```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from collections import defaultdict
import json
import gzip
from urllib.request import urlopen
from collections import defaultdict
```

## → Read Data

!wget http://deepyeti.ucsd.edu/jianmo/amazon/categoryFilesSmall/Movies\_and\_TV\_5.json.c

```
--2022-03-12 00:53:21-- http://deepyeti.ucsd.edu/jianmo/amazon/categoryFilesSma
    Resolving deepyeti.ucsd.edu (deepyeti.ucsd.edu)... 169.228.63.50
    Connecting to deepyeti.ucsd.edu (deepyeti.ucsd.edu) | 169.228.63.50 | :80... connector
    HTTP request sent, awaiting response... 200 OK
    Length: 791322468 (755M) [application/octet-stream]
    Saving to: 'Movies_and_TV_5.json.gz'
    Movies and TV 5.jso 100%[===========] 754.66M 73.4MB/s
                                                                         in 11s
    2022-03-12 00:53:33 (65.9 MB/s) - 'Movies and TV 5.json.gz' saved [791322468/791]
### load the meta data
data = []
with gzip.open('Movies and TV 5.json.gz') as f:
    for 1 in f:
        j = json.loads(l.strip())
        d = \{\}
        d['overall'] = j['overall']
        if 'reviewText' in j:
            d['reviewText'] = j['reviewText']
        else:
            d['reviewText'] = None
        if 'summary' in j:
            d['summary'] = j['summary']
        else:
            d['summary'] = None
        data.append(d)
# total length of list, this number equals total number of products
print(len(data))
```

```
# first row of the list
print(data[0])

3410019
{'overall': 5.0, 'reviewText': "So sorry I didn't purchase this years ago when in the second of the list
}
```

