

"A hurricane, a star, and an insect walk into a bar..."

Popular culture is strongly influenced by so-called 'speculative fiction.' A particularly influential set of books are those that win literary awards.







- The three big awards for Fantasy and Science fiction are the Hugo, the Nebula, and the Locus awards.
- We want to try to:
  - 1. Predict award nominees
  - 2. Predict award winners



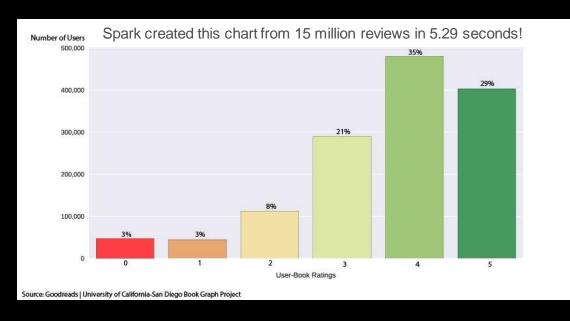
## Really big data...

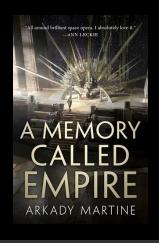
- University of California San
   Diego's (UCSD) Book Graph project
  - This includes Goodreads data from:



- 2.4 Million books
- 830 K authors
- 15 Million reviews
- 2001 2017
- Also scraped Goodreads for 3600 book records for past winners/nominees and 2020 nominees

#### Colab/Spark – fast but not friendly to merges/sub-setting





science-fiction	761 people
sci-fi	707 people
fiction	266 people
2019	238 people
scifi	183 people
space-opera	159 people
fantasy	158 people

7 of 4102 "shelves"

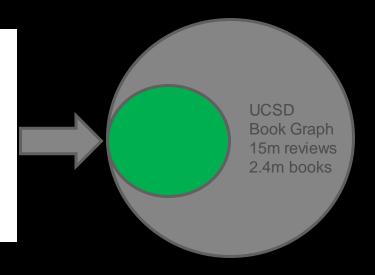
Your session crashed after using all available RAM.

View runtime logs X

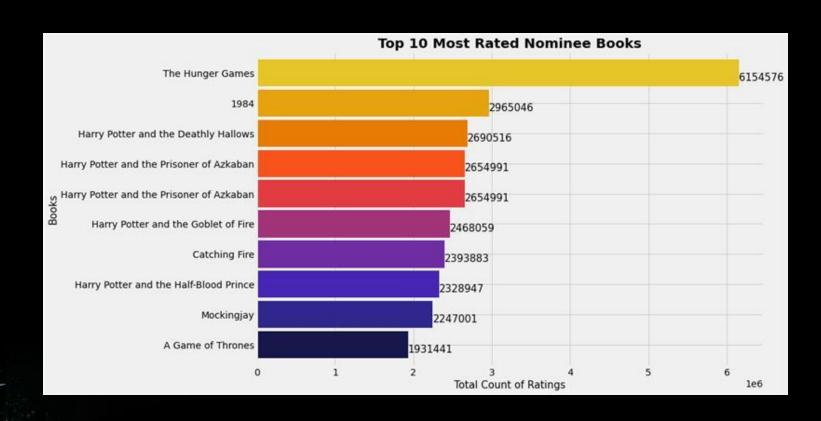


#### UCSD Book Graph Fantasy/Paranormal Subset

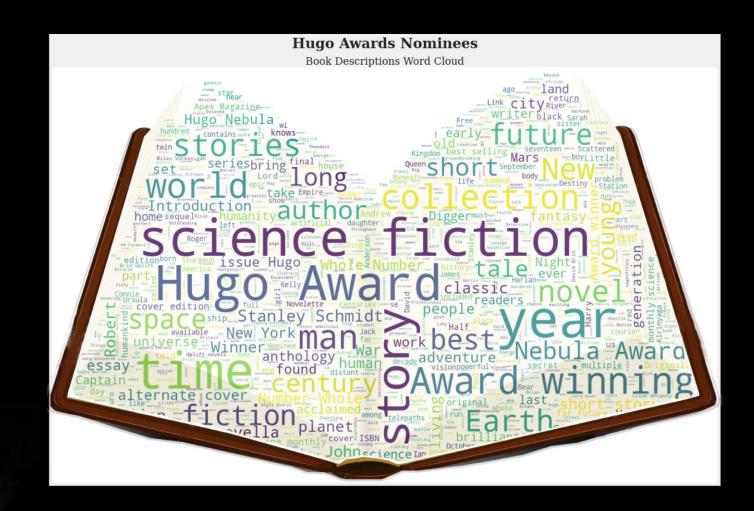
- 3.4m reviews
- 259k books
- 35% of science fiction book awards



