

How was spring break?

1 - I slept the whole time

5 - I worked the whole time

10 - I had parties the whole time

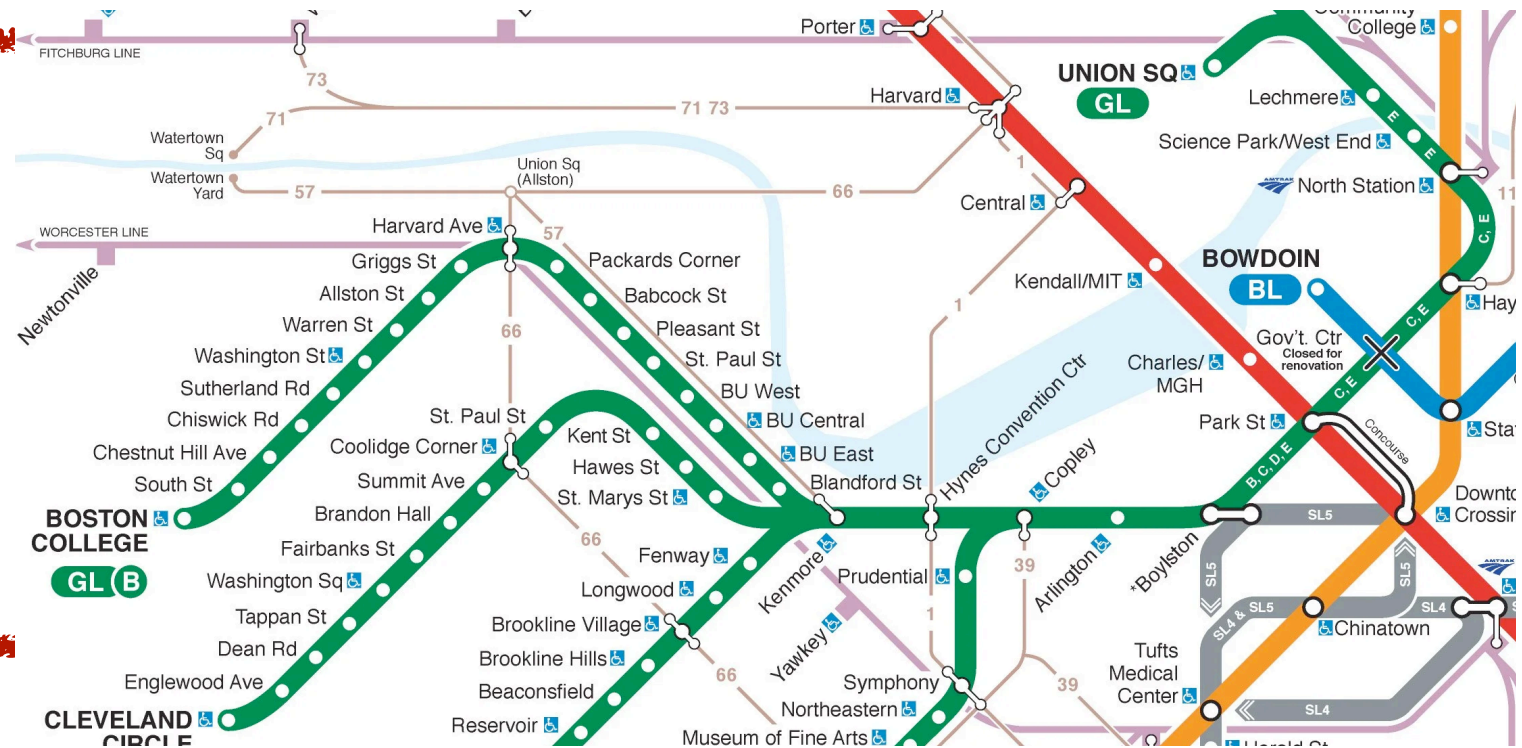


Normal distributions, central limit theorem, cumulative distribution function

GL Green Line Service Changes

Beginning Monday, March 21, Green Line service between **North Station** and **Lechmere** will resume, and regular service on the Union Branch will begin.

- B and C Branch trains will terminate at Government Center
- D Branch trains will terminate at North Station
- E Branch trains will terminate at Union Square



Distributions and probability mass functions

- "All distributions have a **probability mass function**. This tells us how the mass is distributed across outcomes."

- For a binomial distribution, the shape of this function depends on p and $n \rightarrow \# \text{ of trials}$