### Clean Data

```
df = pd.DataFrame(data)
# Drop data with no review text
df = df[~df.reviewText.isnull()].copy()
# Create dataframes for each star rating
df 1 = df[df['overall'] == 1]
print('num 1-star reviews: ', len(df_1))
df 2 = df[df['overall'] == 2]
print('num 2-star reviews: ', len(df_2))
df 3 = df[df['overall'] == 3]
print('num 3-star reviews: ', len(df_3))
df 4 = df[df['overall'] == 4]
print('num 4-star reviews: ', len(df_4))
df 5 = df[df['overall'] == 5]
print('num 5-star reviews: ', len(df_5))
    num 1-star reviews: 193110
    num 2-star reviews: 172409
    num 3-star reviews: 349641
    num 4-star reviews: 665734
    num 5-star reviews: 2027544
import scipy as sp
import seaborn as sns
print("The resulting theta will be the one that maximizes the likelihood function -> t
#1star
n1 = len(df 1)
theta = len(df_1)/len(df)
X arr = np.random.choice([0, 1], p=[1-theta, theta], size=len(df))
resulting theta1 = sum(X arr)/len(df)
print("The resulting theta will be the one that maximizes the likelihood function for
```

```
#2star
n2 = len(df_2)
theta = len(df 2)/len(df)
X_arr = np.random.choice([0, 1], p=[1-theta, theta], size=len(df))
resulting theta2 = sum(X arr)/len(df)
print("The resulting theta will be the one that maximizes the likelihood function for
#3star
n3 = len(df 3)
theta = len(df 3)/len(df)
X arr = np.random.choice([0, 1], p=[1-theta, theta], size=len(df))
resulting_theta3 = sum(X_arr)/len(df)
print("The resulting theta will be the one that maximizes the likelihood function for
#4star
n4 = len(df 4)
theta = len(df 4)/len(df)
X_arr = np.random.choice([0, 1], p=[1-theta, theta], size=len(df))
resulting_theta4 = sum(X_arr)/len(df)
print("The resulting theta will be the one that maximizes the likelihood function for
#5star
n5 = len(df 5)
theta = len(df 5)/len(df)
X arr = np.random.choice([0, 1], p=[1-theta, theta], size=len(df))
resulting theta5 = sum(X arr)/len(df)
print("The resulting theta will be the one that maximizes the likelihood function for
    The resulting theta will be the one that maximizes the likelihood function -> the
    The resulting theta will be the one that maximizes the likelihood function for 1
    The resulting theta will be the one that maximizes the likelihood function for 2
    The resulting theta will be the one that maximizes the likelihood function for 3
    The resulting theta will be the one that maximizes the likelihood function for 4
    The resulting theta will be the one that maximizes the likelihood function for 5
#MAP
n = 10000
X arr = np.ones(n)
alpha = 2
beta = 2
#1 star
theta = len(df 1)/len(df)
```

```
X_arr = np.random.choice([0, 1], p=[1-theta, theta], size=n)
print(sum(X_arr) / n)
new_beta = np.asarray([[alpha + sum(X_arr[:i+1]), beta+(i+1-sum(X_arr[:i+1]))] for i i
new_beta = np.insert(new_beta, 0, [alpha, beta], 0)
beta X = np.linspace(0, 1, 1000)
my color = 'black'
fig, ax_arr = plt.subplots(ncols=6, figsize=(16,6), sharex=True)
for i, iter in enumerate([0, 5, 50, 200, 500, 1000]):
    ax = ax_arr[i]
    a, b = new_beta[iter_]
    beta_Y = sp.stats.beta.pdf(x=beta_X, a=a, b=b)
    ax.plot(beta_X, beta_Y, color=my_color, linewidth=3)
    if a > 1 and b > 1:
        mode = (a-1)/(a+b-2)
    else:
        mode = a/(a+b)
    ax.axvline(x=mode, linestyle='--', color='k')
    ax.set_title('Iteration %d: $\hat{\\theta}_{MAP}$ = %.2f'%(iter_, mode))
fig.tight layout()
#2 star
theta = len(df_2)/len(df)
X_arr = np.random.choice([0, 1], p=[1-theta, theta], size=n)
print(sum(X arr) / n)
new_beta = np.asarray([[alpha + sum(X_arr[:i+1]), beta+(i+1-sum(X_arr[:i+1]))] for i j
new beta = np.insert(new beta, 0, [alpha, beta], 0)
beta X = np.linspace(0, 1, 1000)
my color = 'black'
fig, ax arr = plt.subplots(ncols=6, figsize=(16,6), sharex=True)
for i, iter in enumerate([0, 5, 50, 200, 500, 1000]):
    ax = ax arr[i]
    a, b = new beta[iter ]
    beta Y = sp.stats.beta.pdf(x=beta X, a=a, b=b)
    ax.plot(beta X, beta Y, color=my color, linewidth=3)
    if a > 1 and b > 1:
        mode = (a-1)/(a+b-2)
    else:
        mode = a/(a+b)
    ax.axvline(x=mode, linestyle='--', color='k')
    ax.set title('Iteration %d: $\hat{\\theta} {MAP}$ = %.2f'%(iter , mode))
fig.tight layout()
#3 star
theta = len(df 3)/len(df)
X arr = np.random.choice([0, 1], p=[1-theta, theta], size=n)
print(sum(X arr) / n)
new beta = np.asarray([[alpha + sum(X arr[:i+1]), beta+(i+1-sum(X arr[:i+1]))] for i j
new beta = np.insert(new beta, 0, [alpha, beta], 0)
```

```
beta X = np.linspace(0, 1, 1000)
my_color = 'black'
fig, ax arr = plt.subplots(ncols=6, figsize=(16,6), sharex=True)
for i, iter in enumerate([0, 5, 50, 200, 500, 1000]):
    ax = ax_arr[i]
    a, b = new beta[iter ]
    beta_Y = sp.stats.beta.pdf(x=beta_X, a=a, b=b)
    ax.plot(beta_X, beta_Y, color=my_color, linewidth=3)
    if a > 1 and b > 1:
        mode = (a-1)/(a+b-2)
    else:
        mode = a/(a+b)
    ax.axvline(x=mode, linestyle='--', color='k')
    ax.set title('Iteration %d: $\hat{\\theta} {MAP}$ = %.2f'%(iter , mode))
fig.tight layout()
#4 star
theta = len(df_4)/len(df)
X_arr = np.random.choice([0, 1], p=[1-theta, theta], size=n)
print(sum(X_arr) / n)
new beta = np.asarray([[alpha + sum(X arr[:i+1]), beta+(i+1-sum(X arr[:i+1]))] for i j
new_beta = np.insert(new_beta, 0, [alpha, beta], 0)
beta_X = np.linspace(0, 1, 1000)
my color = 'black'
fig, ax arr = plt.subplots(ncols=6, figsize=(16,6), sharex=True)
for i, iter_ in enumerate([0, 5, 50, 200, 500, 1000]):
    ax = ax arr[i]
    a, b = new beta[iter ]
    beta Y = sp.stats.beta.pdf(x=beta X, a=a, b=b)
    ax.plot(beta X, beta Y, color=my color, linewidth=3)
    if a > 1 and b > 1:
        mode = (a-1)/(a+b-2)
    else:
        mode = a/(a+b)
    ax.axvline(x=mode, linestyle='--', color='k')
    ax.set_title('Iteration %d: $\hat{\\theta}_{MAP}$ = %.2f'%(iter , mode))
fig.tight layout()
#5 star
theta = len(df 5)/len(df)
X arr = np.random.choice([0, 1], p=[1-theta, theta], size=n)
print(sum(X arr) / n)
new beta = np.asarray([[alpha + sum(X arr[:i+1]), beta+(i+1-sum(X arr[:i+1]))] for i j
new beta = np.insert(new beta, 0, [alpha, beta], 0)
beta X = np.linspace(0, 1, 1000)
my color = 'black'
fig, ax arr = plt.subplots(ncols=6, figsize=(16,6), sharex=True)
for i, iter_ in enumerate([0, 5, 50, 200, 500, 1000]):
    ax = ax arr[i]
```

