



NASA GLOBE UCLA CLOUD GAZE

My Project Objective

- Use three datasets:
 - CLOUD GAZE Cloud Cover
 - CLOUD GAZE Cloud Type
 - GLOBE Observations
- Work with these datasets to:
 - Examine usability of the data to catch and fix errors before other scientists use the data
 - Explore different averaging methods to find the best way to average cloud cover values
 - Compare averages from CLOUD GAZE, GLOBE, and the satellites

Example observation: • One image for each direction: • N, E, S, W, Up











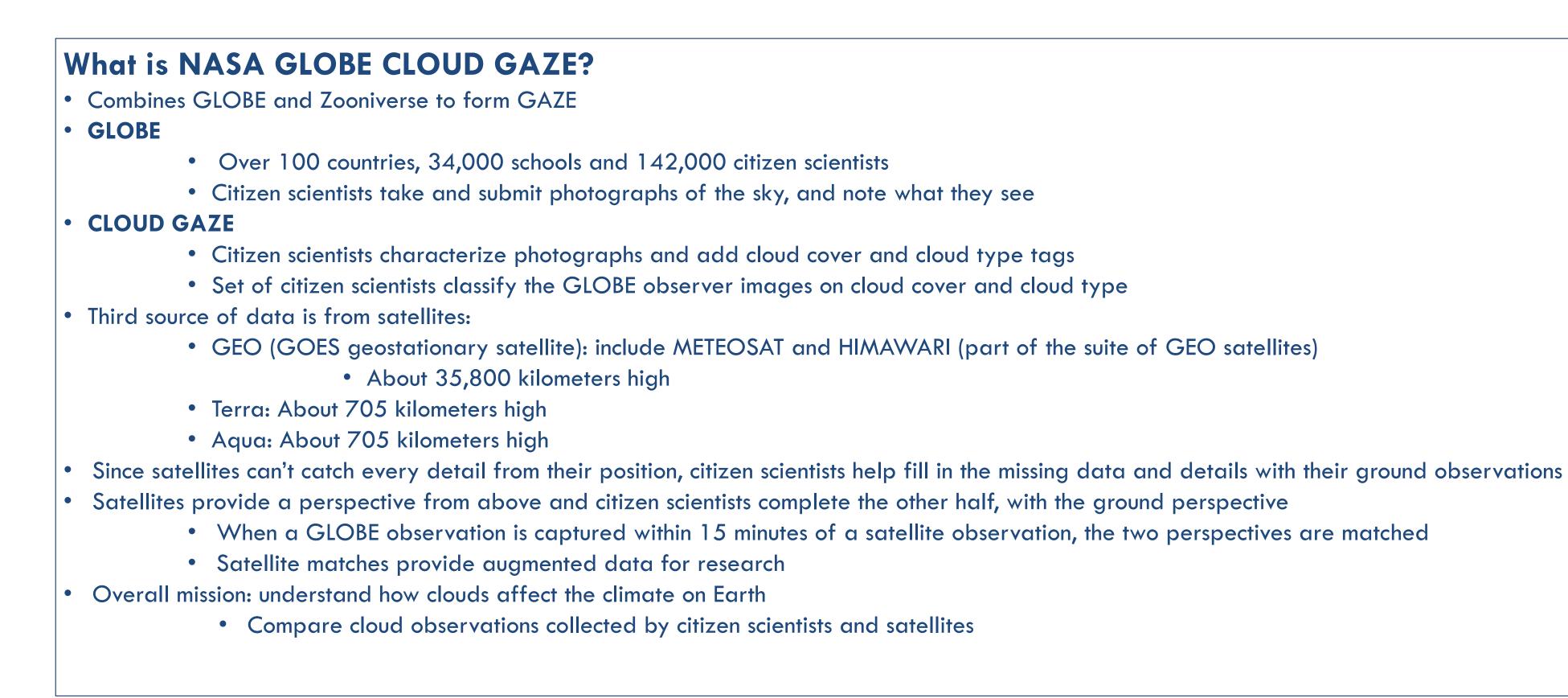
Exploring the data

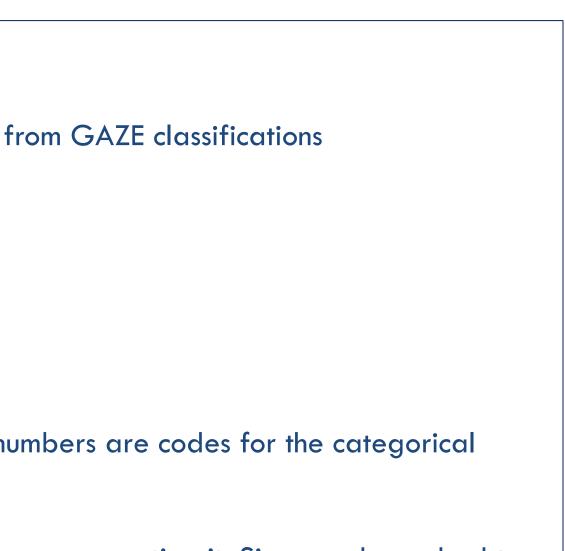
- I focused on averaging cloud cover: 'Most Likely' column
- 'Most Likely' represents the most likely answer for total cloud cover from GAZE classifications 'Most Likely' cloud cover data:
 - Few (less than 10%)
 - Isolated (10-20%)
 - Scattered (25-50%)
 - Broken (50-90%)
 - Overcast (more than 90%)
 - Other (clouds and sky more than 25% blocked, not clouds or sky)
- 'Most Likely' data is represented by integers ranging from 0-5, however these numbers are codes for the categorical values
- Limitations of the data
 - Unique structure: Each row is a separate observation with five images representing it. Since each row had to be treated as its own population in a sense, I was limited on how to find averages and compare values. The population of each observation was too small to use any type of statistical test or complex statistical method, and the data was not continuous which prevented it from meeting the assumptions of many tests and methods in statistics.
 - Categorical: The 'Most Likely' column of data is categorical which limits the possible averaging methods to just qualitative averages, which is equivalent to finding the mode.

Averaging methods

- Qualitative average vs numerical average
 - Finding the mode was a problem here because many of the observ multimodal data, which means that there were multiple modes bec were equally common values.
- My steps taken in order to be able to properly analyze categorical average of t
