

# TESTS & DESCRIPTIVE STATISTICS

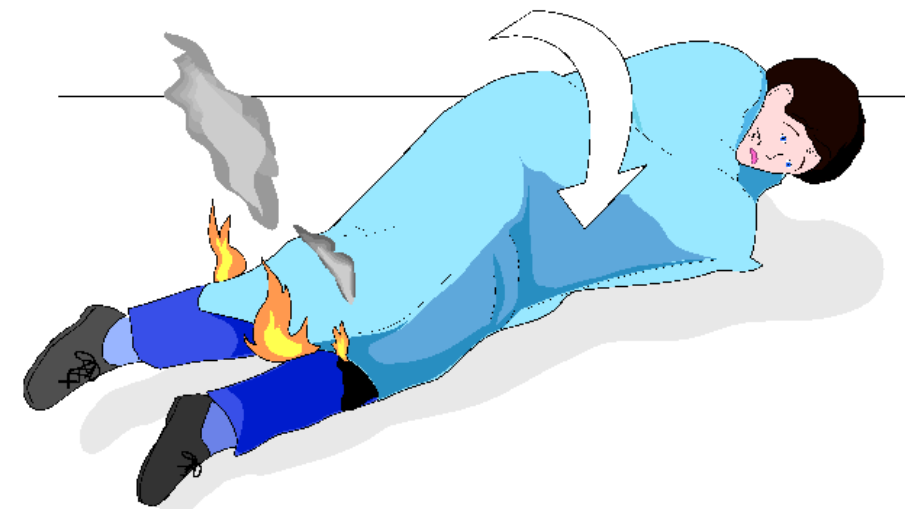
10.2.2020

# PROBLEM SET 2

\* was due TODAY!

# PROBLEM SET 3

\* is assigned TODAY! (due in 2 weeks)



# RECAP

- \* bernoulli distribution
  - \*  $X \sim \text{Bernoulli}(q)$
- \* p-values
  - \* what is the probability (under the null hypothesis) that you would see something *at least* as extreme as what you did see?

# BINOMIAL DISTRIBUTION



- \* it turns out you don't need to simulate lots of experiments to test things about Bernoulli RVs
- \* because you can compute it exactly using the **binomial distribution**!

# BINOMIAL DISTRIBUTION

- \*  $Y \sim \text{Binomial}(n, q)$  #  $Y$  is a Binomial RV with  $n$  trials and probability  $q$  on each trial
- \*  $Y$  is number of 1's that came up in  $n$  samples from a bernoulli distribution with parameter  $q$
- \* (this is exactly what we've been doing—generating binomial RVs!)



# BINOMIAL WHAT?

- \* If you flip a weighted coin (where  $\text{Pr}(\text{heads}) = q$ )  $n$  times, what's the probability that you get  $k$  heads?



# BINOMIAL WHAT?

- \* Simpler question: if you flip a coin twice, what's the probability of heads both times?
- \* What's the probability of one heads and one tails?
- \* What's the probability of tails both times?



# BINOMIAL WHAT?

- \* The probability that two things both happen is the product of the probabilities of each thing happening
- \* (assuming that the two things are independent)
- \* This can be extended to an arbitrary number of things

# BINOMIAL WHAT?

- \* Back to the original question: if you flip a weighted coin (where  $\text{Pr}(\text{heads}) = q$ )  $n$  times, what's the probability that you get  $k$  heads?

# BINOMIAL WHAT?

- \* Let  $p_1$  = the probability of flipping a coin  $k$  times and getting heads every time
- \* Let  $p_2$  = the probability of flipping a coin  $(n-k)$  times and getting tails every time
- \* Let  $c$  = the number of ways to choose  $k$  things out of  $n$  things

# BINOMIAL WHAT?

- \*  $\text{Pr}(k \text{ heads in } n \text{ flips}) = p_1 * p_2 * c$

# BINOMIAL TEST

- \* (As last time) If we flipped a coin 100 times and got 63 heads, is it a fair coin?
- \* Formally: if the coin was fair ( $q=0.5$ ), what is the probability that we would see a result at least as extreme as 63 in 100 trials?

# BINOMIAL TEST

- \* How do we compute this probability?
- \* We can simulate, as we did before
- \* But, since we know the Binomial distribution we can just compute the probability for each  $k$  and sum!