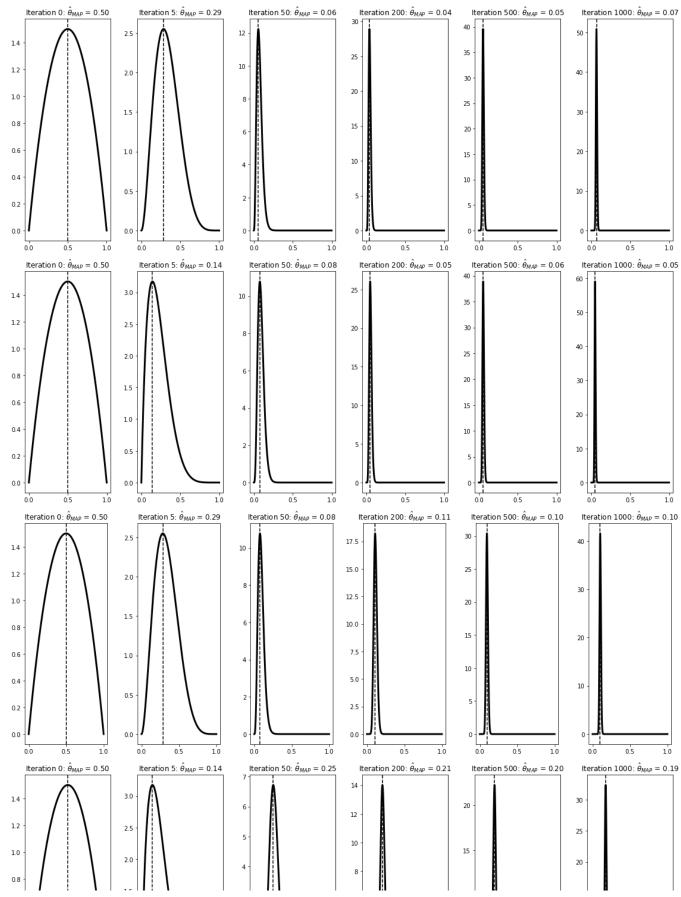
```
a, b = new_beta[iter_]
beta_Y = sp.stats.beta.pdf(x=beta_X, a=a, b=b)
ax.plot(beta_X, beta_Y, color=my_color, linewidth=3)
if a > 1 and b > 1:
    mode = (a-1)/(a+b-2)
else:
    mode = a/(a+b)
ax.axvline(x=mode, linestyle='--', color='k')
ax.set_title('Iteration %d: $\hat{\\theta}_{MAP}$ = %.2f'%(iter_, mode))
fig.tight_layout()
```

0.0524

0.0481

0.1015

0.1897
0.5813



import random

```
random.seed(42069)
# Randomly sample n_train (100k) datapoints from each rating to form training data
n_{train} = 100000
data 1 = df 1.to dict('records')
data 2 = df 2.to dict('records')
data 3 = df 3.to dict('records')
data 4 = df 4.to dict('records')
data_5 = df_5.to_dict('records')
train = random.sample(data_1, n_train) + random.sample(data_2, n_train) + random.sampl
random.shuffle(train)
len(train)
    500000
import string
# Using NLTK Library
import nltk
# import the stopwords collection from the nltkcorpus module
from nltk.corpus import stopwords
nltk.download('stopwords')
sWords = set(stopwords.words())
# New bag-of-words after removing punctuation capitalization and stopwords
wordCount = defaultdict(int)
punctuation = set(string.punctuation)
ascii lower = set(string.ascii lowercase)
```

```
whitespace = set(string.whitespace)
for d in train:
    r = ''.join([c for c in d['reviewText'].lower() if not c in punctuation and (c in
    for w in r.split():
        if not w in sWords:
            wordCount[w] += 1
print('num words case-insensitive and no punc: ', len(wordCount))
    [nltk data] Downloading package stopwords to /root/nltk data...
                   Unzipping corpora/stopwords.zip.
    [nltk data]
    num words case-insensitive and no punc: 407782
# Sort by popularity
counts = [(wordCount[w], w) for w in wordCount]
counts.sort()
counts.reverse()
words = [x[1] \text{ for } x \text{ in counts}[2:750]]
print('top 20 popular words: ')
for i in range(20):
    print(words[i], wordCount[words[i]])
print('top 20 least popular words: ')
for i in reversed(range(1, 20)):
    print(words[-i], wordCount[words[-i]])
words = set(words)
    top 20 popular words:
    like 177915
    good 176169
    story 115759
    would 114727
    great 112150
    time 105217
    really 104703
    much 95481
    even 92317
    get 88493
    see 84373
    first 83358
    well 83019
    movies 77819
    dont 77125
    watch 74726
    bad 70924
    love 67341
    dvd 65663
    could 65145
    top 20 least popular words:
    dumb 5743
```

```
bunch 5734
shots 5730
killing 5727
personal 5715
edition 5690
potential 5677
doctor 5639
thriller 5638
showing 5625
brilliant 5622
solid 5613
leaves 5589
issues 5567
score 5563
acted 5560
musical 5559
mention 5557
worked 5546
```