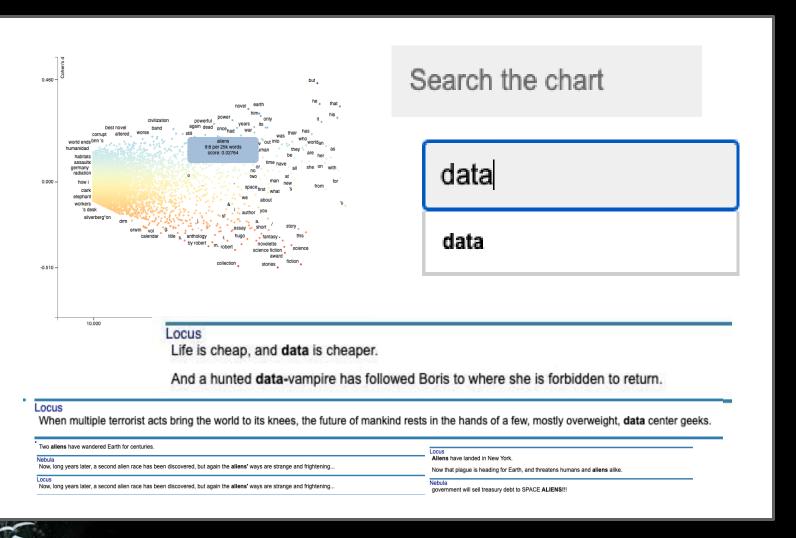
## Observe Nominated Books



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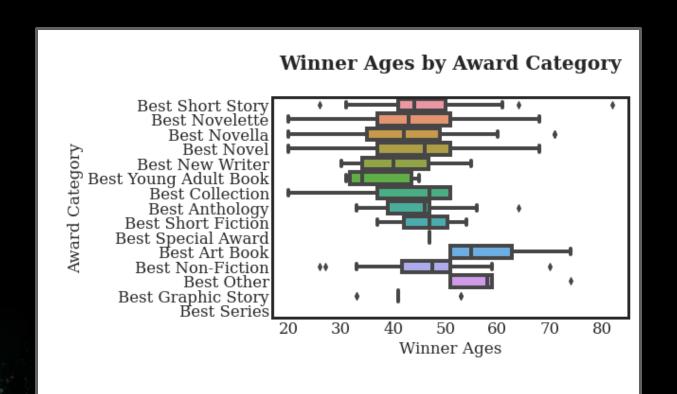
https://jonkaplan18.github.io/jonkaplan18/index.html

#### Occupations:

- Writer
- Librarian
- Anthropologist
- Publisher
- Psychologist
- Linguist
- Hypnotist
- Klingon language expert
- Lawyer
- Aerospace engineer
- Chemical engineer
- Futurist
- Truck Driver
- Marine biologist
- Psychic medium
- Occultist

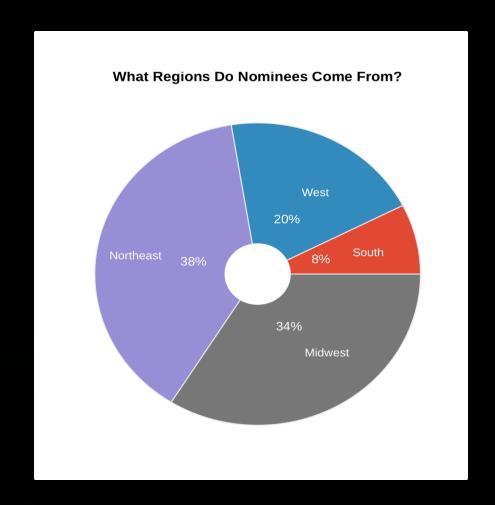


Average Age: 44 years old



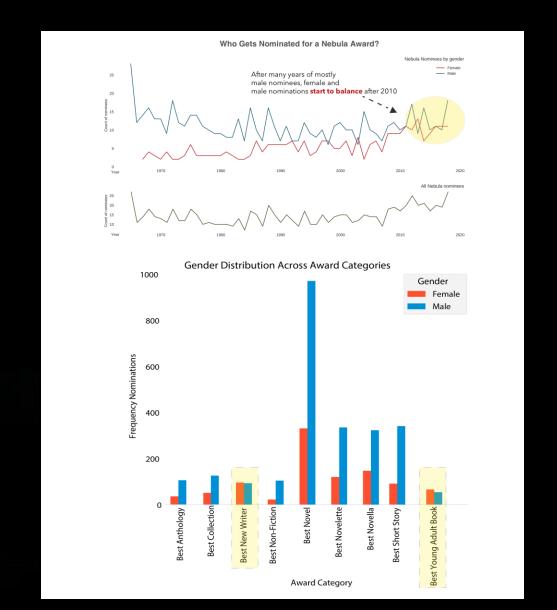
Authors by USA Region:

