In lecture, we have seen the important uses of estimation, center, and spread. Today's discussion worksheet is focused primarily on a small recap of them all.

## Estimation

Estimation is the idea of trying to estimate an unknown value using data. In lecture, we saw the example of estimating the number of warplanes based on a (possibly small) sample. In many situations, we can estimate the same quantity in multiple ways. For example, we could take the average of our sample and multiply it by 2 as one estimate for the total number of warplanes, but we could also use the maximum element in a sample.

Often, we're interested in seeing how an estimate varies with different random samples. If we have access to the full data, we can simulate many random samples, take an estimate from each, and examine a histogram of those estimates. We sometimes call this a "sampling distribution."

In terms of coding, we often accomplish this using *iteration*, using a for loop in order to make many estimates and create a distribution.

## Center

Center, average, or mean all refer to essentially the same value. One way to calculate it is to take the sum of all elements, and then divide by the amount of elements present, a method you might be familiar with. Another way to calculate our mean is to multiply each value by its proportion in the set, and then summing each of these values up. This calculation might be helpful particularly in the case of histograms. For example, if we had a list [3,3,4,4,4,4,5,6]. We could add up all of the elements in the list and divide by 8, the length of the list, but it might be easier to see that we have a 0.25 proportion of 3s, 0.5 proportion of 4s, .125 proportion of 5s, and a .125 proportions of 6s. Hence, our center can be calculated by: 3\*.25+4\*.5+5\*.125+6\*.125, which is 4.125. Notice, also, that this element is not in our original list. In this way, the mean is the weighted average for the distinct values, where the weights are proportions. So, the mean is the balance point of a histogram.

So, what is the relationship between the mean and the median? The median is the 50 percent point of the data, which is not necessarily equal to the mean. If the mean is larger than the median, then we will say our data is right-skewed, as there exists a tail, causing the mean to be pulled to the right.

Luckily, we can use the function np.mean in order to take the mean of an array or of a column of a table easily for us.

## **Spread**

Now, we have seen why the center of a distribution is important, but it's also important to note how far away, on average, elements are from the center. So, we first look at the deviations of each of the elements from the average. This causes some deviations to be negative, so we should square these, and taking the average of these gets us the *variance*, the average squared deviation from the mean. But, depending on what units we are working with, we now have a units squared, so after all of that we need to take the square root. This calculation gets us the *standard deviation*, an important tool in exploring the spread of a distribution. Hence, the SD is the root mean square of deviations from average.

In code, np. std does this long computation for us on arrays.

So, why do we use SDs? Because, a bulk of our entries are guaranteed to be within a positive or negative distance of a few SDs from our mean. What's a few? Usually, it is no more than 5. This leads us to an im-

portant equation called Chebyschev's inequality, which says that for some number z, the amount of entries that within z \* SD of the mean is at least  $1 - \frac{1}{z^2}$ . Notice the phrase "at least", which says that the proportion can be more, but never less.

Another useful tool is standard units, which are often denoted z. In general, a standard unit is  $\frac{\text{value} - \text{mean}}{\text{standard deviation}}$ , with a standard unit of 0 meaning that our value is equal to our mean. The farther away we get from the mean, the farther our z gets from 0, in either the positive of negative direction.

## **Practice Problems**

- 1. Here is a list of numbers: [0.7, 1.6, 9.8, 3.2, 5.4, 0.8, 7.7, 6.3, 2.2, 4.1, 8.1, 6.5, 3.7, 0.6, 9.9, 8.8, 3.1, 5.7, 9.1]. Using it, answer the following questions
  - (a) Without doing any math, guess whether the average is around 1, 5, or 10.
  - (b) Without doing any math, guess whether the SD is around 1, 3, or 6.
- 2. Suppose we have two lists, one is [5, 4, 6, 3, 7, 2, 7], and the other one is [5, 4, 6, 3, 7, 2, 7, 5, 5, 5, 5]. Which of the lists has a smaller standard deviation? Explain your answer, but do not do any calculations.
- 3. Use the three histograms below for the following questions.
  - (a) Match each histogram with a unique possible average: 40, 50, 60.
  - (b) Match each histogram with a description: The median is greater than the average, The median is less than the average, The median is equal to the average
  - (c) Is the standard deviation of histogram 3 closer to 5, 15, or 50?
  - (d) True or false: Histogram 1 has a standard deviation much smaller than that of histogram 3. Explain.

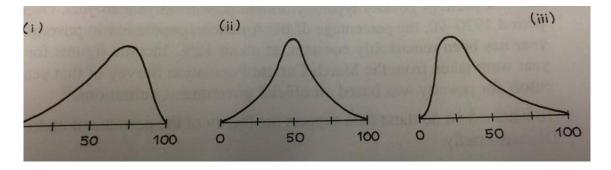


Figure 1: Histograms for question 3

- 4. A study on college students found that men had an average weight of 66 kg and a standard deviation of 9 kg, while the women had an average weight of about 55 kg and a standard deviation of 9 kg. If you took the men and women together, would the standard deviation of their weights be smaller, equal to, or bigger than 9 kg? Why?
- 5. Assume we are given the following table:

Color	Shape	Amount	Price
Red	Round	4	1.30
Green	Rectangular	6	1.20
Blue	Rectangular	12	2.0
Red	Round	7	1.75
Green	Rectangular	9	1.40
Green	Round	2	1.0

Use the table, named marbles, from discussion 4 to compare the distribution of prices for round marble packages versus rectangular marble packages. Use code in order to get the necessary aspects of our distribution.