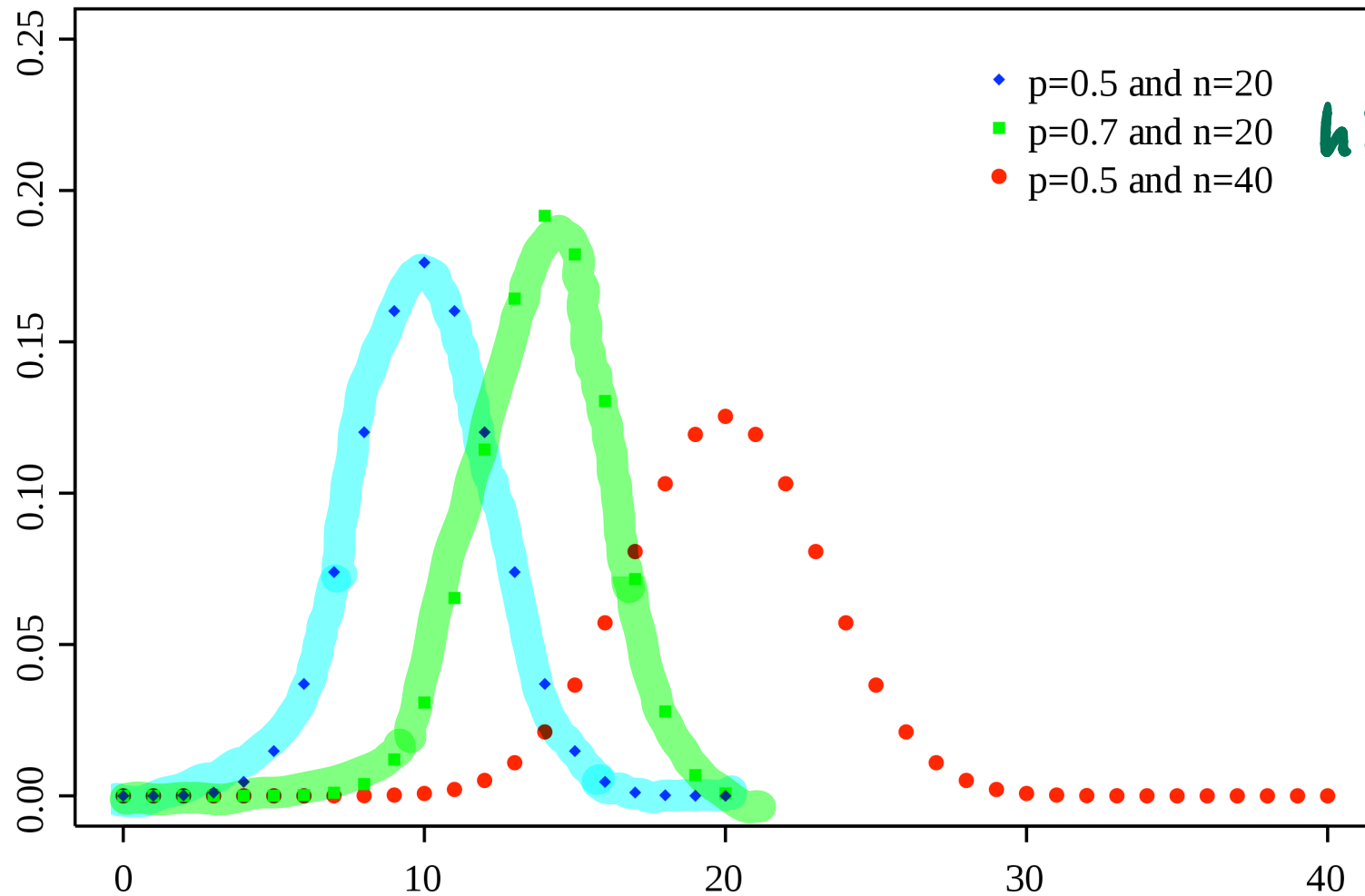
- For a poisson distribution, the shape of this function depends on $\lambda \rightarrow \text{rate of occurrence}$

