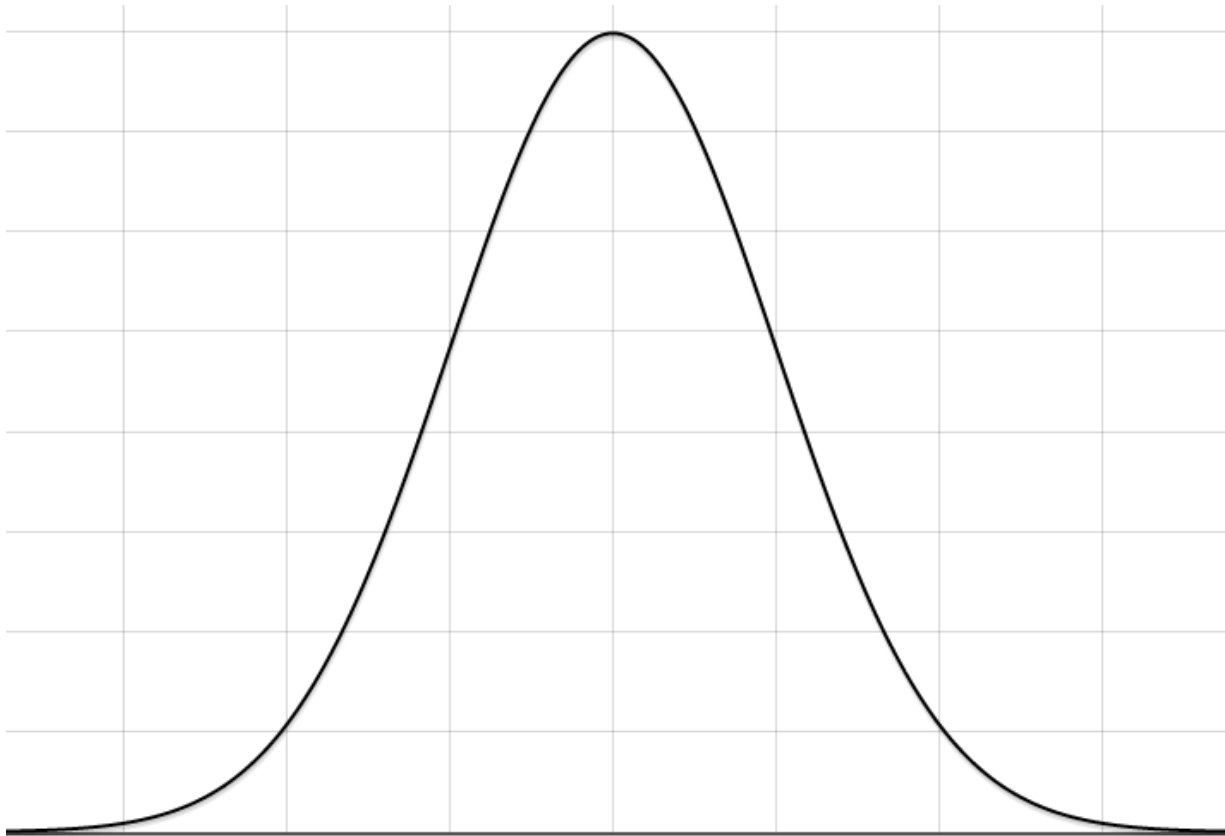
Density Curve

A ***density curve*** is a theoretical model to describe a distribution.

- Distribution for
 - **Individual** measurements in **population** (for a quantitative variable)
 - **Sampling distribution** for a **statistic**
- All density curves:
 - have an area under the curve of 1 (100%)
 - **give proportions/percents as areas under the curve**