# BINOMIAL TEST

- \* In reality we would always use **`scipy.stats.binom_test`**

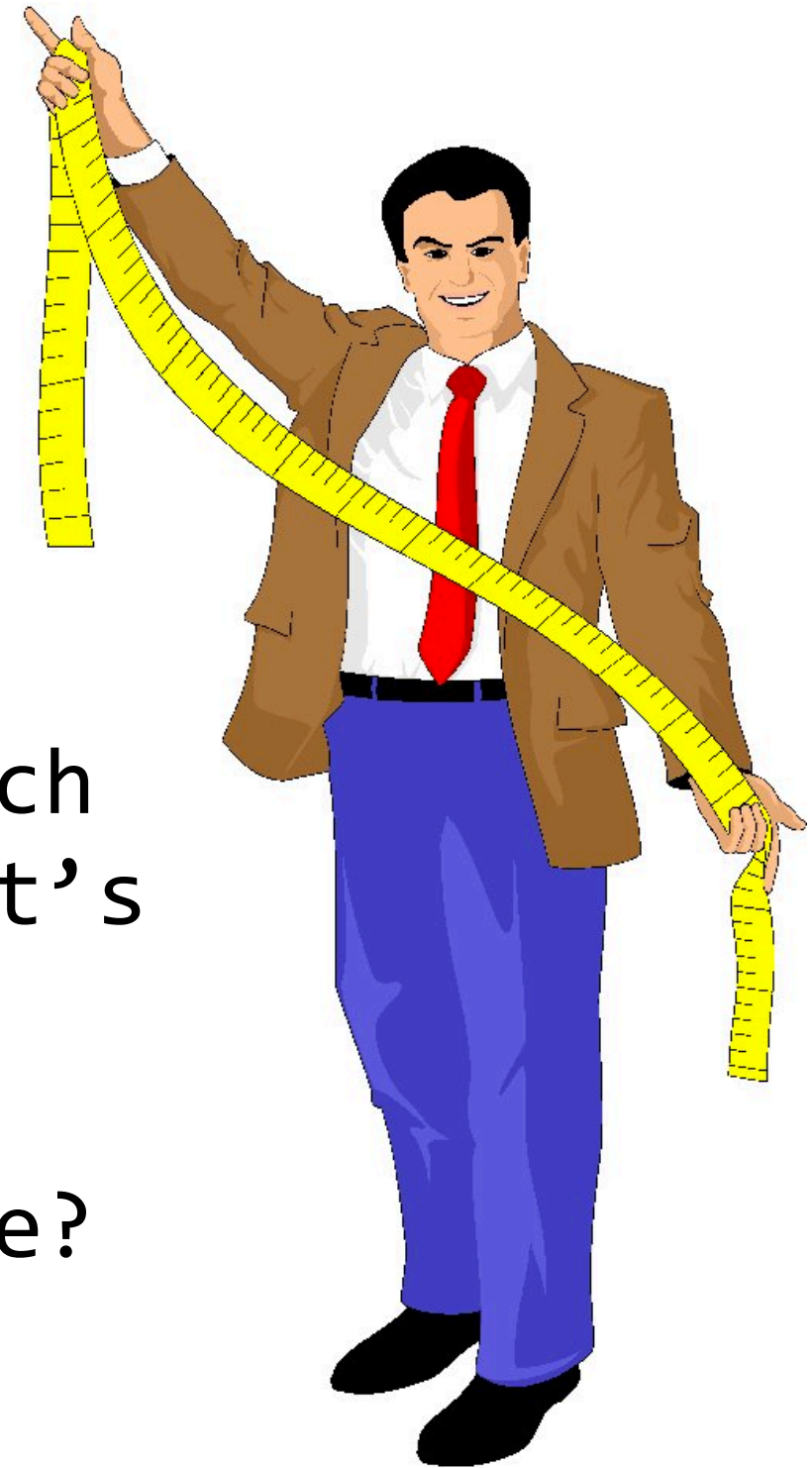
# MEAN

- \* as we've already seen when talking about numpy, the **mean** of a collection of numbers is the same as the average
- \* i.e.  $\text{mean}(\text{arr}) = \text{sum}(\text{arr}) / \text{len}(\text{arr})$



# VARIABILITY

- \* What if we want to measure how variable the data is around the mean?
- \* We could do compute how far each data point is from the mean—let's call this the deviation
- \* What will the mean deviation be?



# VARIABILITY

- \* The mean deviation is always zero!
- \* So obviously we can't just average deviations to get a sense of how variable the data is
- \* One thing we could do is take the mean *squared* deviation
- \* This is the *variance* of the data

# VARIABILITY

- \* Variance can also be obtained using `arr.var()` in numpy

# VARIABILITY

- \* Another useful number is the square root of the mean squared deviation
- \* This is the *standard deviation*
- \* Standard deviation can be obtained using `arr.std()` in numpy

**END**