p
↓
likelihood
of success

Distributions and probability mass functions

- For a binomial distribution, the shape of this function depends on

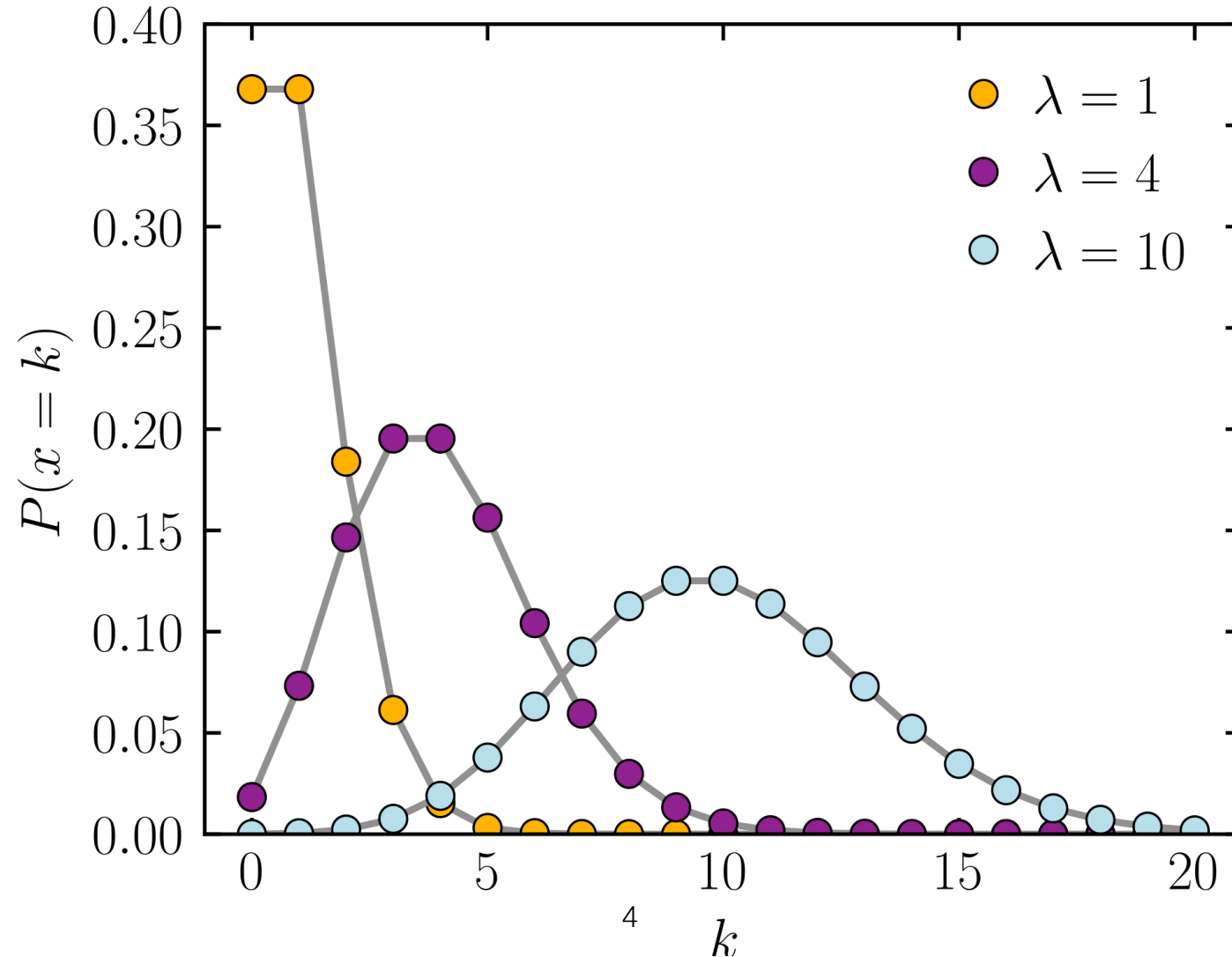
P and n



higher P
value

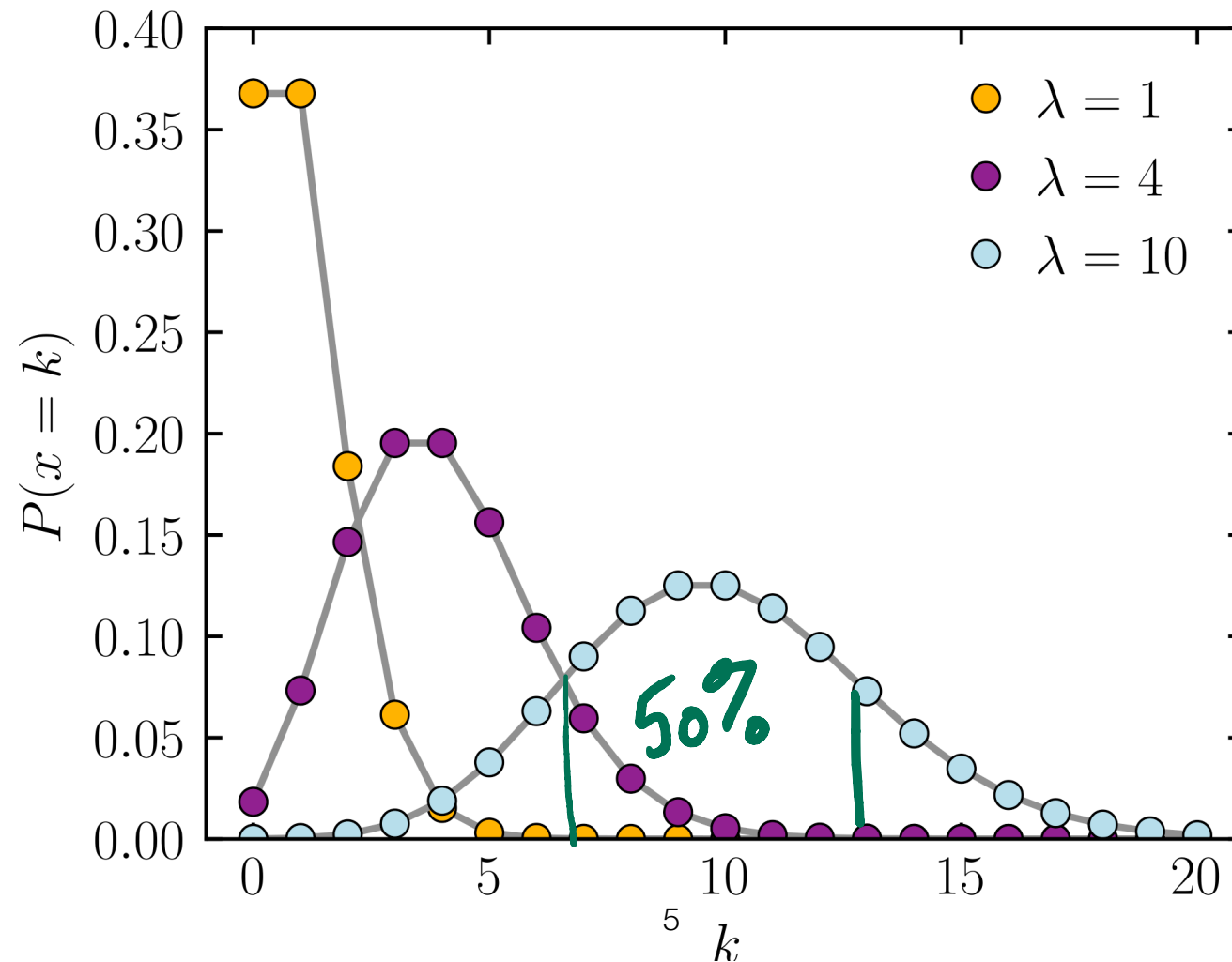
Distributions and probability mass functions

- For a poisson distribution, the shape of this function depends on λ



Distributions and probability mass functions

- The area under the curve for a probability mass function sums to: one
- This means that one of the outcomes will happen



Distributions and probability mass functions

- We want to pick the best probability distribution for the events that we're observing to create the best model/the model that is closest to the ground truth.

