

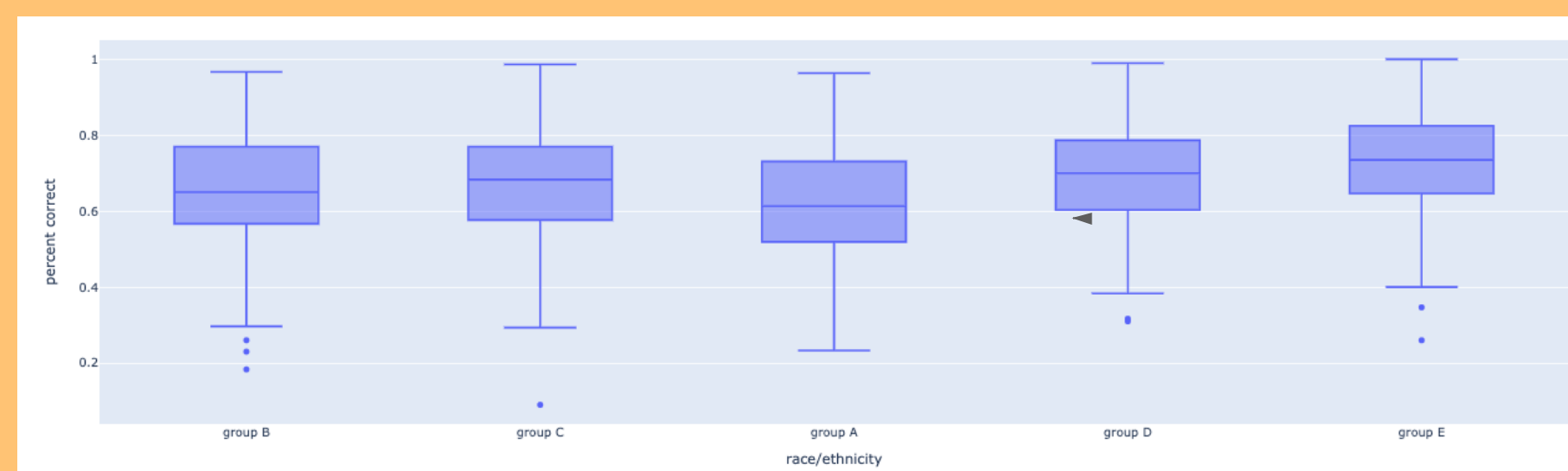
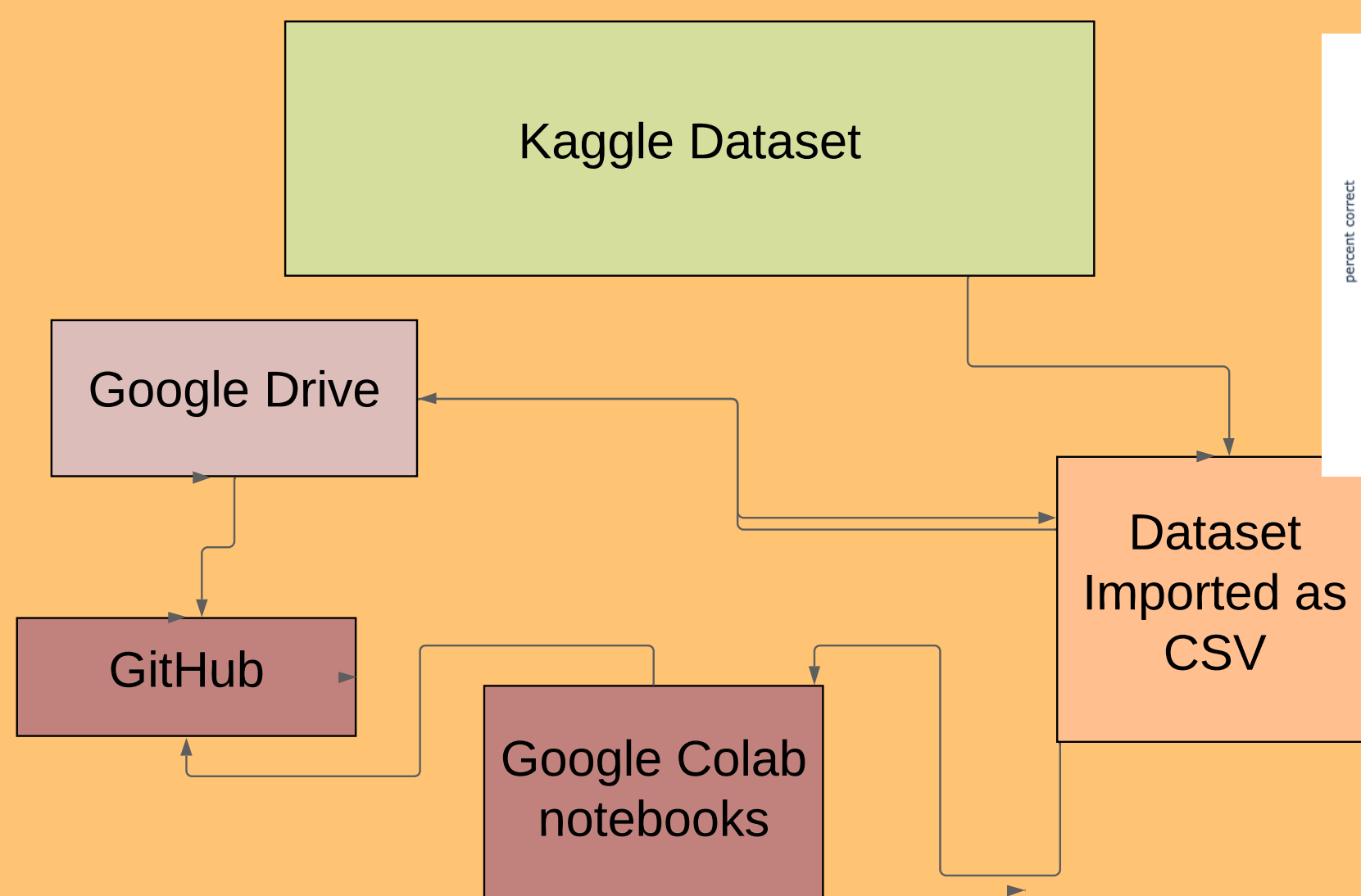