## Feature Extraction

```
wordId = dict(zip(words, range(len(words))))

def feature(datum):
    feat = [0]*len(words)
    r = ''.join([c for c in datum['reviewText'].lower() if not c in punctuation and (of for w in r.split():
        if w in words:
            feat[wordId[w]] += 1
    return feat

# Get smaller training sample
small_train = random.sample(train, int(len(train)/2))

X = [feature(d) for d in small_train]
train_set = np.asarray(X)
train_set.shape
    (250000, 748)
```

# K-Means Algorithm

```
def calcSqDistances(X, Kmus):
    N,D = X.shape
    K = Kmus.shape[0]
```

```
sqDist = np.zeros((N, K), dtype=np.float32)
    for i in range(N):
        for j in range(K):
            sqDist[i,j] = np.linalg.norm(X[i] - Kmus[j])
    return sqDist
def determineRnk(sqDmat):
    m = np.argmin(sqDmat, axis=1)
    return np.eye(sqDmat.shape[1])[m]
def recalcMus(X, Rnk):
#
      N, D = X.shape
#
      Kmus = np.zeros((len(Rnk[0]), D))
#
      for k in range(len(Rnk[0])):
#
          Rnk_new = np.stack([Rnk[:, k], Rnk[:, k]], axis=1)
#
          Kmus[k,:] = np.sum(np.multiply(X, Rnk_new), axis=0) / np.sum(Rnk[:,k], axis=
      return Kmus
    return (np.divide(X.T.dot(Rnk), np.sum(Rnk, axis=0))).T
def runKMeans(K):
    X = train set
    N, D = X.shape
    Kmus = np.zeros((K, D))
    rand inds = np.random.permutation(N)
    Kmus = X[rand inds[0:K],:]
    maxiters = 1000
    for iter in range(maxiters):
        sqDmat = calcSqDistances(X, Kmus)
        Rnk = determineRnk(sqDmat)
        KmusOld = Kmus
        Kmus = recalcMus(X, Rnk)
        if np.sum(np.abs(KmusOld.reshape((-1, 1)) - Kmus.reshape((-1, 1)))) < 1e-6:
    return Kmus, Rnk
kmus5, Rnk = runKMeans(5)
kmus5
    array([[0.08152832, 0.03834253, 0.22514463, ..., 0.08791874, 0.18458227,
             0.05953182],
            [0.00478015, 0.00590329, 0.03084578, ..., 0.00408564, 0.01022224,
             0.009782751,
            [0.15018022, 0.04365238, 0.4989988, ..., 0.18662395, 0.3360032,
             0.156988391,
            [0.03231975, 0.02803816, 0.12084052, ..., 0.03204955, 0.07029285,
             0.03161308],
```

## Sanity Check and Results

5 Examples of Cluster 1 Reviews

return Rnk

```
cluster_1_idx = []
for i in range(len(Rnk)):
    if np.where(Rnk[i] == 1)[0] == 0:
        cluster_1_idx.append(i)

rev1 = [small_train[i] for i in cluster_1_idx]
rev1[:5]

[{'overall': 3.0,
```