- Binomial distributions: discrete r.v.s - only success/failure, # of successes in a given # of trials

- Poisson distributions: discrete r.v.s. - only happened/didn't happen, # of occurrences in a time span

• both of these are for discrete variables
↳ $B(p, 1 \dots 3)$ to sum to one

A. binomial

↳ new / not new

↳ newness of trains is independent

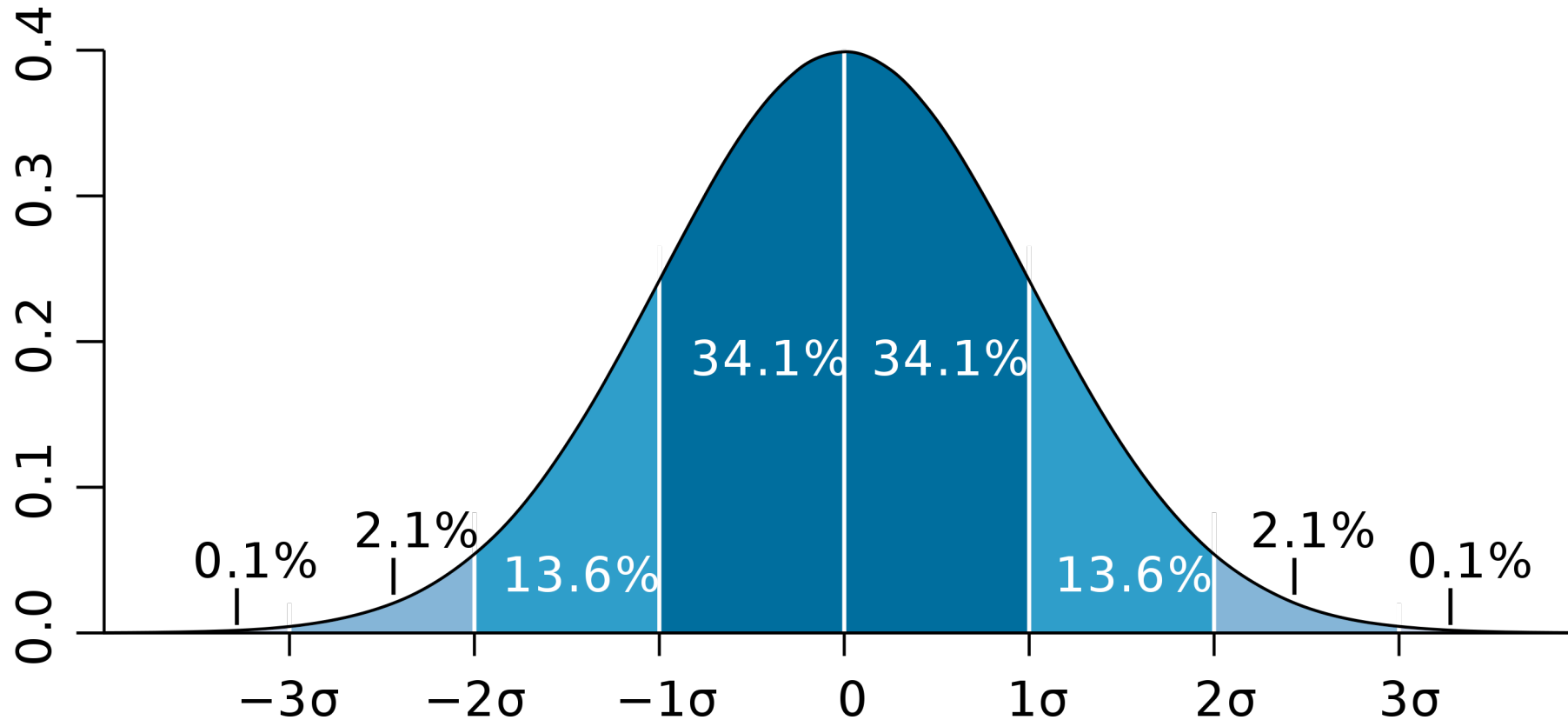
for all train examples, ^{this lecture} ↓ assume that
train newness / travel time etc are
independent

Normal Distributions

- A **normal (or Gaussian)** distribution is a bell-shaped curve.

std dev *variance*

- Previously, we used it to ground what σ means (and σ^2)



Normal Distributions

- A **normal** distribution is a bell-shaped curve that models a random variable that has a default mean, an expected variation about that mean, and whose values are real-valued

