## - Midterms

- You may be getting frustrated that you are getting good results from the models you are applying
  - Data science is exploratory
  - Data science is difficult
  - You've learned a lot so far
  - Feature engineering
  - Understanding the dataset
  - Ask interesting questions
  - Get interesting answers about the data
  - 80% report
    - □ More important that you can explain and makes sense than one you can't explain but performs better
    - □ Make sure you focus on this report
    - □ Why you are doing what you are doing
    - ☐ You should have a good understanding of the tools and techniques
    - ☐ This is more important than the kaggle competition
    - □ You should mention what is relevant to the general result that it's
    - □ Don't need all the code you did, but should have everything that is relevant
  - 20% competition
    - ☐ Hopefully you are viewing it as fun

- Follow rubric that is on Piazza
- No word count
  - Try to limit to 2? pages with an appendix
  - Concise is better
  - Make sure you are clear
  - Takes about 2 pages to address those things in the rubric
- Report and leader board are independent from each other
- 30% seen on kaggle board now, will be graded on remaining 70%
  - Already technically done, you just don't see it right now
- You all have a lot of opportunity still to get an A in this class
- You can use kaggle kernels, google co-lab, cloud resources
  - You shouldn't need to , you can do a lot with very little
- Validation
  - This reduces the amount you have for training
  - More data isn't necessarily better
  - If you skip this, you may have a workflow problem
  - There are other methods of validation we can discuss
    - □ Cross-validation
    - ☐ K-fold cross validation
  - Look at the test set

- □ Look at what we are asking you to predict
  - Are there particular movies or users that is better for predicting on the testing set
- Worth knowing what you are being asked to make predictions on
  - This may be overfitting, but that is kind of the task you have
- You can do non-machine learning to predict in this way
  - Recommendation systems
- If TFDF vectorizer is having a problem
  - Pull out specific words that relevant to each group
  - Top words per score
  - Top words overall
  - Top words that are different
- Another little trick:
  - If you look at the data
  - A lot of words are used across all the scores
  - That confuses the model
    - That makes it struggles between 2 and 4 and 1 and 5 since a lot of the same words are used
    - Maybe first try to predict whether the review is positive or negative, THEN use the other words to predict the score