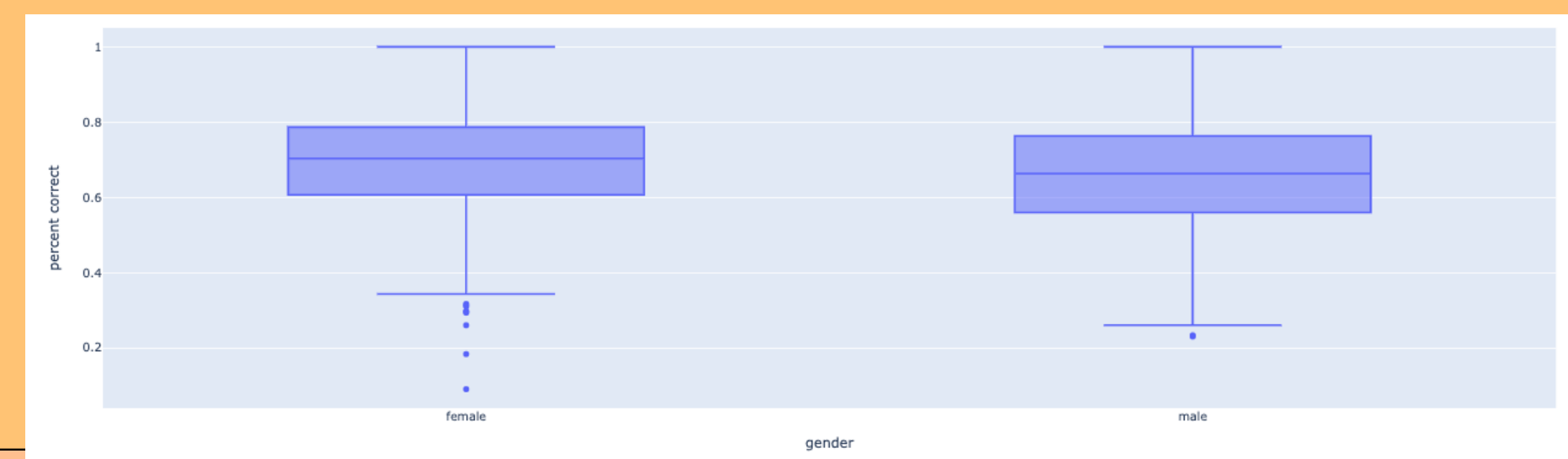