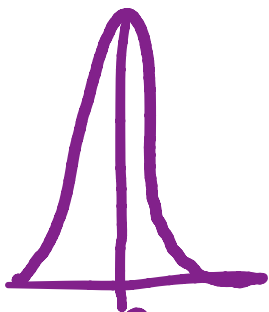
↳ σ

- in contrast to discrete

- heights

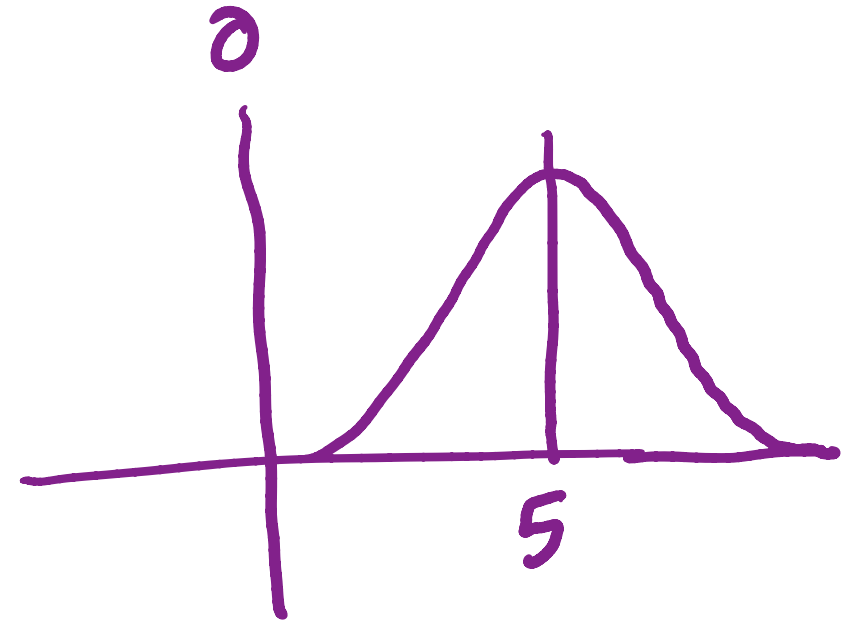
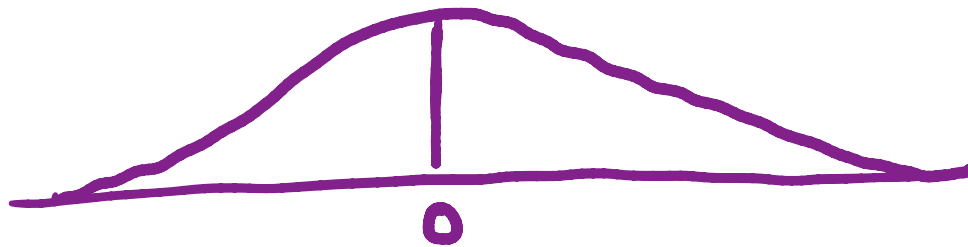
- times

small σ



small σ

large σ



Normal Distributions

- A **normal** distribution is a bell-shaped curve that models a **real-valued random variable**.

↗ vs. mass S

- It is defined by the **probability density function** $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$

$f(5)$

- Where μ is the mean (expected value) and σ is the standard deviation of the random variable

- If we say that $X \sim N(5,4)$, this means:

X is a r.v. w/ mean 5 and variance 4

ICA Question 2: Normal distributions and pdf

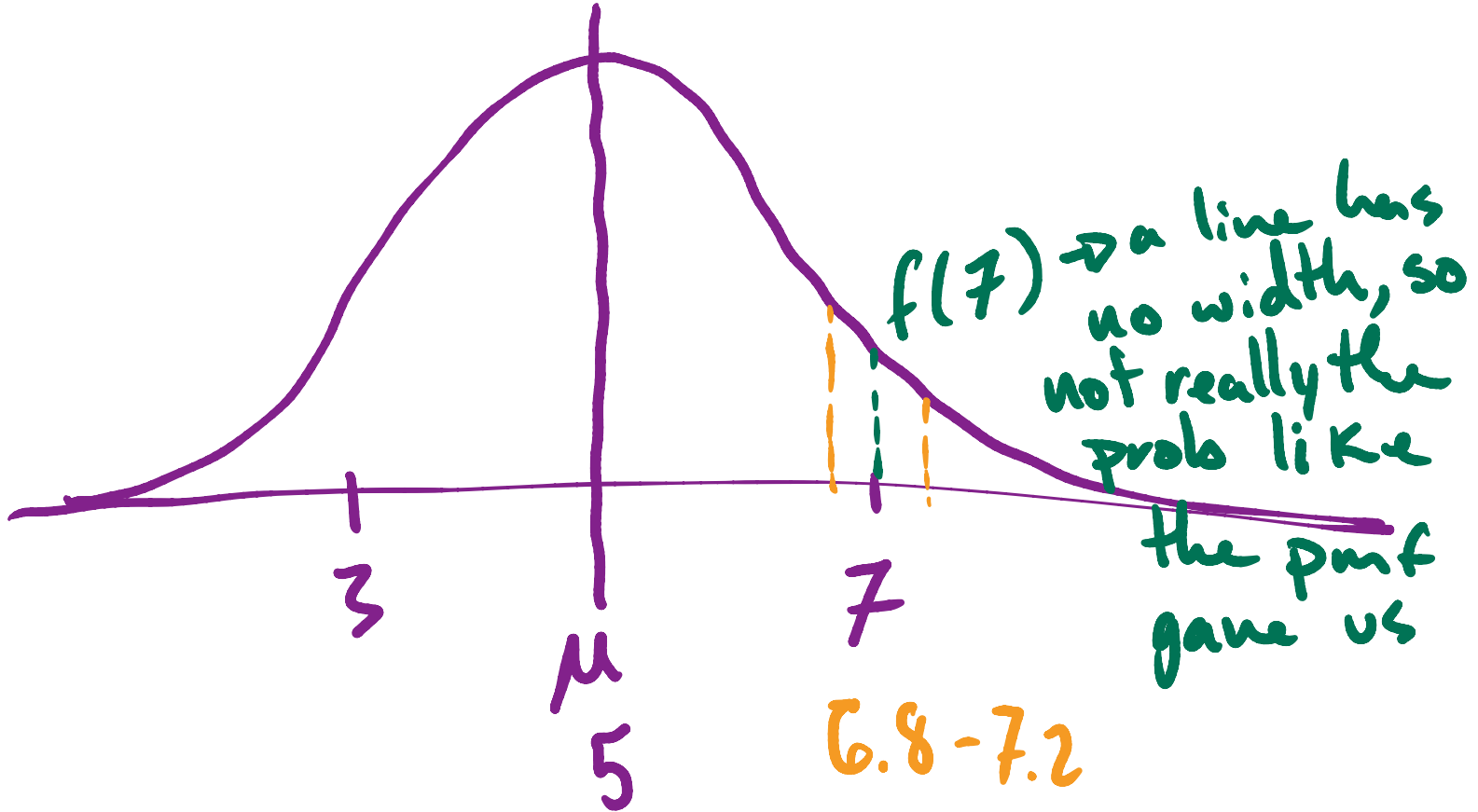
The times between North Station and Union Square have a mean of 5 minutes and a variance of 4 minutes.