 - Convert integer to a meaningful total cloud cover numeric value
 - Take the midpoint of each category's interval
 - Assign this value as the new 'Most Likely' value • Take average of all five directions using the meaningful numeric vo
 - Convert numeric average back to one of the given categories base cloud cover intervals

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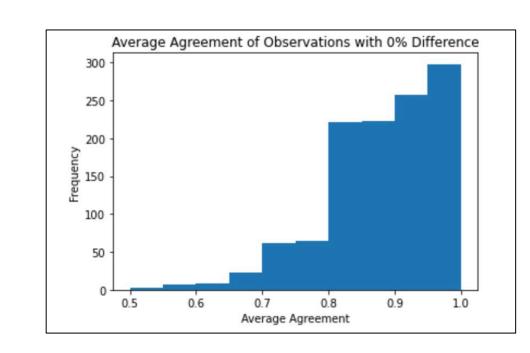


Measuring precision of averages

- Method: percent difference
- Measures difference between two related values
- 43% of the data has 0% average percent difference
- 11% of the data has over 100% average percent difference
- Once I found the averages, I wanted to examine how precise and reliable my average was so I used percent difference to measure the difference between two values.
- I found the percent difference between my average and each direction. I averaged them together for a single percent difference for each observation.
- Almost half of the observations (43%) had a percent difference of zero, which means that all the values were the same, which indicates a precise average that I found. Looking at the plot to the right, you can see that its distribution is mostly concentrated around lower average percent difference.
- Only about 11% of the data had 100 or more percent difference, meaning that the values were very different.

Percent Difference and Agreement

- When analyzing the values, I was able to verify that agreement corresponds to percent difference. • When there is a higher percent difference, there is lower agreement and the cloud cover data tends to have higher standard deviation. This means that the data is more spread out and there is more variation in the responses of the classifications. As a result, a high percent difference is inevitable since there is less agreement in the values.
- The left plot shows the average agreement of observations with 0% difference and the distribution is concentrated around higher agreement.