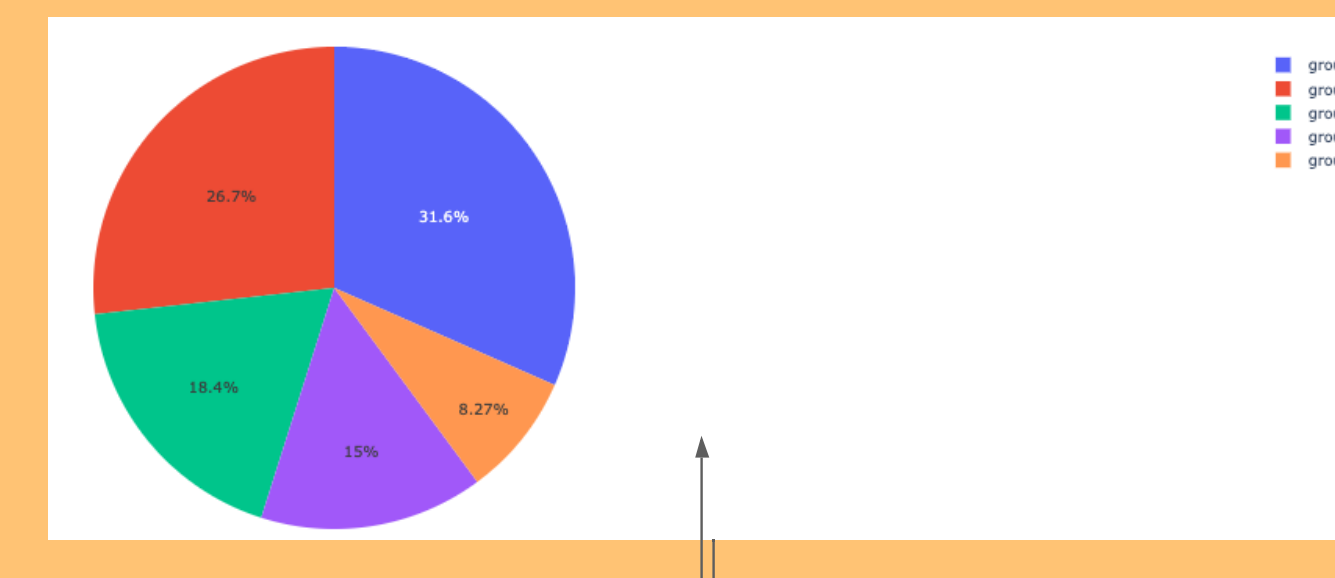
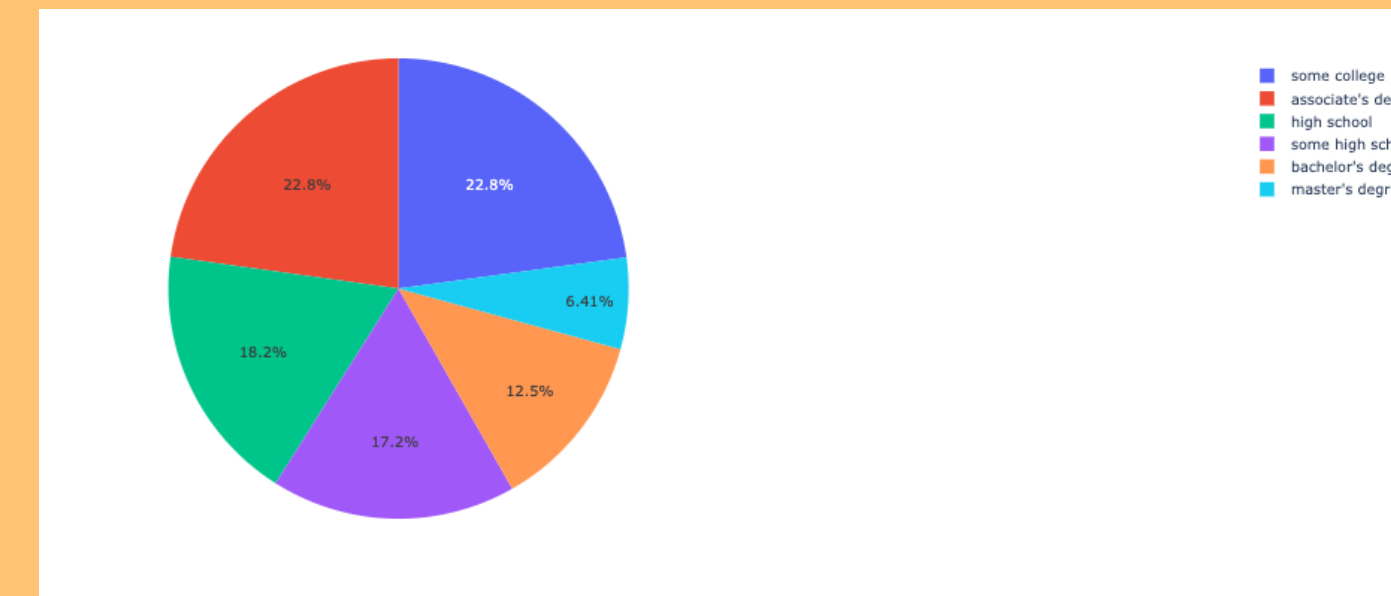