Mostly Northeast & Midwest



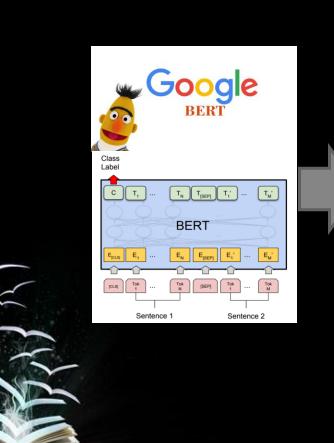


Gender: Male-heavy, but changing





## Model – picking nominees with BERT









75% ACCURACY

# Model – picking nominees with BERT + sklearn

model	params	fit_time_avg	score_time_avg	f1_avg	accuracy_avg	cv_elapsed
RandomForestClassifier	{"max_depth": 2}	0.175958	0.025919	0.491692	0.967309	0:00:00.410419 elapsed
RandomForestClassifier	{"max_depth": 10}	0.263998	0.047518	0.539130	0.966471	0:00:00.633970 elapsed
RandomForestClassifier	{"max_depth": 100}	0.281441	0.051397	0.547964	0.961861	0:00:00.677039 elapsed
svc	{"kernel": "linear", "C": 1}	165.117654	0.013786	0.496810	0.967477	0:05:30.274687 elapsed
svc	{"kernel": "rbf", "C": 1}	0.109521	0.055421	0.496801	0.967393	0:00:00.339447 elapsed
SVC	{"kernel": "rbf", "C": 10}	0.263779	0.067080	0.523079	0.966555	0:00:00.672040 elapsed
svc	{"kernel": "poly", "C": 1}	11.360120	0.017424	0.518310	0.967896	0:00:22.766108 elapsed
GaussianNB	{"priors": [0.03, 0.97]}	0.005442	0.005059	0.578774	0.911484	0:00:00.027622 elapsed
	RandomForestClassifier RandomForestClassifier RandomForestClassifier SVC SVC SVC	RandomForestClassifier {"max_depth": 2} RandomForestClassifier {"max_depth": 10} RandomForestClassifier {"max_depth": 100}  SVC {"kernel": "linear", "C": 1}  SVC {"kernel": "rbf", "C": 1}  SVC {"kernel": "rbf", "C": 10}  SVC {"kernel": "poly", "C": 1}	RandomForestClassifier         {"max_depth": 2}         0.175958           RandomForestClassifier         {"max_depth": 10}         0.263998           RandomForestClassifier         {"max_depth": 100}         0.281441           SVC         {"kernel": "linear", "C": 1}         165.117654           SVC         {"kernel": "rbf", "C": 1}         0.109521           SVC         {"kernel": "rbf", "C": 10}         0.263779           SVC         {"kernel": "poly", "C": 1}         11.360120	RandomForestClassifier         {"max_depth": 2}         0.175958         0.025919           RandomForestClassifier         {"max_depth": 10}         0.263998         0.047518           RandomForestClassifier         {"max_depth": 100}         0.281441         0.051397           SVC         {"kernel": "linear", "C": 1}         165.117654         0.013786           SVC         {"kernel": "rbf", "C": 1}         0.109521         0.055421           SVC         {"kernel": "rbf", "C": 10}         0.263779         0.067080           SVC         {"kernel": "poly", "C": 1}         11.360120         0.017424	RandomForestClassifier         {"max_depth": 2}         0.175958         0.025919         0.491692           RandomForestClassifier         {"max_depth": 10}         0.263998         0.047518         0.539130           RandomForestClassifier         {"max_depth": 100}         0.281441         0.051397         0.547964           SVC         {"kernel": "linear", "C": 1}         165.117654         0.013786         0.496810           SVC         {"kernel": "rbf", "C": 1}         0.109521         0.055421         0.496801           SVC         {"kernel": "rbf", "C": 10}         0.263779         0.067080         0.523079           SVC         {"kernel": "poly", "C": 1}         11.360120         0.017424         0.518310	RandomForestClassifier         {"max_depth": 2}         0.175958         0.025919         0.491692         0.967309           RandomForestClassifier         {"max_depth": 10}         0.263998         0.047518         0.539130         0.966471           RandomForestClassifier         {"max_depth": 100}         0.281441         0.051397         0.547964         0.961861           SVC         {"kernel": "linear", "C": 1}         165.117654         0.013786         0.496810         0.967477           SVC         {"kernel": "rbf", "C": 1}         0.109521         0.055421         0.496801         0.967393           SVC         {"kernel": "rbf", "C": 10}         0.263779         0.067080         0.523079         0.966555           SVC         {"kernel": "poly", "C": 1}         11.360120         0.017424         0.518310         0.967896