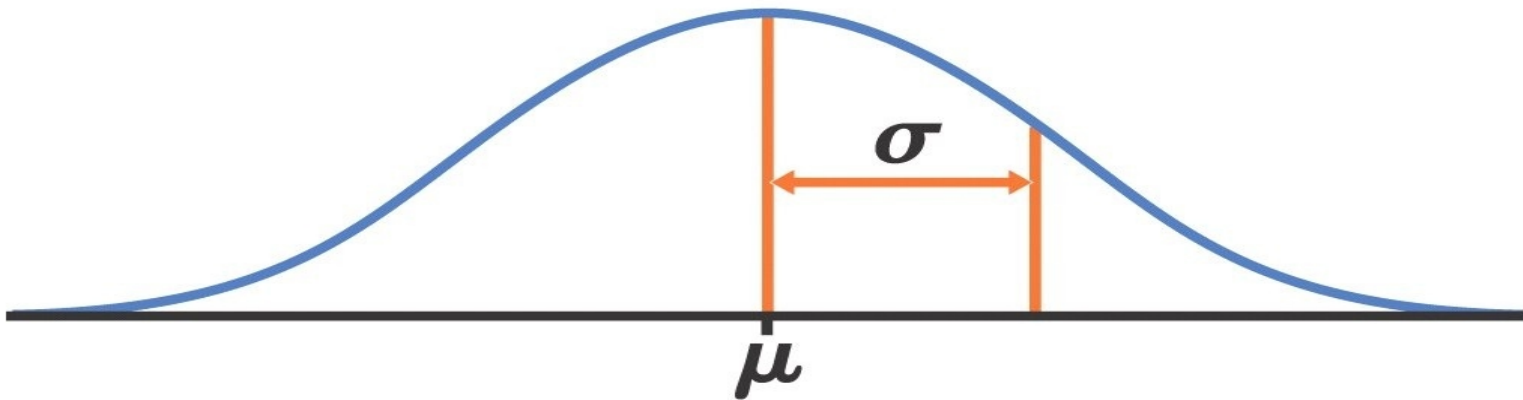
Normal Distribution

A ***normal distribution*** has a symmetric bell-shaped density curve.



The Normal Model: $X \sim N(\mu, \sigma)$

- The mean and SD determine how a normal density curve looks.
- The normal model **parameters** are
 - μ = model **mean (center)**
 - σ = model **SD (variability)**