'reviewText': "I saw this movie today, at 1:30 in the morning. And I was very excited. 'Cause I love Jason and his Friday the 13th movies.\nI always liked them, and watched them since I was very little, I got scared like a baby with Jason(Of course that was many years ago. \nThe movie starts with a great slasher beginning. Rowan(Lexa Doig) and Jason(Kane Hodder) are in the same room and only one of them will survive...No. Rowan and Jason get both frozen in time when she suddenly tryes to scape from him(The year is 2010).\nAfter, like 400 years, this two frozen bodys are found by a goup of students. The take them to their ship and bring Rowan back to like with the new technology. \nBut she's not the only one that wakes up, Jason begins his large(Very large, I dare you to try to count them) body count, just after he gets up from bed.\nJason X is a real well done movie. It's visually amazing, the athmosphere it's creepy, and it's a real big step for the Friday the 13th movies. \nActing was pretty good for a Friday movie, I specially enjoyed Mellysa Ade's(Jannesa) and Lexa's(Rowan) performances.\nJannesa was like the funny character, always making sarcastic jokes.\nAnd Rowan is the main character, she's very pretty.\nBut they were

nothing compared to the master of Friday the 13th, that's right, I'm talking

about Kane Hodder. He's perfect for Jason, he was born for becoming Jason. He's a real horror king.\nWhat I really loved here was Crystall lake 3-D version, I mean, it was like and old Friday the 13th movie mixed up with a new one. It was just great. Uber-Jason looked awesome, but I just wanted a longer part with him.\nWhat I really hated here was KM-14, it was so stupid!!!.\nI rate this with 3 stars out of five, 'cause the movie was a little too stupid and sarcastic for a Friday movie. But I have to say that is one of the best sequels in the series.\nThe go like this(From great to bad)\n1-Part 6(Awesome)\n2-Part 4(Tommy rules!)\n3-Part X(Cool)\n4-Part 1(The original, but not the best)\n5-Part 7(Really cool kills)\n6-Part 3(Just liked it 'cause it was 3-D and Jason gets his mask in there)\n7-Part 8(Loved the Manhattan secuence, the rest of the movie sucked)\n8-Part 2(Not bad sequel, not great either)\n9-Part 5(There's no Jason, so ther's no more to say)\n10-Part 9(What a BAD movie)\nJASON X it's a real horror movie for fans, but it's true. If your not a Jason fan, you'll never like this movie. \nBut I did, it's pure fun, I mean it doesn''t make any sense, but It was an hour and a half enterteining flick.\nJASON X\nDirected by J. Isaacs\nStarring: Kane Hodder,\nLexa Doig, Lisa Ryder\nRunning time: 93 minutes\nRated: R\nOrigina: U.S.A.\nYear: 2002\nFINAL GRADE: C+(7)", 'summary': 'JASON IN SPACE? Only good for fans'},

{'overall': 1.0,

'reviewText': "Before I go any further, you may want to know where this is coming from. I am 15, Catholic, home-schooled, and look upon the use of contractions in formal writing as the first sign that a book is bad. Rowling uses a lot of contractions.\n\nI cannot compare the movie to the book, because I have never had the time necessary to waste on it - though it only took me 2 hrs to finish the first two.\n\nThe movie was just stupid - any lover of plot or logic will do well to keep away.\n\nItem #1: under-age wizards are not alowed to use magic away from the school.\n\nThis is a sensible enough rule, but it seems as though a) it is never enforced; or b) an exception can be made for 'the one who lived' (or whatever he is called)\n\nItem #2: Harry is disturbed that his qod-father is after him, and holds him in contempt for being a murder.\n\nCome on!! I mean, his life is always in danger because of Valdemort anyway, so lighten up!! As for not liking his god-father's being a murder, what about Harry's blowing-up his aunt?!? That's attempted murder!\n\nItem #3: Harry's godfather was one of his parents' best friends.\n\nWitches and wizards obviously cannot believe in God. How can these atheistic/agnostic witches have godparents? Why would they even baptize their kids? This brings up a similar point: why do they celebrate Christmas? Wouldn't they more then likely celebrate Winter Solstice or the Feast of Thor or something?\n\nItem #4: a

#### 5 Examples of Cluster 2 Reviews

```
cluster_2_idx = []
for i in range(len(Rnk)):
    if np.where(Rnk[i] == 1)[0] == 1:
        cluster_2_idx.append(i)

rev2 = [small_train[i] for i in cluster_2_idx]
rev2[:5]

[{'overall': 3.0, 'reviewText': 'Great', 'summary': 'Three Stars'},
    {'overall': 1.0,
        'reviewText': "This movie would lead you to believe you can not Kiss before
```