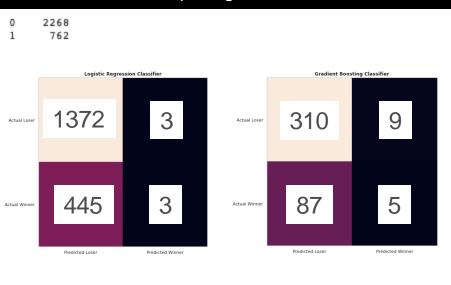




92-97% ACCURACY

# Model – Picking Winners

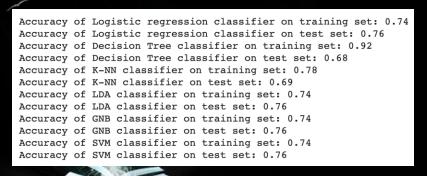
#### Imbalanced: 75% at picking non-winners



#### Balanced: 84% at picking winners



#### 5 true positives



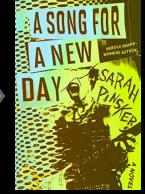
#### 552 true positives

```
Accuracy of Logistic regression classifier on training set: 0.56
Accuracy of Logistic regression classifier on test set: 0.52
Accuracy of Decision Tree classifier on training set: 1.00
Accuracy of Decision Tree classifier on test set: 0.84
Accuracy of K-NN classifier on training set: 0.80
Accuracy of K-NN classifier on test set: 0.70
Accuracy of LDA classifier on training set: 0.61
Accuracy of LDA classifier on test set: 0.58
Accuracy of GNB classifier on training set: 0.56
Accuracy of GNB classifier on test set: 0.55
Accuracy of SVM classifier on training set: 0.55
Accuracy of SVM classifier on test set: 0.53
```

## Recommend – if we just had more compute...

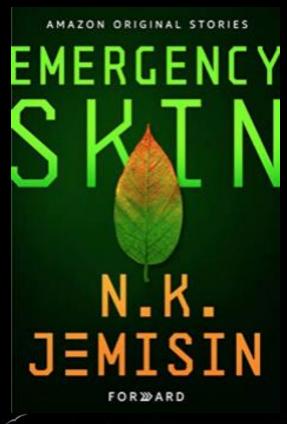
SciFi Book Awards	Accuracy
Predict Nominees (2007-2017)	75% (BERT) 92-97% (BERT + sklearn)
Predict Winners (2020)	84% (sklearn balanced)

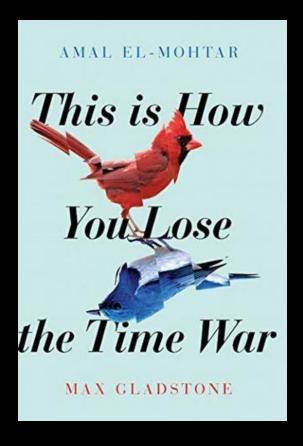




Challenges	
Spark syntax	Great for fast queries but very hard to subset and merge
Colab resources	GPU support ok but limited RAM
BERT/simple transformers	Good results on text classification but requires large compute (GPU's)

## Hugo 2020 Predicted Winners









Check our predictions August 1, 2020

## Resources and Data

- https://sites.google.com/eng.ucsd.edu/ucsdbookgraph/home
- Devlin, J., Ming-Wei, C., Lee, K., & Toutanova, K. (2019). BERT: Pre-training of deep bidirectional transformers for language understanding. Ithaca: Cornell University Library, arXiv.org. Retrieved from https://search-proquestcom.libezproxy2.syr.edu/docview/2118630252?accountid=14214
- https://gluon-nlp.mxnet.io/examples/sentence\_embedding/bert.html
- https://github.com/ThilinaRajapakse/simpletransformers
- https://github.com/huggingface/transformers
- <a href="https://simpletransformers.ai/littps://medium.com/swlh/simple-transformers-multi-class-text-classification-with-bert-roberta-xlnet-xlm-and-8b585000ce3a">https://simpletransformers.ai/littps://medium.com/swlh/simple-transformers-multi-class-text-classification-with-bert-roberta-xlnet-xlm-and-8b585000ce3a</a>
- Nominees Dataframe
- nom\_awards = pd.read\_csv('https://www.dropbox.com/s/1 mll9m9r27wy5iz/nominees\_stats\_awards\_all.csv?dl=1') #this is the original GoodReads scraped dataset without additional demographic info.
- nom\_awards2 =

  pd.read\_csv(https://www.dropbox.com/s/ysuekksts86beag/final\_nominations\_df.csv?dl=1") #this is the version with the merged demographic info from Worlds Without End scrape