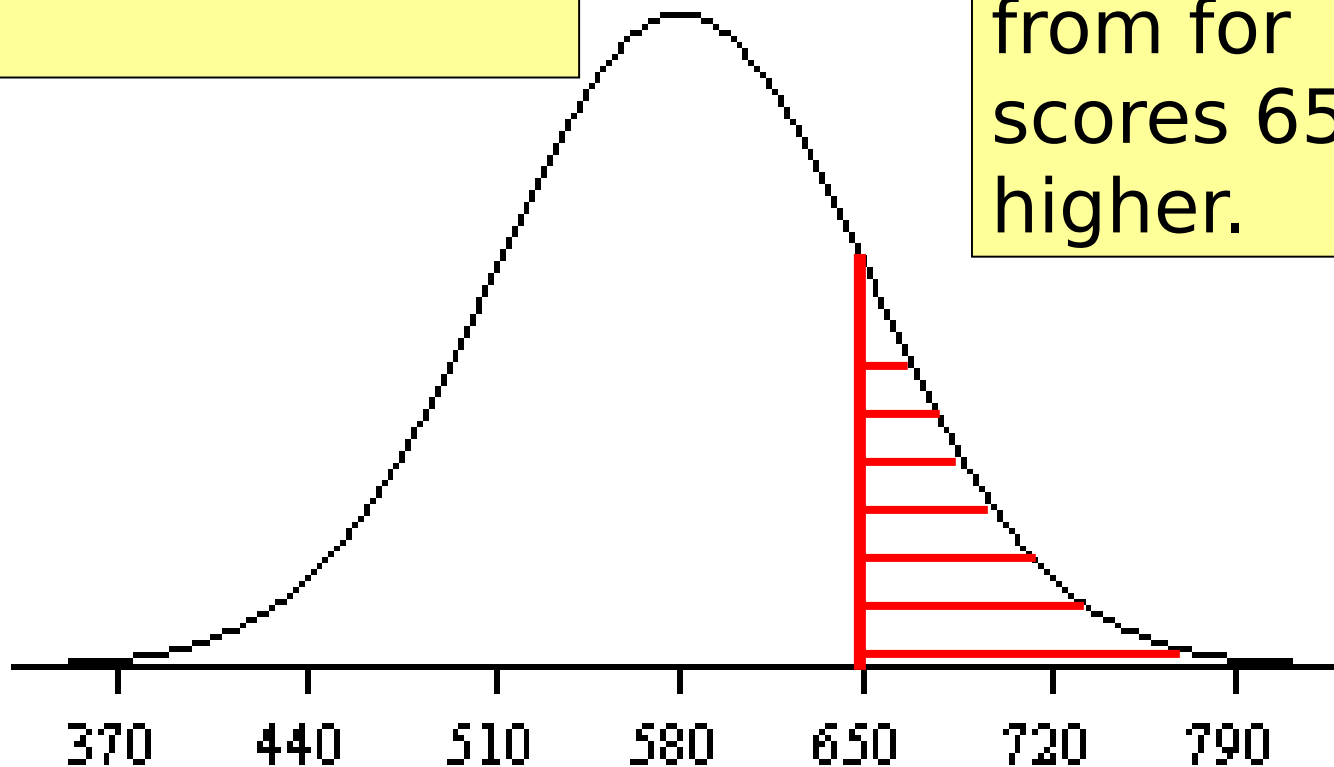


Example: A Population

Verbal SAT $\sim N(580, 70)$

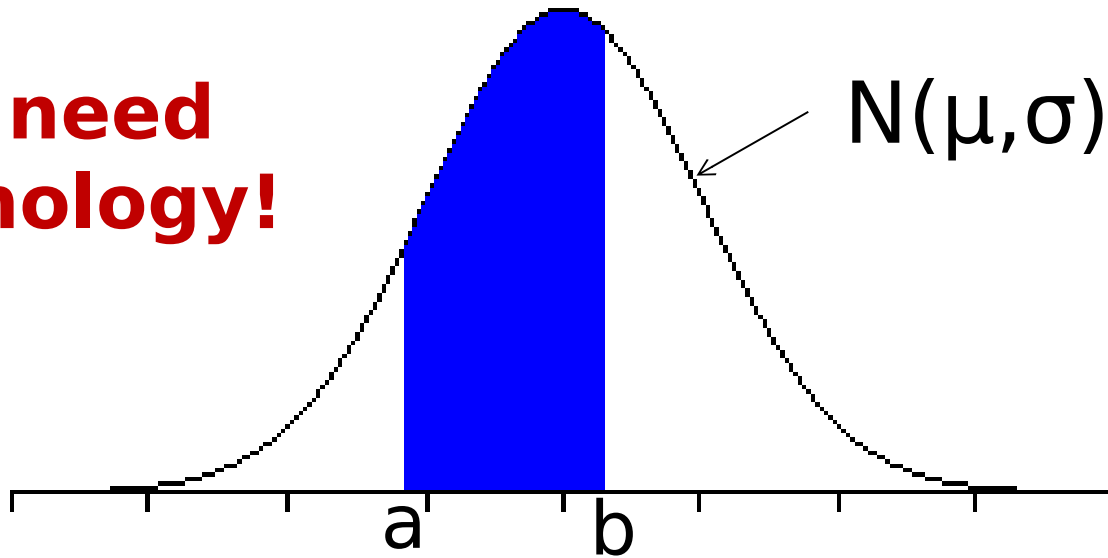
What proportion of people score above 650?

Area under the curve from for scores 650 & higher.



How can we find areas under a normal density?

We need technology!