What is $f(5)$ if $X \sim N(5,4)$ and $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$?

yes, this is all in the exponent!

0.1997

PdFs (real-valued r.v.)



Central Limit Theorem

- The **central limit theorem** says that if a population has a mean μ and a standard deviation σ , if we take enough samples, with replacement, the samples' means will be normally distributed.

- Requirements:

- a coin flip - heights
- die rolls

- observations must be independent from/dependent on one another
- the mean and the variance must be defined and finite
- observed random variables must/don't have to be normally distributed

• sample size? sufficiently big enough

ICA Question 3: central limit theorem

```
# for coin flipping
import random
# for graphing
import matplotlib.pyplot as plt

# central limit theorem
def flip_coin(times):
    return [random.randint(0, 1) for t in range(times)]

samples = YOUR NUMBER HERE
times = YOUR OTHER NUMBER HERE
averages = []
for sample_num in range(samples):
    coin_flips = flip_coin(times)
    averages.append(sum(coin_flips) / len(coin_flips))

plt.hist(averages)
plt.show()
```

Cumulative Distribution Functions

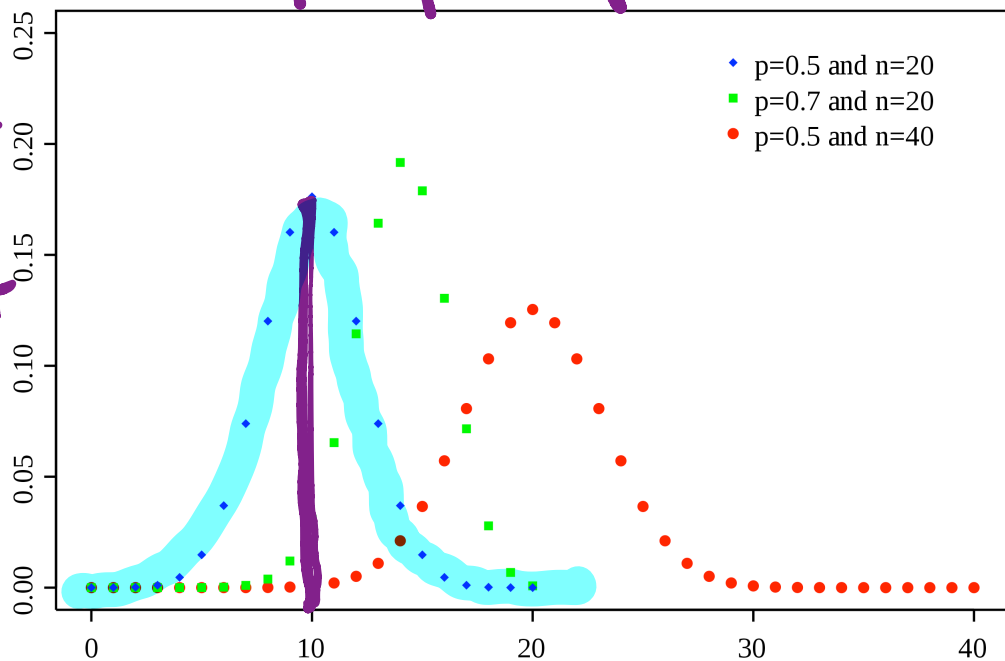
- A probability mass function or a probability density function tells us "what is the probability that a random variable will take this value according to the underlying distribution"