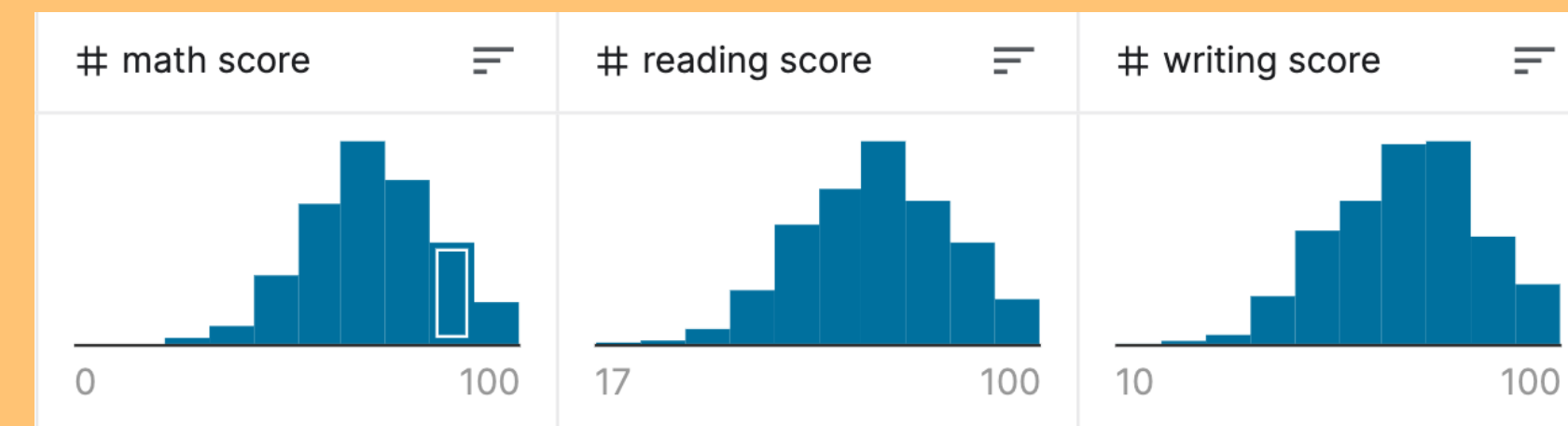
High Score Test Scores