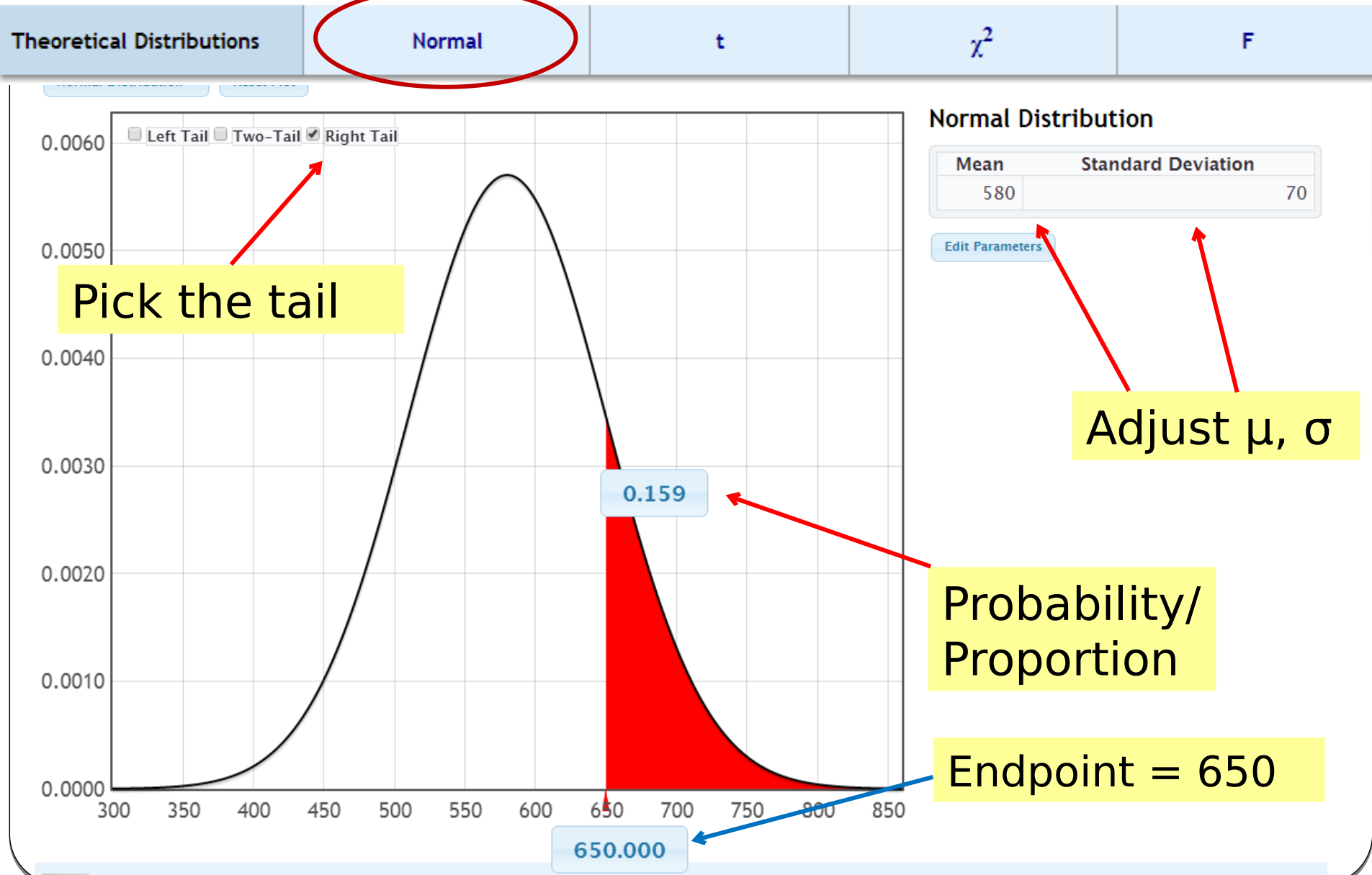


ARGH!

Calculus!

