

**Assessing Evaluation Bias  
amongst readers towards ardently  
favored/despised authors**

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# Motivation

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The key motivation behind this project is to provide some mathematical support to the arguments of a “cult-writer”, or “hated by the critics writer” etc.

## Data Details:

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Amazon Review Dataset (Julian McAuley et al) - Books

Total Size: 50Mn reviews, for 2.5 Mn books on Amazon up to 2018

Utilized for project: sample @ 100K reviews

Why? Only considered books with 50 reviews, done by reviewers with who've written 50 reviews, aka '50-core'

# Task 1

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**Highlight Ardent Promoters and Detractors for each work (author-book)**

Metric:

**Reviewers that rated highly positive or very negative for more than 80% on a particular book.**

How:

- K-Means clusters for items enough data available for 3 distinct clusters per book per author**
- Measure the changes in ratings upon removal of the highlighted book-reviewer combo**

# Approach and Setup

Author	Book	User	Rating
XYZ	Abridged Bird Law	Deandra	1.5
XYZ	Abridged Bird Law	Frank	5
XYZ	Abridged Bird Law	Charlie	4.5
XYZ	Abridged Bird Law	Mac	1.5
XYZ	Denim Chicken and other recipes	Dennis	5
XYZ	Denim Chicken and other recipes	Mac	3
XYZ	Denim Chicken and other recipes	Charlie	2
XYZ	Denim Chicken and other recipes	Deandra	4.5
XYZ	Denim Chicken and other recipes	Cricket	3
XYZ	Denim Chicken and other recipes	Frank	5

# Approach and Setup

Author	Book	User	Rating
XYZ	Abridged Bird Law	Deandra	1.5
XYZ	Abridged Bird Law	Frank	5
XYZ	Abridged Bird Law	Charlie	3.5
XYZ	Abridged Bird Law	Mac	4.5
XYZ	Denim Chicken and other dreams	Dennis	5
XYZ	Denim Chicken and other dreams	Mac	3.5
XYZ	Denim Chicken and other dreams	Charlie	4
XYZ	Denim Chicken and other dreams	Deandra	2
XYZ	Denim Chicken and other dreams	Cricket	5
XYZ	Denim Chicken and other dreams	Frank	5

# Approach and Setup

Author	Book	User	Rating
<del>XYZ</del>	<del>Abridged Bird Law</del>	<del>Deandra</del>	<del>4.5</del>
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<del>XYZ</del>	<del>Denim Chicken and other dreams</del>	<del>Deandra</del>	<del>2</del>
XYZ	Denim Chicken and other dreams	Cricket	5
<del>XYZ</del>	<del>Denim Chicken and other dreams</del>	<del>Frank</del>	<del>5</del>

# Results 1

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- The user clusters were bound by hard coded logic
- 99% of the reviews did NOT fall into the category. Reviewers fluctuated between 0.75-1.25 perc for all samples.
- The lack of significant results probably due to the high number of reviews

## User Clusters:

	mean	std	count
Low	1.72	0.45	3.14%
Normal	3.79	0.41	36.66%
High	5.00	0.00	60.19%

## Task 2

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**Evaluate distortions in product recommendations due to inflated/deflated ratings for particular authors and works**

Metric:

**Distortion =**

**$1 - (\text{Recommendation Prior Removal AND Recommendation Post Removal}) / \text{Recommendation Prior Removal}$**

How:

**Fetch collaborative fill/item-based suggestions before and after dataset cleaning. Higher distortion means worse effects.**



# Results

## Example:

```
{'0001844423', '0003302245', '0007284241', '0007350961',  
'0140431179', '0142406635', '045141411X', '0545123267',  
'1477809732', 'B00A9WRH3M'}
```

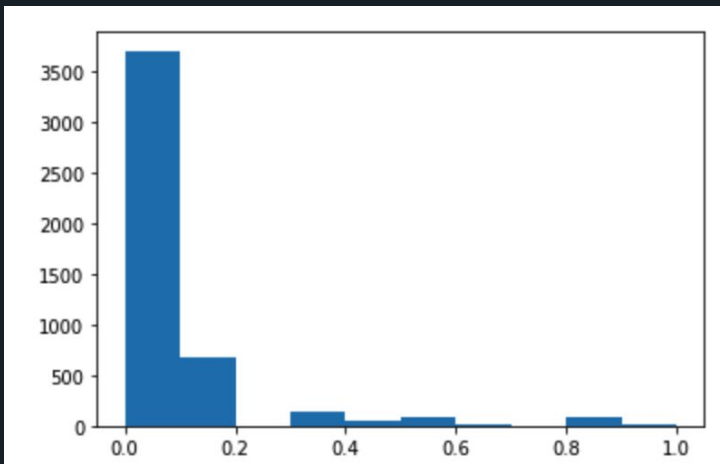
v/s

```
{'0001048767', '0003302245', '0007284241', '0007350961',  
'0140431179', '0142406635', '045141411X', '0545123267',  
'0545123267', 'B00A9WRH3M'}
```

Distortion = 0.8

Distortion over the dataset: 11.3  
[SD: 15.4]

## Distortion Density:



# Future Steps

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- Results to be recalculated 5/10/20-core data, i.e., more data, along with categorical focus as well
- Robustness of recommendations should be tested with other learning/stationary models such as kNN, Matrix Factorization/SVD
- De-biasing is a fairly well explored area in the field of recommendations, with existing methods that provide some mathematical guarantees/outlines of effect
- Demographics affected, as well focus on authors that are most affected by the biases