marriage.... We are a Christian Family I've been married to my wife for 43 years... our Children have both been married long a time. one for 16 years the other for llyears My wife and I waited until we got married.. however we did Kiss..... This Movie is Over the TOP....", 'summary': 'Over the Top'}, {'overall': 2.0, 'reviewText': 'This movie has no idea whether its a buddy cop comedy, or a parody of one. At points its serious, and then it is the complete opposite the next minute, to the point of being rediculous. Will Ferell never "goes there" with his classic Anchorman rants, and Mark Wahlberg plays the most annoying character ever. It was just a mess of a movie, and I wanted to give it a low star rating, b/c I was sickened to see 4 stars, and so many positive reviews. Click the dislike button now, you freaking no-lifes. I\'m not gonna respond to hostile comments either, so just to get this out of the way: SCREW YOU.', 'summary': 'A throwaway movie. Waste of time.' }, {'overall': 4.0, 'reviewText': 'Was a decent action movie with a very basic revenge story line. No issues with streaming or audio.', 'summary': 'Four Stars'}, {'overall': 3.0, 'reviewText': 'This time Sylvester Stallone is against a bunch of ax wielding serial killers, with nylons over their faces, who terrorize the LA nights...\n\nWhen a fashion model happens to see the ugly face of a sadistic psychopath (Brian Thompson), Ingrid (Brigitte Nielsen) becomes the main target of the secret "New World" society stopping at nothing to slain her...\n\nLieutenant Marion Cobretti (Stallone), in his gun metal-gray classic Mercury, and armed with guns, knives, grenades, and firearms, is assigned to protect the statuesque blonde...\n\nThe movie is too violent and too bloody and contains one of the most interesting car chase sequences ever filmed...', 'summary': '"I don\'t deal with psychos, I put them away!"'}]

#### 5 Examples of Cluster 3 Reviews

```
cluster_3_idx = []
for i in range(len(Rnk)):
    if np.where(Rnk[i] == 1)[0] == 2:
        cluster_3_idx.append(i)

rev3 = [small_train[i] for i in cluster_3_idx]
rev3[:5]
    {'overall': 2.0,
```

'reviewText': 'If one were to look at Inside on any level other than that of a splatter horror film in the vein of a poor man\'s Lucio Fulci, it would be a sad response to the current state of horror. Inside is a film that seems like it will only appeal to those who think gore— no matter how realistic and well executed it may be— means a film is effective. That would be a falsity. As Inside both lacks in nearly every department aside from conjuring some adamantly

gruesome images, and is the antithesis of both thought and suspense.\n\nGood horror movies for me are not unlike that of suspense or thrillers. In fact they are often much, much better because of their sense of both baroque and dire atmosphere and a feeling of all-bets-are-off bleakness. Whether you are Alfred Hitchcock, Brian De Palma, or Dario Argento you must understand all three of these genres. As Hitchcock put it: "There is no terror in the bang, only in the

anticipation of it. "\n\nInside is all about the bang. Any sense of anticipation, (with the exception of one humorously ironic scene involving a toaster,) is shoddy and amateur: often being telegraphed long before it happens. Suspense and actual tensions should always be a necessity. Gore without this is simply for pussies. As Inside is a film dependent on startling images, much like the aforementioned films of Lucio Fulci. Yet, in the case of effects, although it may have improved on realistic bloodshed, it sadly lacks the best element of that auteur\'s Grindhouse output in which the dreamlike and otherworldly whimsy of films like The Beyond is rather traded for nothing of real artistic merit or value.\n\nThough this writer does not have a problem at all with bloodshed, it must be said that after you see a gore scene once, no matter how well executed, you become desensitized; it\'s just not that effective anymore. It is the bang without a menacing triggerman -- yet, if you have a scene, gore or not in tact, and have an ample amount of suspense built around character development, pacing, tension and attack and release, then you have a film that will withstand multiple viewings.\n\nOther than bloodshed, Inside tries to make a bleak atmosphere through its use of shadows (doesn\'t work, by the way) and its angry lead character. Yet, the film does not earn it, rather the character of the pregnant Sarah (Alysson Paradis) feels as static as a walking dead extra in a Romero zombie film. Though the audience knows she is mad at the world and misses her husband, but it falls into telling not showing character development. Inside "tells" us to feel empathetic for this character and all the victims due to their ordeal, but it does not establish any sort of relationship with any of them of any worth. Although this may seem intentional, especially involving Sarah\'s possible more than friendship with her boss, this sense of mystery certainly doesn't help, rather it just seems lazy. \n\nOf course if one where to really want to they can pick apart many dumb character moves, as well. However, one must know that often you simply won\'t have a movie without them. Horror, especially this genre, utilizes a narrative that relies on mistakes, often as a commentary --- whether intentional or not, it depicts humans for what they are, flawed animals that often make mistakes when drastic times call for drastic measures. Dumb character moves can be forgiven if they resemble the character. It\'s again too bad that Inside doesn\'t have much character depth or layers, which does not exactly help matters when the bad logic comes into play.\n\nThe half-twist comes as braindead and unclear as the rest of the movie -- even negates itself in logic somewhat when Sarah mentions that she was told there were, "No survivors." A scene involving a believed to be dead character that can be compared to something out of a zombie-movie seems more hokey than anything in context to the rest of this otherwise played-straight film. The final shot of Inside, also, rather feels like it has false feelings of grandeur.\n\nIn the end, Inside may have a taboo feel like some of those golden-age horror movies, and possibly can be seen for its no-holds-barred approach with some of its noteasily-shaken images. It\'s too bad the budget\'s load was blown on gore shots than a good script.\n\nScore: 5.0 / 10 (in 0.5 increments)', 'summary': 'Not Truly Terrifying'}]