$$Area = \int_a^b \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

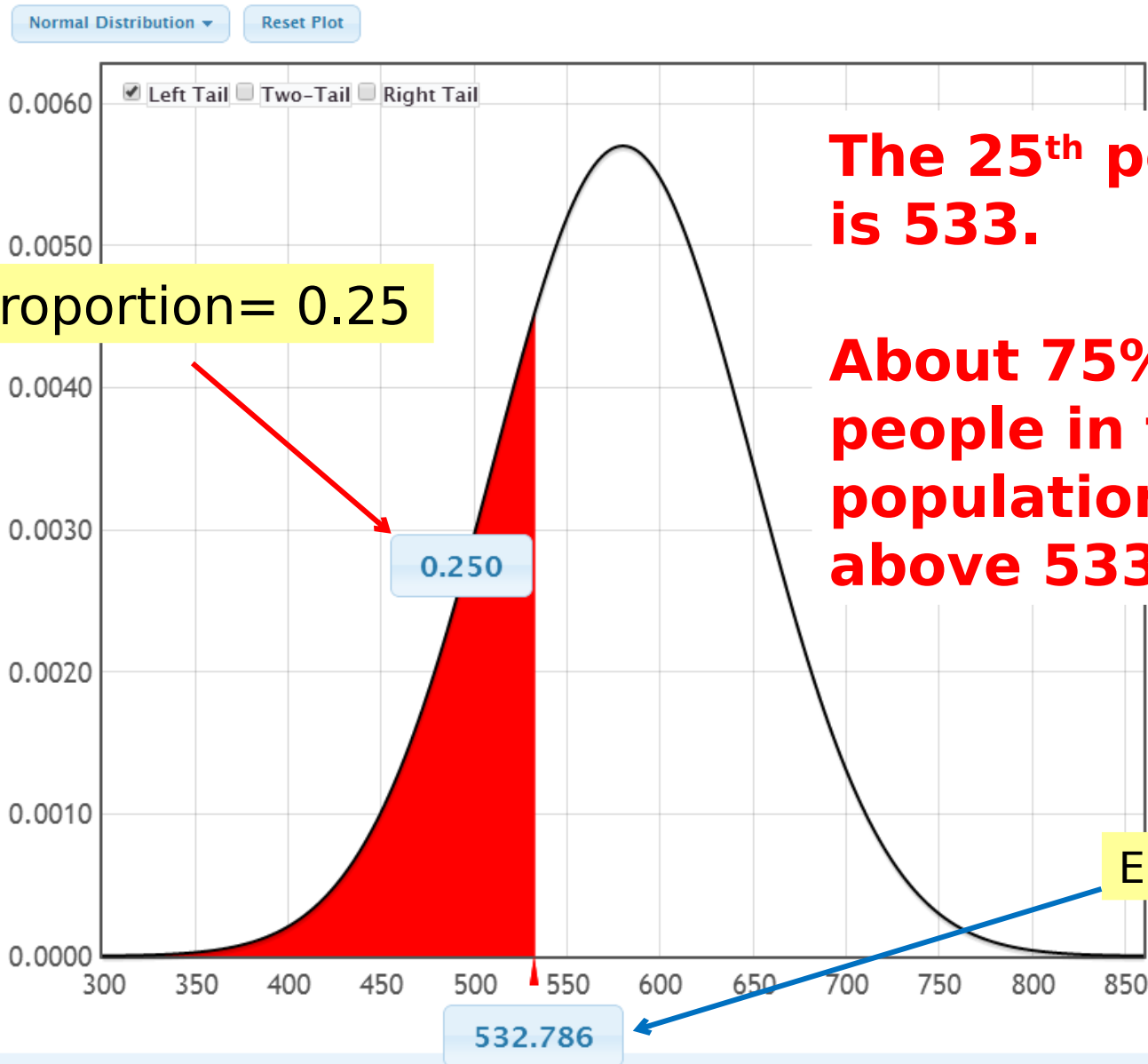
(1a) StatKey – Verbal SAT population



Example: Verbal SAT scores

- About 16% of people in this population had a score of 650 or higher.
- What score is the 25th percentile?
- without Statkey – it will be a score below 580 (median & mean) and above 440 (2 SD below 580)
- using Statkey – adjust the left-tail area to be 0.25.

(1b) StatKey - Verbal SAT population



The 25th percentile is 533.

About 75% of people in the population score above 533.

Endpoint = 533

Example: Verbal SAT scores

- What percent of the population had a score of 650 or higher?
- Using R enter:
`> 1 - pnorm(650, 580, 70)`

What score is the 25th percentile?

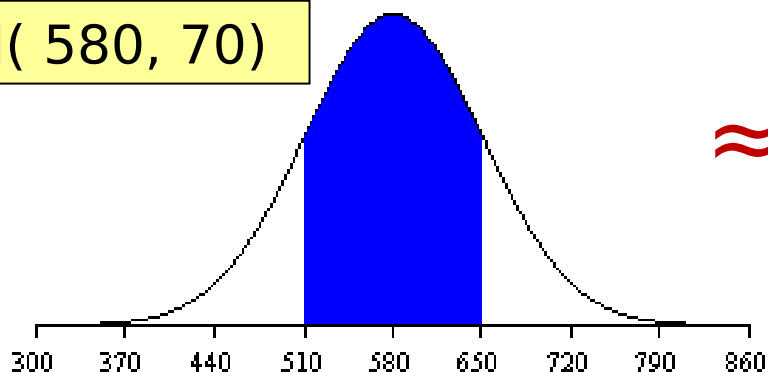
- Using R enter:
`> qnorm(.25, 580, 70)`

Finding Probabilities for $N(\mu, \sigma)$

Big Idea for Normal Models: **All that really matters is the number of standard deviations from the mean.**

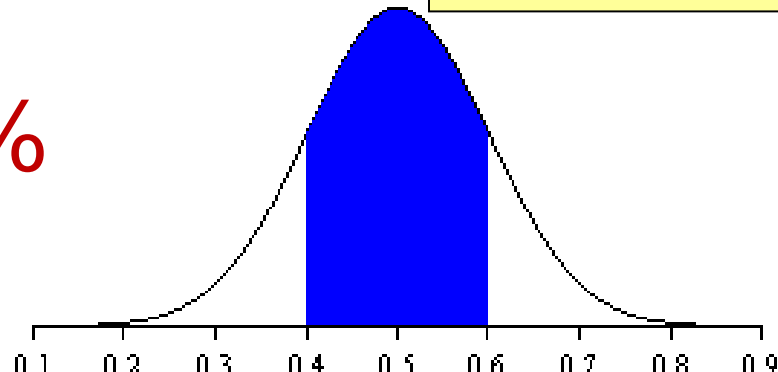
About what proportion should be within **one** standard deviation of the mean?