DH 100 Theory and Methods | Channing Lee | 5/30

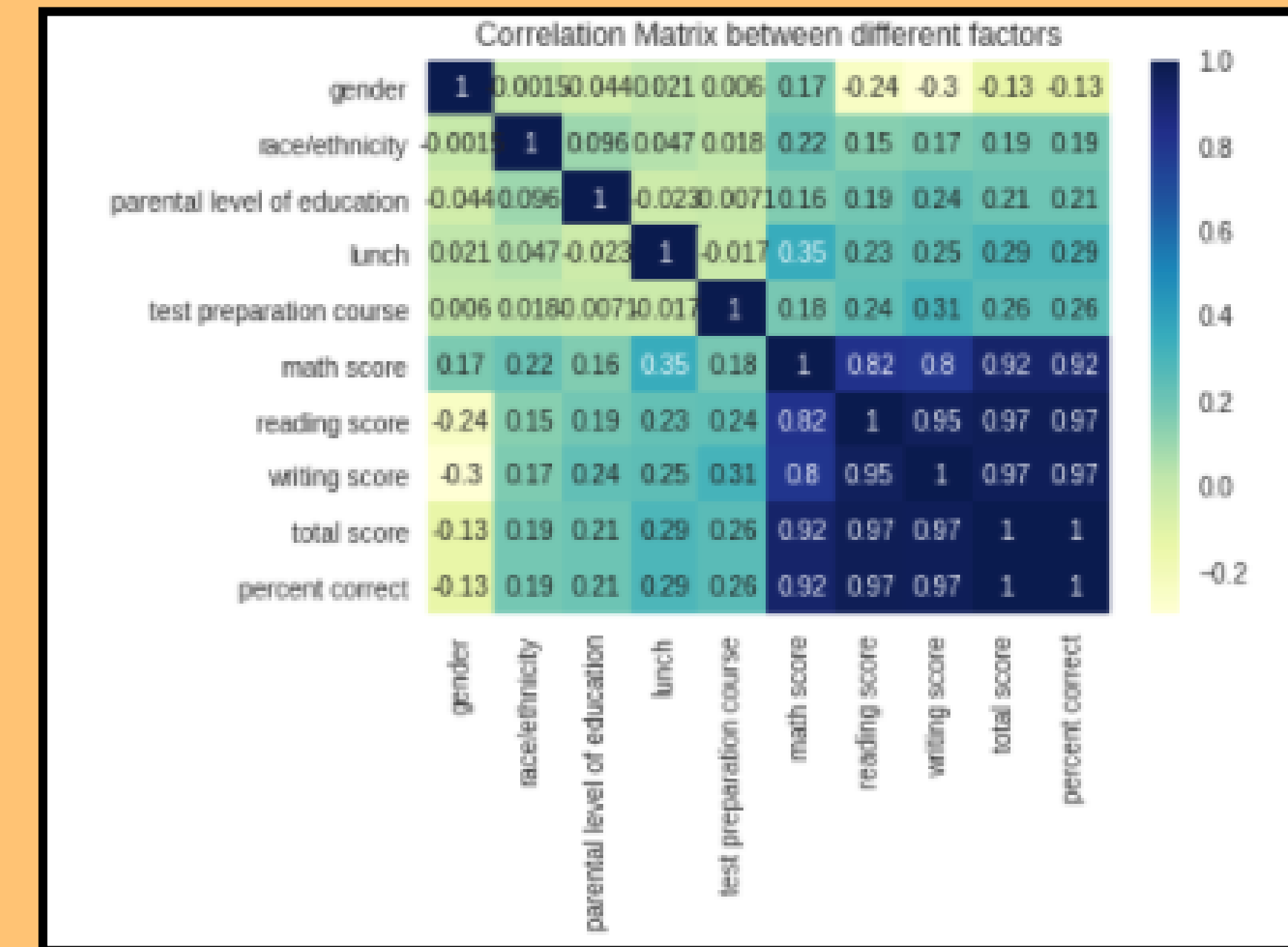
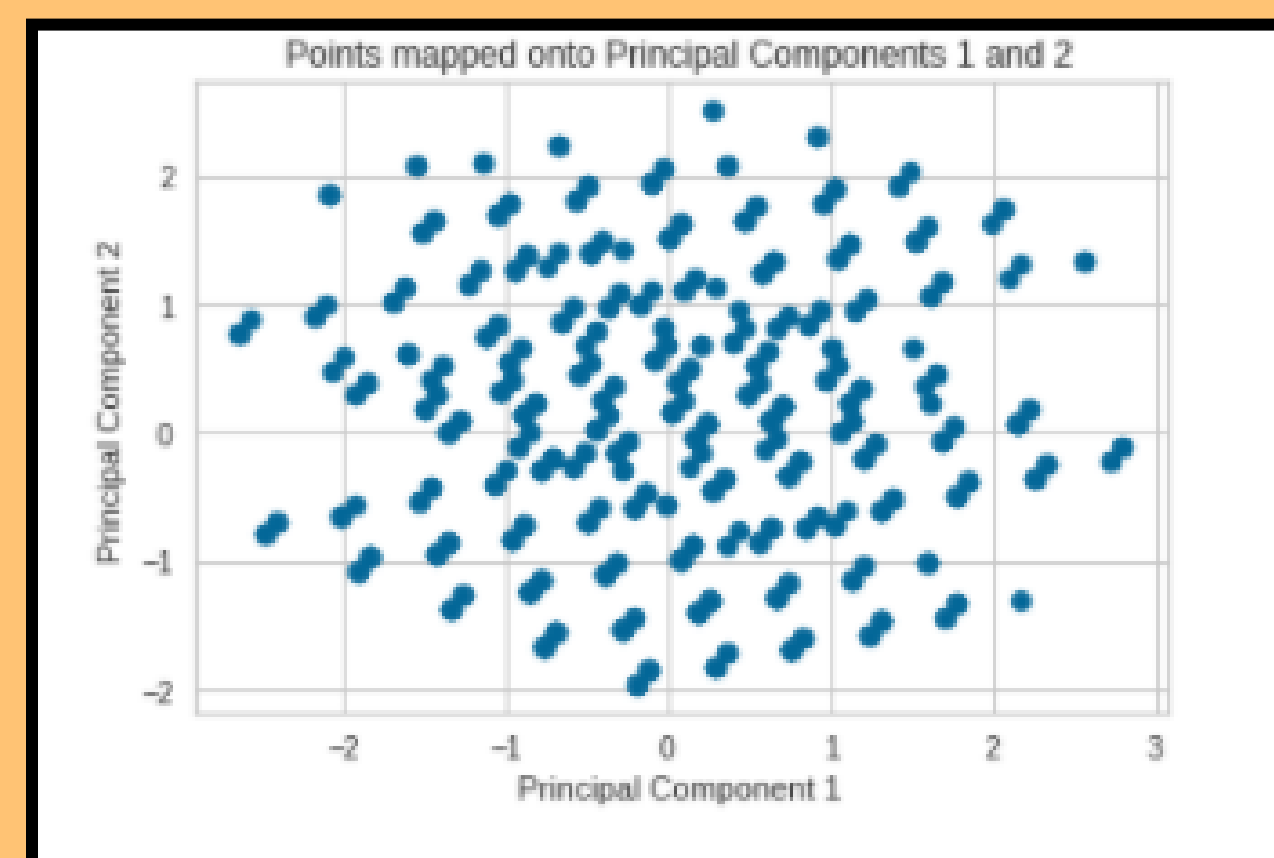