#### 5 Examples of Cluster 4 Reviews

```
cluster_4_idx = []
for i in range(len(Rnk)):
    if np.where(Rnk[i] == 1)[0] == 3:
        cluster 4 idx.append(i)
```

```
rev4 = [small_train[i] for i in cluster_4_idx]
rev4[:5]
```

[{'overall': 3.0,

'reviewText': 'I\'m sure a lot of people have asked what went wrong with "Twin Peaks: Fire Walks with Me." It adds nothing new to the brilliance of the series and it\'s purpose seems to be to show the audience what kind of person Laura Palmer was, which we already pick up from the series. Story and characters are sacrificed for the purpose of shocking people with Laura Palmer\'s actions, but it just doesn\'t work. Things are vulgar, annoying, and often hard to take. Plus, since we already know who killed Laura Palmer (from the TV series), there is nothing new to be said. In the film is too long for its own good, with the entire 20 odd opening minutes a pointless addition to the rest of the story; it just slows things down and becomes boring. While entertaining at times, especially when Laura Palmer comes into the picture (after the excruciating start), this was extremely disappointing. It seems the whole concept of Twin Peaks has went to David Lynch\'s head a little bit. This being the movie, David Lynch decided to turn up the sexual antics and the bizarre plot.\nAside from that, the setting is breathtaking and bizarre all at once, and Sherly Lee is just superb as Laura Palmer. However, this just doesn\'t do any justice whatsoever to the TV series.',

'summary': 'Watch at your own risk!!!'},
{'overall': 1.0,

'reviewText': 'Thank you to the one star reviewer who gave me a heads up about this. As I result, I will contribute a one-star without watching the movie.\n\nSince Bridesmaids--one of the most disgusting movies I have ever had to stop watching--I have started to watch for this s\*\*\* (so to speak). I wish they would develop a new rating sytem that would warn us. I can handle sex and "foul" language, and I really don\'t need to be warned about them. But fecal scenes--oh, yeah, and vomit scenes--are right up there with violence, as far as I\'m concerned. Could we have a special star system for the gross-out factor?', 'summary': 'More fecal scenes!'},

{'overall': 3.0,

'reviewText': "THE TROLLENBERG TERROR or THE CRAWLING EYE is a British production written by Jimmy Sangster and directed by Quentin Lawrence in 1958.\n\nGiant crawling eyes are hiding in a radioactive fog resting on the slopes of the Trollenberg mountain in Switzerland. When infortunate climbers enter the fog, they are simply decapitated by the eyes. By the tentacles of the eyes, I mean. Meanwhile, a young British girl with telepathic powers is attracted by the mountain. The eyes send two zombies to kill her but she's saved by the hero Forrest Tucker, a United Nations observer.\n\nWell, that's a screenplay, isn't it? A little bit irrational maybe, but, remember, we're in the fifties and strange things happened then in the world like the birth of John Travolta, for instance, so let's be tolerant. The copy presented by Image is superb and the actors are surprisingly good for this kind a production. Recommended if you're a movie lover.\n\nA DVD zone they're here but they are so stupid.",

'summary': "I'M WATCHING YOU"},
{'overall': 2.0,

'reviewText': 'Olympus Has Fallen is one of the worst movies ever.\nThe story is only possible if a multitude of idiotic screw ups ALL occur.\n1. Unidentified aircraft cruises into DC w/o ID and early prevention.\n2. Unidentified aircraft opens fire, likely without sophisticated aiming systems, and takes down 2 fighter planes within seconds.\n3. Unidentified aircraft opens fire, likely without sophisticated aiming systems, and pretty much takes out the whole security team on the roof.\n4. A huge crew of terrorists 1) get pass customs, 2)

none get picked up by CIA/FBI/NSA or what have you, 3) get access to heavy artillery w/o being caught by aforementioned agencies, 4) modify vehicles without notice near DC.\n5. After the aircraft guns down the units on top, the large crew of terrorists with heavy artillery guns down the remaining armed