$N(580, 70)$

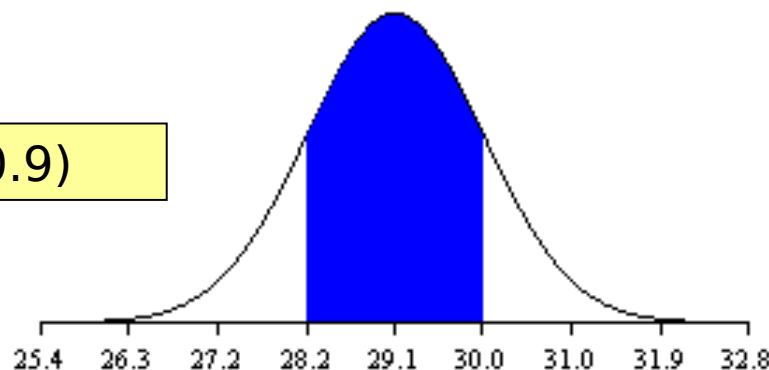


$\approx 68\%$

$N(0.5, 0.1)$



$N(29.1, 0.9)$



Big Idea for Normal Models:

All we need is a z-score.

Standard Normal

$$\mu=0, \sigma=1 \rightarrow Z \sim N(0,1)$$

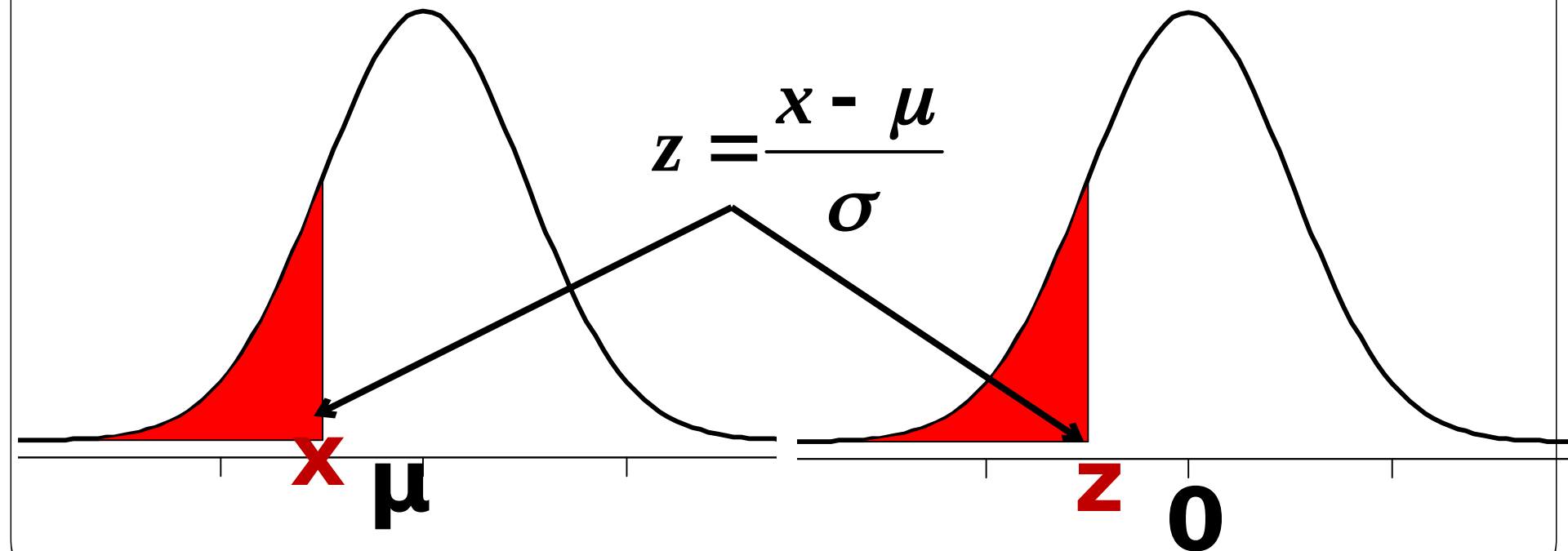
Connecting any Normal model to the standard normal model