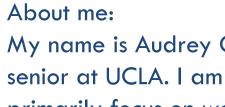
1 1	North Numeric Most Likely	East Numeric Most Likely	South Numeric Most Likely	West Numeric Most Likely	Up Numeric Most Likely	Numeric Most Likely Average	Average Most Likely (Categorical)
rvations had cause there	70.0	70.0	70.0	70.0	70.0	70.0	Broken
the data:	5.0	5.0	5.0	5.0	5.0	5.0	Few
	5.0	5.0	5.0	17.5	5.0	7.5	Few
alues sed on the	37.5	70.0	37.5	37.5	37.5	44.0	Scattered
	70.0	70.0	70.0	70.0	70.0	70.0	Broken

Audrey Cabrera **Data Science Intern Summer 2022**

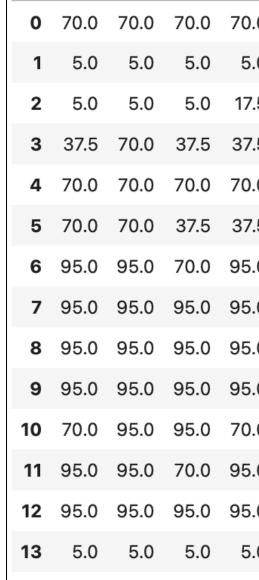


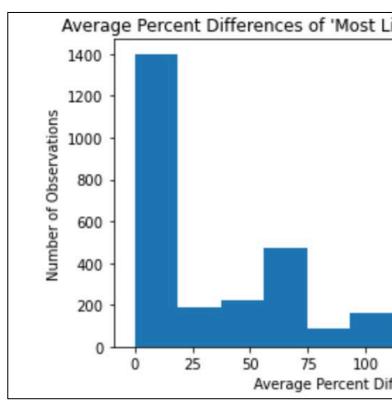
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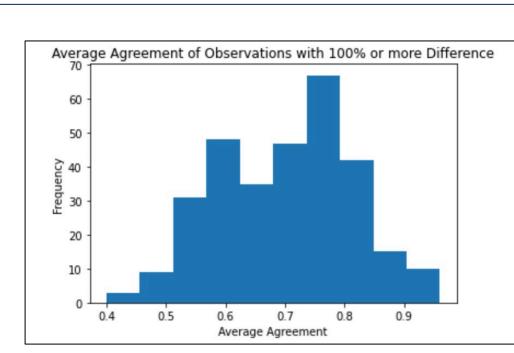


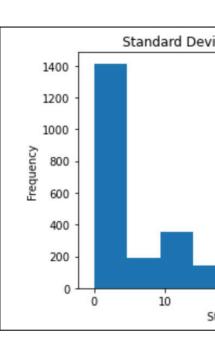
grad	uate.						
	Table	e of	Reca	lculc	ıted	Averaç	ges:
	Ν	E	S	W	Up	My Avg	Avg % Diff
0	70.0	70.0	70.0	70.0	70.0	70.0	0.00
1	5.0	5.0	5.0	5.0	5.0	5.0	0.00
2	5.0	5.0	5.0	17.5	5.0	7.5	48.00
3	37.5	70.0	37.5	37.5	37.5	44.0	21.88
4	70.0	70.0	70.0	70.0	70.0	70.0	0.00
5	70.0	70.0	37.5	37.5	70.0	57.0	28.79
6	95.0	95.0	70.0	95.0	95.0	90.0	9.32
7	95.0	95.0	95.0	95.0	95.0	95.0	0.00
8	95.0	95.0	95.0	95.0	95.0	95.0	0.00
9	95.0	95.0	95.0	95.0	95.0	95.0	0.00
10	70.0	95.0	95.0	70.0	95.0	85.0	14.41
11	95.0	95.0	70.0	95.0	70.0	85.0	14.41
12	95.0	95.0	95.0	95.0	95.0	95.0	0.00
13	5.0	5.0	5.0	5.0	5.0	5.0	0.00
14	70.0	70.0	0.0	37.5	70.0	49.5	66.10





• Whereas in the middle plot, we can see that there is lower agreement on average for the data with 100% or more difference.







My name is Audrey Cabrera and I am a rising senior at UCLA. I am majoring in statistics and primarily focus on working with data. I intend to pursue a career in the data science field after I

Standard Deviations of Most Likely Values