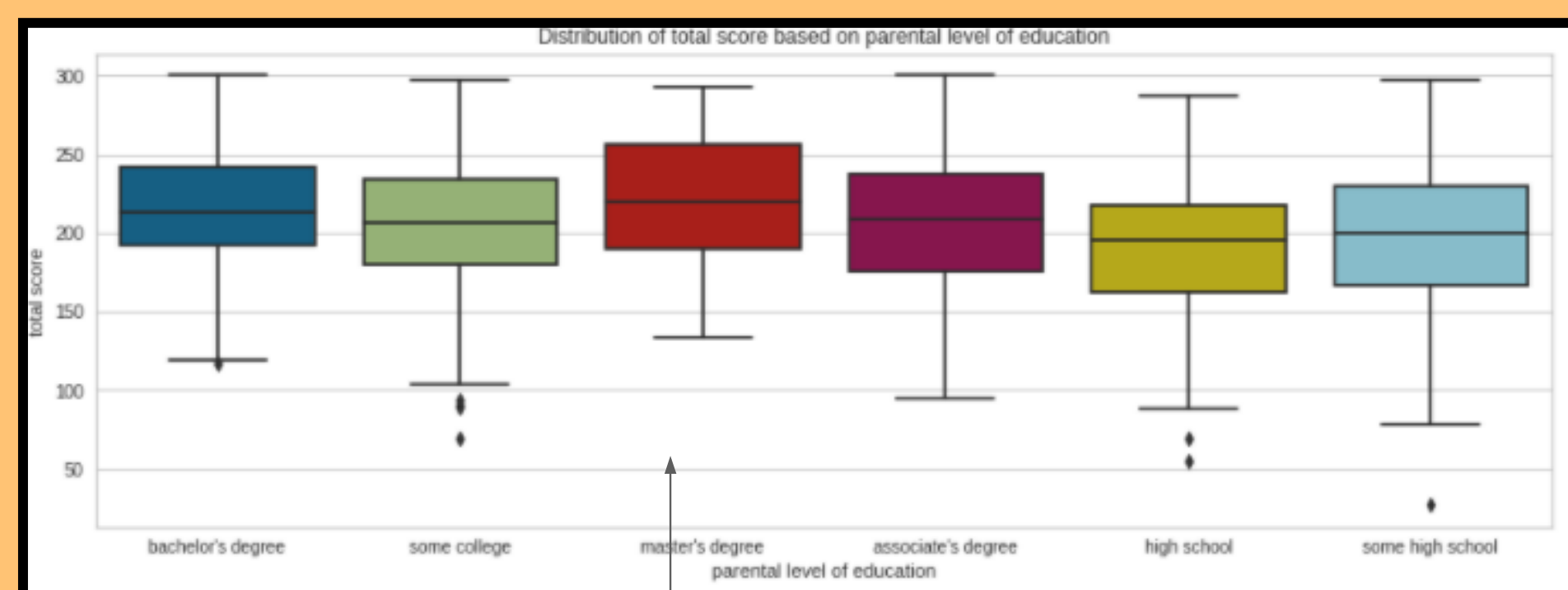
1) Dataset
<https://www.kaggle.com/spscientist/students-performance-in-exams>
(PDF, TXT, CSV)