$f(10, \dots)$

- A **cumulative distribution function** tells us the probability that a random variable will take a value less than or equal to a target value \rightarrow **x-axis**

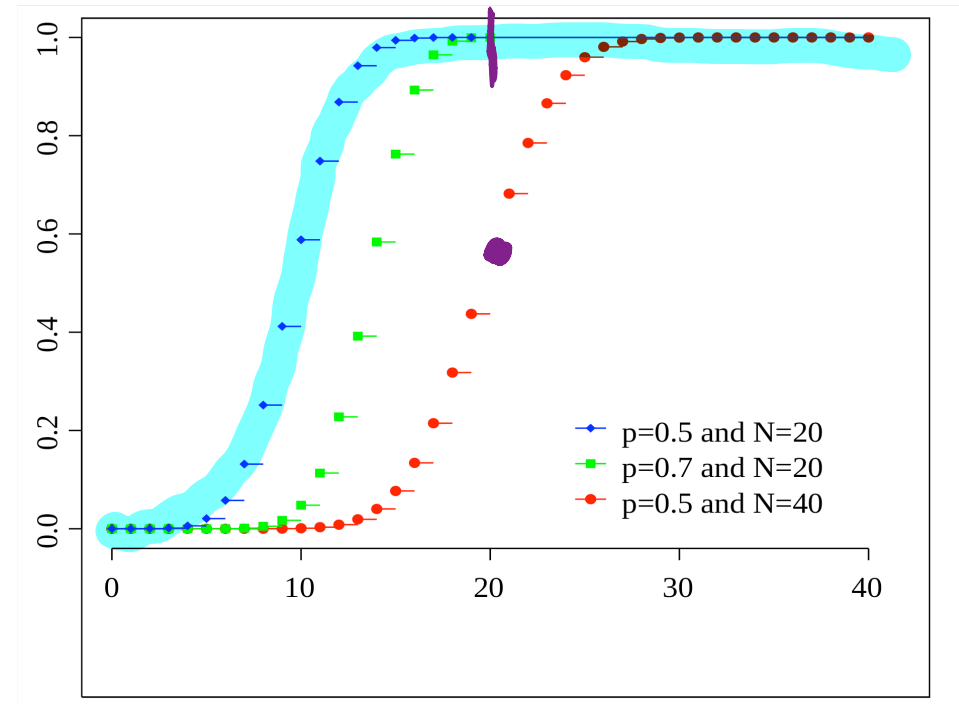
binomial - pmf

width: 1
times
height



↓

0



Cumulative Distribution Functions

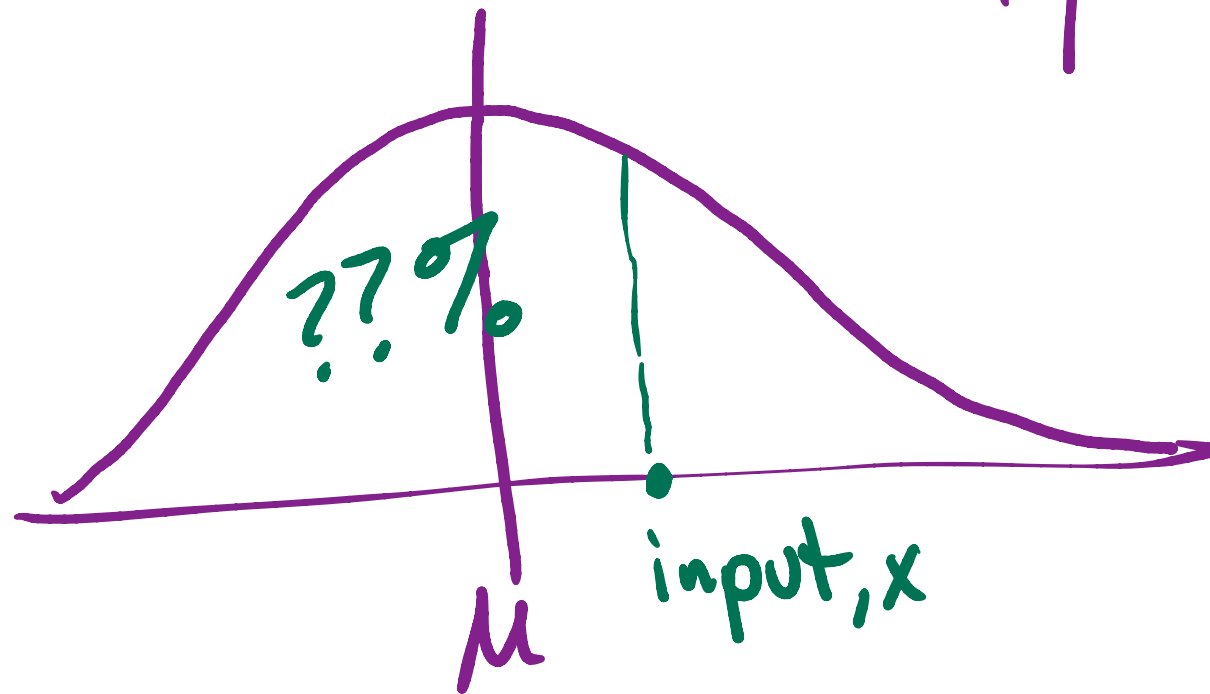
- A **cumulative distribution function** tells us the probability that a random variable will take a value less than or equal to a target value.