#### 5 Examples of Cluster 5 Reviews

```
cluster_5_idx = []
for i in range(len(Rnk)):
    if np.where(Rnk[i] == 1)[0] == 4:
        cluster_5_idx.append(i)

rev5 = [small_train[i] for i in cluster_5_idx]
rev5[:5]
```

[{'overall': 4.0, 'reviewText': "Wrestlemania is the biggest event of the year for wrestling fans around the world. We have seen some great matches in Wrestlemania past such as Hogan/Andre(WM3), Hogan/Warrior(WM6) and so many others. So did this PPV live up to the hype or did it sink like the Titanic? Well I'll go down the list of matches match by match. So lets get started: (I'll be rating everything out of 5 stars)\n\nMoney in the Bank was the first match of the night, and while it wasn't the greatest Money in the bank it was still very enjoyable and as always included tons of crazy spots. My only gripe was even though I am a fan of CM Punk I would have liked to have seen Shelton or MVP win it also it could have been a tad bit longer. 4 out of 5\n\n25 Diva Battle Royal was very bad. First of all the divas didn't have a proper entrance, they all just came out together while Kid Rock played. So, you didn't know who was all in the match except for a few divas here and there. I found out the next day that former WWE diva Joy Giovani was in it, but I would of never known that because well we were never told or she was never seen. So, that was a huge disappointment, as the match which could have been decent ended up being one giant mess. 1 out of 5\n\nJericho vs The Legends was ok, at first it was boring and slow and then when it came down to just Jericho and Steamboat it took off. Steamboat for being up there in years looks like he did years ago. He looks like he can still go and put on a good 20 min match. I hope at Backlash Jericho faces Steamboat in a one on one match. 3 out of 5\n\nJeff Hardy vs. Matt Hardy was better then I expected, I remember when these two quys feuded years ago and their matches weren't that good. Well this match was a lot better then I had thought but still could have been better then it was. The double table spots was cool as was the Jeff jumping over the ladder and missing Matt. The ending with the Twist of Fate using the chair was sick and was a great ending to this match. I doubt this feud is over so lets hope these two can continue to put on great matches. 3 out of 5\n\nJBL (c) vs. Rey Mysterio for the IC title was a joke, literally maybe thats why Rey came out as the joker. I don't have much to say about this match as it was only 21 seconds long. So all im saying is it sucked. 1 out of 5 stars\n\nThe Undertaker vs. Shawn Michaels, OMG this match was great and I mean great. This should have been the main event. I t was so good it made the 2 world title matches look like crap compared to it and it made it hard to follow it up. The match is worth buying the event just for this match alone. The spot when Taker flew over the top rope and on to the camera man was sick as hell, I thought Taker and the cameraman were dead. The ending with both guys hitting their finishers and only getting a two count was awesome this match has to be Taker greatest Mania match ever and also one of the best matches we have seen in WWE

in a while. 5 out 5\n\nAs for the last two matches, for the Edge (c) vs. John Cena vs. Big Show it was boring and I really can't remember all that happend in the match except for the spots where Cena picked up Big Show and Edge at the same time and Big SHow stuckn the ropes, and Cenas entrance with 300 Cena clones. Cena winning was predictable as Big Show has only won 1 Wrestlemania match in his life. 2 out 5\n\nTriple H vs Randy Orton was a complete and total let down, they had this really good feud going and had it very personal with great segments leading up to the match. The match in no way lived up to its expectations and as I said thats the reason why the Taker/HBK match should have been last it was just too good to have any other match follow it up. The ending was dumb and nothing special really happened in this match. I thought we were going to have someone come down and possibly help Orton win and turn on HHH but nothing. 2.5 out of 5\n\nOverall, it was pretty good, match of the night was Taker/HBK and while there were some matches that sucked, it was Mania and thats good enough for me. [...] I still enjoyed it and I'm still glad I ordered it. The DVD will be out on May 19, according to Amazon as its up for Pre-order, so I'll be picking it up and recommend that you do to.\n\nReview brought to you by Dreadfulentertainment.com",

'cummary'. 'Not the hest but still a good event'l

#### Mean Rating of Each Cluster

```
cluster_1_ratings = [d['overall'] for d in rev1]
cluster_2_ratings = [d['overall'] for d in rev2]
cluster_3_ratings = [d['overall'] for d in rev3]
cluster_4_ratings = [d['overall'] for d in rev4]
cluster_5_ratings = [d['overall'] for d in rev5]

np.mean(cluster_1_ratings), np.mean(cluster_2_ratings), np.mean(cluster_3_ratings), np.
```

#### Mode Rating of Each Cluster