- Explore feature engineering
- Train and optimize the discriminators
- Implement brutal-force generator
- Evaluation and error analysis
- Detailed report available
- Demo available

• The entire dataset will be available tomorrow

- Explore feature engineering
 - Text encoder: tokenizer -> lemmatizer -> tf-idf
 - Picture encoder: categories -> binary vector + paired similarity
 - Joint encoder: paired similarity

Notes

- Category keywords can be idf reweighted as well
- Scaled similarities present dissimilar features

- Consistency discriminator
 - Input: text picture
 - Output: consistent or inconsistent
 - Encoder: Text; Picture; Joint
 - Dataset: Triplets: text picture; random text picture; text– random picture
 - Training: Binary classification logistic regression
 - Metric: Average precision

- Reasonability discriminator
 - Input: picture
 - Output: reasonable, unreasonable
 - Encoder: Picture
 - Dataset: Pair: text real picture; text fake picture
 - Training: Binary classification logistic regression
 - Metric: Average precision

- Notes
 - interpretability

Generator

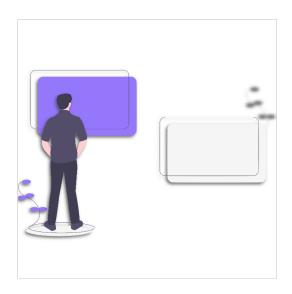
• Input: text

• Output: picture

Objective: maximize weighted sum of scores given by discriminators

• Method: Brutal-force search

• Metric: category-level average F1 score



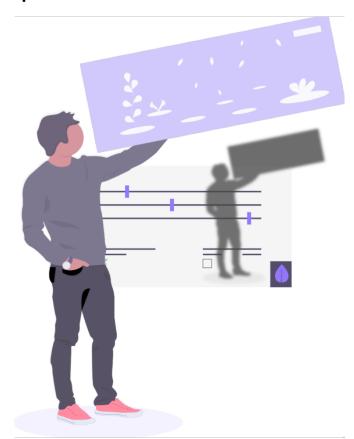
A3212 Character – Person – Stand - Back

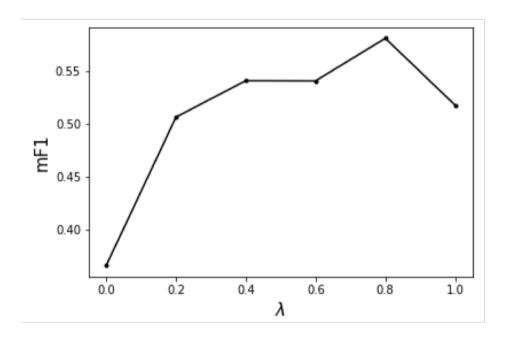
A2121 Surrounding – Room – Abstraction - Chart

[0, 1, 1, 0 | 0, ..., 1, ..., 1, ..., 0, | ..., 1, ..., 1, ..., 1, ..., 0,]

Results

- Input: "A man is showing a chart"
- Output:





Conclusion

- Joint features are critical to the decisions
- Context information may not be that important as for current short sentences

Problem

- Imbalance
 - Some highly frequent keywords dominate the decision through single-token features, and give rise to fallaciously high performance
 - A majority of keywords are too rare to be learned with non-trivial weights
- Similarities are constrained to keywords with the same POS
- No joint feature for bigrams
- No extra care for unseen words
- Random states strongly affect the performance, which probably attributes to the dataset

To-do

- K-folds validation to yield a more robust performance index
- Optimize a congregated loss function rather than train two discriminators separately
- Explore more methods to encode joint information
- Replace rare tokens with their hypernyms to increase recurrence
- Replace unseen tokens with their most similar keywords in the vocabulary
- Pixel-level picture encoder

- Thought
 - Sophisticated embedding methods, e.g. word2vec, multi-layer perceptron, and subsequent models, e.g. sequence, GAN will produce better results?

- Requirement
 - Dataset