- Input: a real-value

5.5

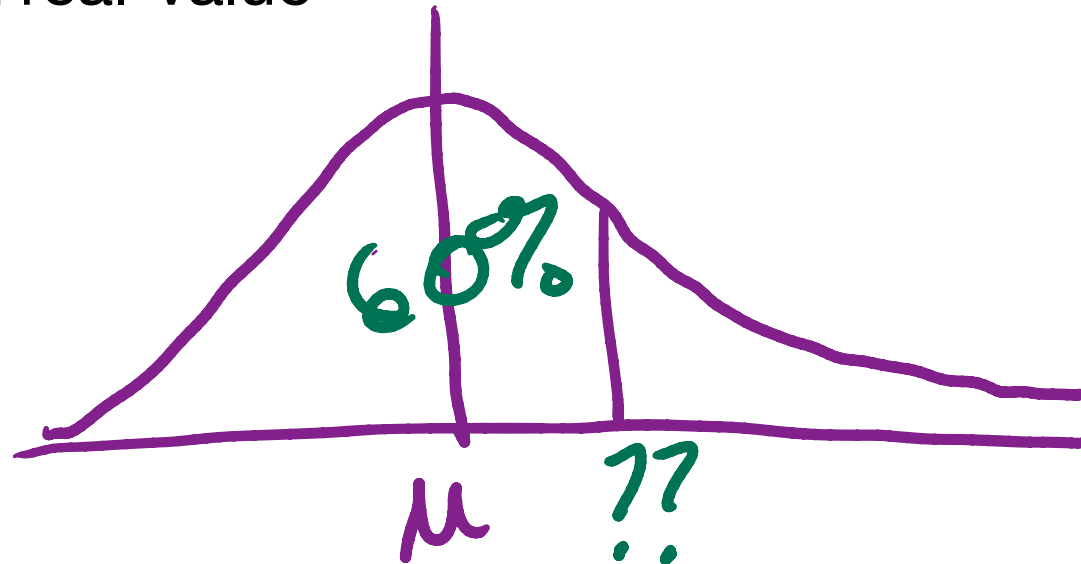
- Output: a percentage chance

probability



Percent Point Functions

- A **percent point function** tells us the value of x for which some percentage of the normal distribution is at or under that value.
- Input: a percentage chance
- Output: a real-value



ICA Question 4: cdf/ppf

The times between North Station and Union Square have a mean of 5 minutes and a variance of 4 minutes. If $X \sim N(5,4)$, then, using scipy, answer the following questions:

- 1) If I know that I'm very lucky and I expect my travel time to be in the **bottom** 10% of times, how long should I budget for my trip?

$$\hookrightarrow \text{ppf for } .1 \rightarrow 2.44$$

- 2) What percentage of trains can I expect to take 5 - 8 minutes (inclusive) for this trip?

$$\hookrightarrow \text{cdf}(8) - \text{cdf}(5) \rightarrow .43 \rightarrow 43\%$$

Mini-project clarifications

- (There's a pinned piazza post with these as well) → will be this afternoon
- submission format? length? no constraints on format/length.
- examples? unfortunately none :(
- thoroughness of the application of math topics? scale this appropriately. If you are working individually, expect to spend about 25 minutes actually writing down the explanation/grounding for each math topic.
- how much time? approx 200 minutes outside of class time — we are assuming that you may need to review the topics from class but that you do not need to learn them from scratch in this estimate

Mini-project clarifications

- what kind of scenarios to consider? up to you! pick ones that demonstrate the math topics that you've chosen well
- is it necessary to code the demonstrations or is it okay to just explain the concept? it is not necessary to code the demonstrations
- what can be included in the research? math topics? code implementation? calculations? yes. Anything you need to do that is not creating your actual deliverable
- is it necessary to collect actual data? no, but fabricated data should be reasonable

Mini-project clarifications

- More questions? post on the pinned piazza post and/or come to Felix's office hours!

Schedule

ICA passcode: "green"

Turn in ICA 16 on Canvas (make sure that this is submitted by 2pm!)

HW 6's final due date is on Tuesday. No late day deductions for this HW!

HW 7 will be released on Thursday, it is due on April 3rd. You'll need content from Thursday's/next Monday's lecture for this HW.

Test 3 (your last in-class test (**Test 4** is during your final exam slot)) is the Thurs. after this one

Mon	Tue	Wed	Thu	Fri	Sat	Sun
March 21st Lecture 16 - normal distributions	Felix OH Calendly HW 6 due @ 11:59pm	Felix OH Calendly	Felix OH Calendly Lecture 17 - hypothesis testing			
March 28th Lecture 18 - t-tests, experimental bias	Felix OH Calendly	Felix OH Calendly	Felix OH Calendly Test 3 (HW 5/6)			HW 7 due @ 11:59pm

More recommended resources on these topics

- Probability density functions: YouTube, 3Blue1Brown -- Why “probability of 0” does not mean “impossible” | Probabilities of probabilities, part 2
- why approximating a normal CDF is hard: Wikipedia, https://en.wikipedia.org/wiki/Normal_distribution#Numerical_approximations_for_the_normal_CDF_and_normal_quantile_function
- Central Limit Theorem: YouTube, Central limit theorem | Inferential statistics | Probability and Statistics | Khan Academy