2) How does factors such as gender, race/ethnicity, parental level of education, and etc affect student test performance

Introduction / "Hook"



- 1) High School Test Scores
- 2) DigHum 100 Theory & Methods in the Digital Humanities
- 3) Dr. Adam Anderson
- 4) Channing Lee



Interpreting your results:

- 1) return to research question & how the results are answered by your methods.
- 2) It should explain how the visuals can be interpreted, and demonstrate your knowledge of the subject matter & corpus.

I have not ran my Machine learning methods yet but I created some visualizations so I can find trends in the data and hopefully develop some intuition.

Conclusions/Further steps

As a result of my research we are able to see that lunch and test preparation have the largest impact on test scores. To put my results into action I would recommend the school to offer lunch to students that may not have access to it. This would significantly inhance a student's cognitive ability and allow them to focus in class resulting in high test scores. Another recommendation would be to put more time into standardize test preparation. Though standarize testing is dying out, it is almost impossible to receive admission to a top ranked school without a good ACT or SAT score. By providing students with a class period every week, students will be prepared for the test and have a bright future ahead of them

Work Cited

include the work you are citing:

- 1) High School Test Scores
- 2) DH 100
- 3) Instructor: Dr Anderson
- 4) Student: Channing Lee.

Seshapanpu, Jakki. "Students Performance in Exams." *Kaggle*, 9 Nov. 2018, www.kaggle.com/spscientist/students-performance-in-exams
<https://github.com/earthimmortal/DigHum100>

Discussion of results

Linear Regression:

Math Score: From the analysis we could see that the Parents level of education had the least amount of impact on Math scores while whether the students had lunch or not made the most amount of difference.

Reading Score: For reading score, gender had barely any impact at all while lunch and test preparation mattered the most

Writing Score: Gender had a negative impact on the scores and Once again Preparation had the most impact on Writing Performance.

Ridge Regression:

Math Score: Parental level of education had the least amount of impact while lunch had the most amount of impace

Reading Score: Gender had a negative impact on Reading and test preparation mattered the most

Writing Score: Gender also had a negative impact on Reading and Test preparation mattered the most.

Lasso Regression:

Math Score: Parent level of education had the least impact. Lunch had the most impact

Reading Score: Gender had a negative impact, Test prep most impact

Writing Score: Gender had negative impact, test prep most impact

Random Forest

Instad of using Random forest for classification, I used it for regression to find predictors of my model. Running Random Forest produced subpar results. Using Root Mean Squared Error as a metric, running Random Forest resulted in a lower Root Mean Squared Error than all the other methods.

PCA:

I was able to find split my dataset into components to reduce the deimnsion of the data. While plotting the the points on the first and second components of principal components, I found a pattern as seen on the left with the points forming three separate planes. This made it apparent to me that the dataset was fabricated and not real.