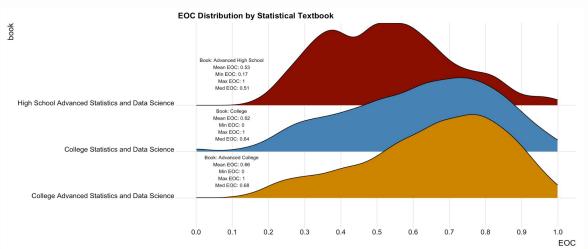
Data Fest 2024

How Best Can We Support Underperforming Students?



Ella Smith, Colin Hassett, Matthew Huang, Justin Mai

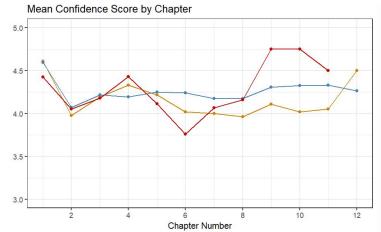
Exploratory Data Analysis

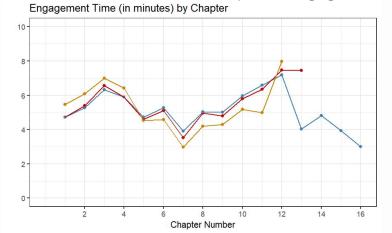


Statistical analysis: Significant

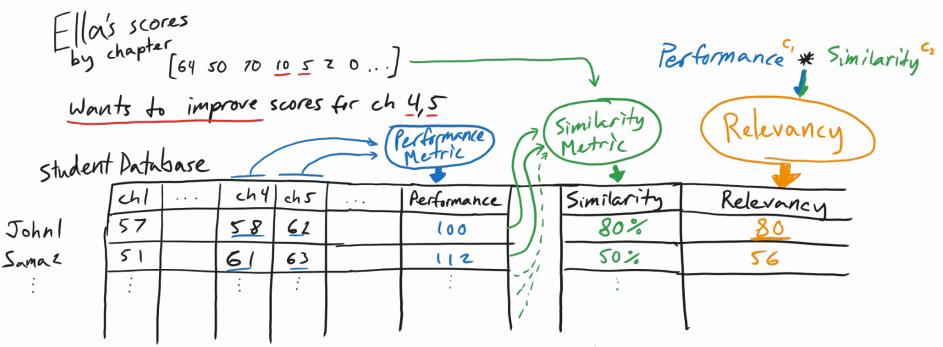
Tukey: confirmed that chapters were different in EOC; when visualizing, EOC was lower in later chapters.

P value: Confirmed had significant evidence of EOC Insignificant evidence that amount of videos impacts engagement...





Filtering Signal from Noise



*Note: The specific performance metric we used also takes into account # of attempts:

score * attempts^(-1/2)

- *Note: The similarity metric considers more than just score data, including book, class, and sentiment
 - The specific measure we used is cosine similarity, but others such as correlation or euclidean distance can be used depending on performance considerations

* C1, C2 are weights to scale depending on importance

Finding Recommendations

Top n Relevant Students:

*	ch4_attempts	ch5_attempts	ch4_correct	ch5_correct	performance2	cosine_sim	weighted_performance2
1	182	84	91	49	5.272705	0.7670921	3.102619
2	189	86	99	55	4.696956	0.7841144	2.887855
3	185	96	101	52	4.500000	0.8062652	2.925286
4	194	86	86	51	4.769728	0.7118950	2.417272
5	167	84	95	45	5.167494	0.6599228	2.250434

Example Insight:

diff_in_attempts Named num [1:5] -50.35 23.61 33.63 84.9 -8.44

By finding average **differences** in **attempts** by chapter, we might recommend that Ella study **less** for chapter 1, more for 2 and 3, and **much more** for 4.

Other Insights to look for:

- Differences in sentiment: Ella's low perception of a chapter's usefulness could prompt an instructor to highlight answering "why should you care?"
- Differences in **engagement time**