Area below x = Area below z

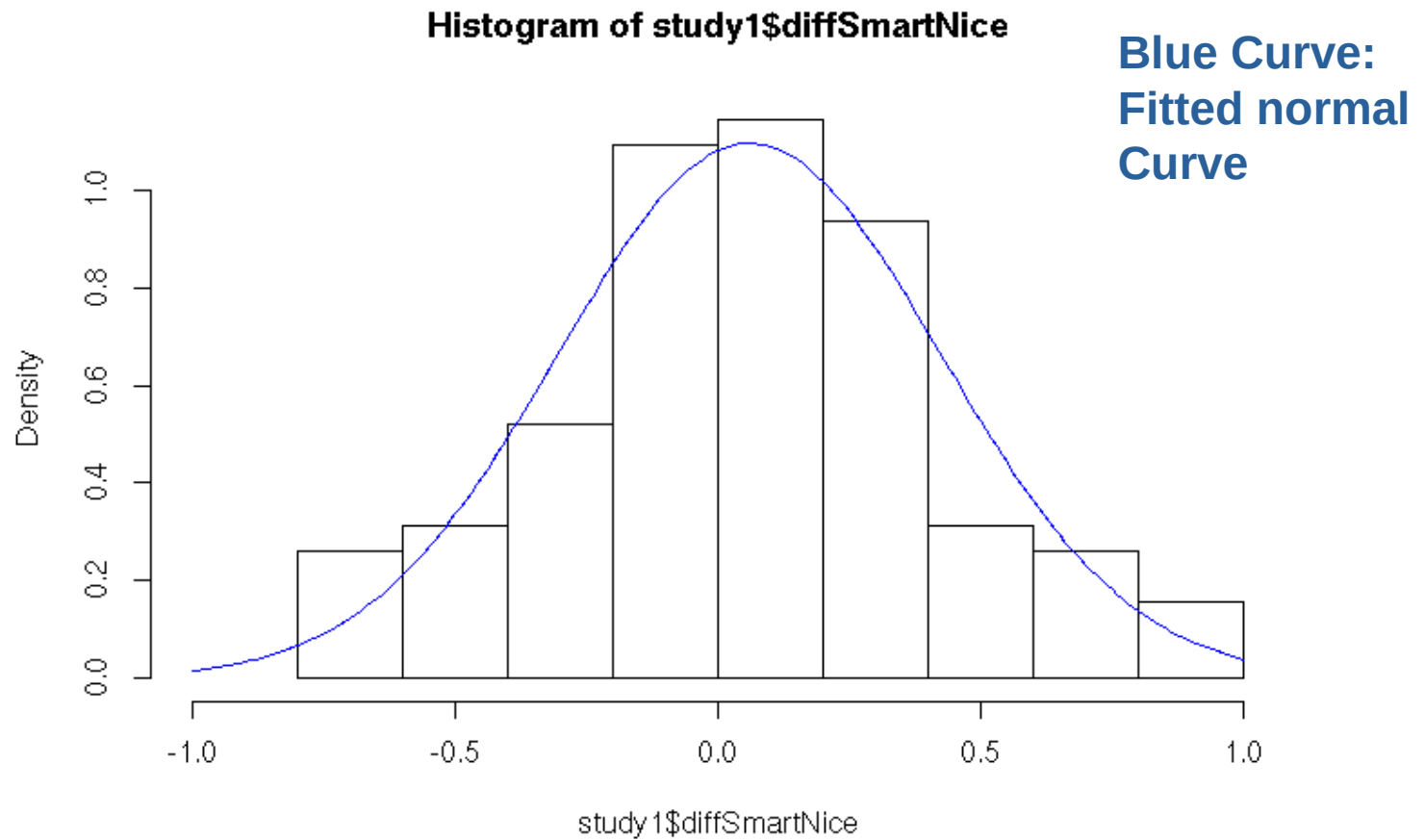
X is $N(\mu, \sigma)$

Z is $N(0, 1)$

$$z = \frac{x - \mu}{\sigma}$$

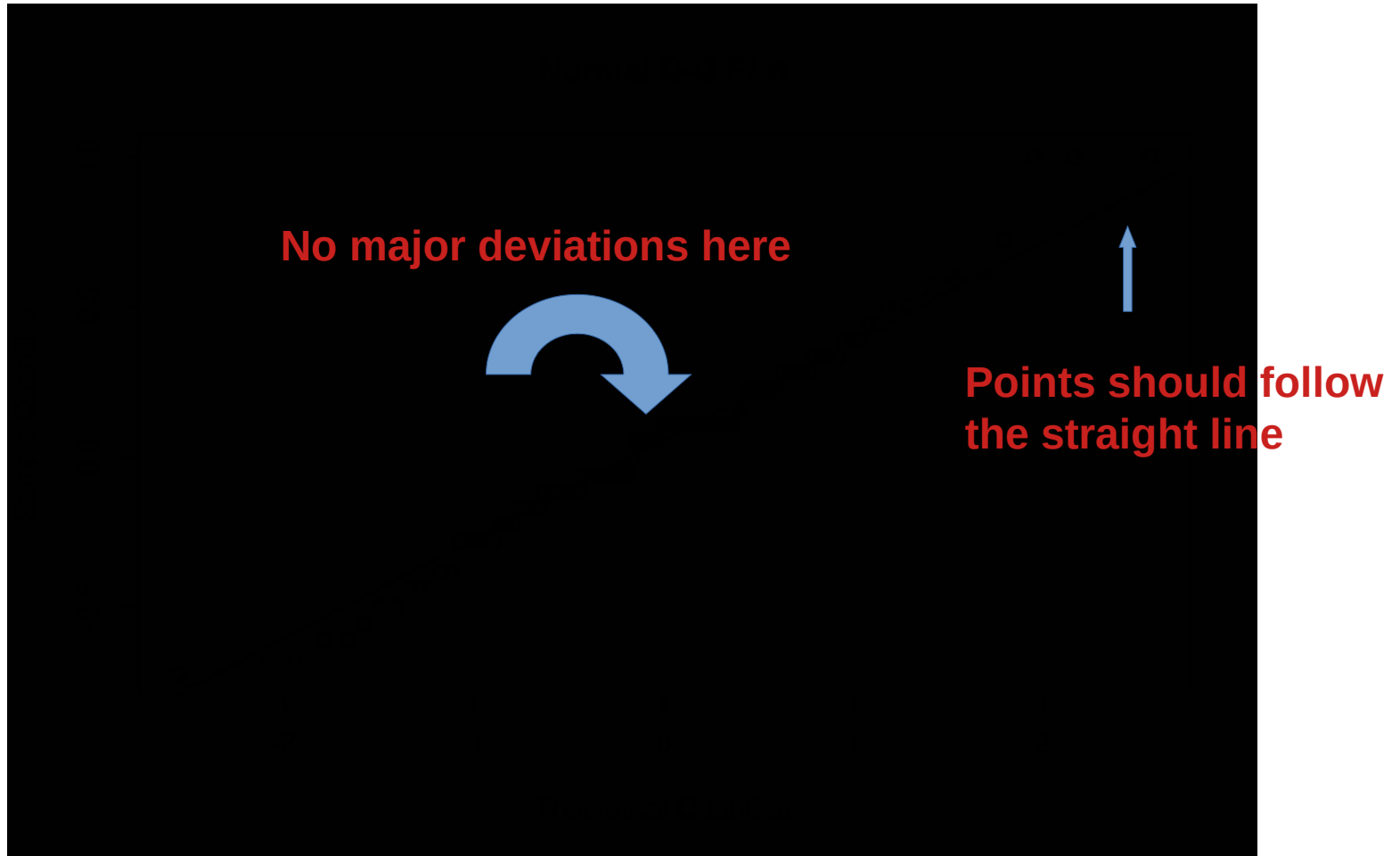


Normal Approximation: Sleep Study 1



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> dnorm(x-values, mean, sd)
```

QQ-plot: Sleep Study 1



The difference between nice and smart scores follows a normal distribution.

Big question

- When have we already been using normal models??
 - Bootstrap distributions – get confidence intervals if a bootstrap distribution is roughly bell-shaped
 - Randomization distributions – many of these are bell-shaped.
- Normal models play a huge role in statistical inference.
- **If we know the (bootstrap/randomization) standard error*** then we can just use a normal model rather than a resampling model (which requires more computational effort).