```
In [1]: # We want to conduct sentiment analysis on the comment data
        # We would be interested in the overall sentiment regarding different dominant
        # Import the required packages
        import pandas as pd
        import numpy as np
        import nltk
        from nltk.tokenize import word tokenize
        from nltk.stem.porter import PorterStemmer
        from nltk.corpus import stopwords
        import string
        from nltk.stem import WordNetLemmatizer
        from nltk.corpus import wordnet
        from wordcloud import WordCloud
        from sklearn.feature_extraction.text import TfidfVectorizer
        import re
        import gensim
        from gensim.models import Phrases
        from gensim.models.phrases import Phraser
        from gensim.utils import simple preprocess
        import matplotlib.pyplot as plt
        from gensim.models.phrases import Phrases, ENGLISH CONNECTOR WORDS
```

```
In [2]: # Load in the data and preprocess it ~ we generally expect that the stopwords
# encountered in the past, we will also be using vader lexicon ~ suitable for
# perhaps account for abbreviations
txt_df = pd.read_csv(r'C:\Users\huang\OneDrive\Documents\ECO481\text_df.csv')
# Create re function to remove punctuations
regex = re.compile('[%s]' % re.escape("[-,\.!?:'â€[™]"))
def test_re(s): # From Vinko's solution, with fix.
    return regex.sub("", s)
# perhaps we also need to drop duplicated columns, this is dependent on subset
txt_df.drop_duplicates(subset = ['User ID', 'Comments'], inplace = True)
txt_df.reset_index(inplace = True)
txt_df.head()
```

#### Out[2]:

	index	Unnamed: 0	User ID	Comments	Upvotes
0	0	0	jbe3ase	This seems to address a lot of the feedback so	250
1	1	1	jbe3kcn	Cruor's Vow would work with all combat styles	97
2	2	2	jbe8g6l	During the initial proposal I was like: hmm, f	6
3	3	3	jbfo38h	Given the power of the book, I don't think we	1
4	4	4	jbe3exk	For Glacies' Vow, is the bonus magic damage de	72

```
In [3]: # preprocess and then create bigram models:
        # this is a bit challenging ~ some of these words are outright filler like nam
        # these refer to a neutral object so to speak, like you could say bad/qood thi
        # comments to words
        def comments to words(comments):
            11 = []
            for comment in comments:
                11.append(gensim.utils.simple preprocess(str(comment), deacc=True))
            return 11
        txt_words = comments_to_words(txt_df['Comments'])
        # Set up bigrams
        bigram = Phrases(txt words, min count=1, threshold=2, delimiter=' ')
        bigram phraser = Phraser(bigram)
        print(bigram phraser)
        ps = PorterStemmer()
        stop = set(stopwords.words('english'))
        # Now get me some punctuations
        punctuations = set(string.punctuation)
        lemmatizer = WordNetLemmatizer()
        # Add in the abbreviations
        stop.add('tbh')
        stop.add('imo')
```

FrozenPhrases<5331 phrases, min\_count=1, threshold=2>

```
In [4]: # Now we process the texts:
        # Should I pre-process the text? Probably. Remove punctuations and stopwords..
        # perhaps put headlines into lower-case as well...
        ps = PorterStemmer()
        stop = set(stopwords.words('english'))
        # Now get me some punctuations
        punctuations = set(string.punctuation)
        lemmatizer = WordNetLemmatizer()
        # Apply some early pre-processing
        comment list = []
        for item in txt_df['Comments']:
            comment_list.append(test_re(item).lower())
        txt df['Comments Simplified'] = comment list
        # Attach tags to words ~ hopefully lets us lemmatize more easily
        def get wordnet pos(word):
            """Map POS tag to first character lemmatize() accepts"""
            tag = nltk.pos_tag([word])[0][1][0].upper()
            tag_dict = {"J": wordnet.ADJ,
                         "N": wordnet.NOUN,
                        "V": wordnet.VERB,
                        "R": wordnet.ADV}
            return tag_dict.get(tag, wordnet.NOUN)
        # How will it handle names?
```

```
In [5]: # preprocess function
        def preprocess(text):
            # STOPWORD REMOVAL
            doc1 = text.lower()
            doc2 = doc1.split()
            doc3 = [val for val in doc2 if val not in stop]
            doc4 = " ".join(doc3)
            # print(doc4)
            # PUNCTUATION REMOVAL
            # would this be like 'contains', that is if it contains punctuations, then
            doc5 = [val for val in doc4 if val not in punctuations]
            # print(doc5)
            doc6 = "".join(doc5)
            # print(doc6)
            # Lemmatize instead, more useful than stem
            # but we need to determine if noun, verb or whatever
            words = word tokenize(doc6)
            # print(words)
            # associate with right tags
            tags = [get wordnet pos(word) for word in words]
            ind = 0
            lem_final = []
            while ind in range(len(words)):
                lem = lemmatizer.lemmatize(words[ind], tags[ind])
                lem final.append(lem)
                ind += 1
            final = " ".join(lem final)
            return final
        # Make bi-grams and also the processed comments
        processed list = []
        bigram_list = []
        for item in txt df['Comments Simplified']:
            processed list.append(preprocess(item))
            bigram_list.append(bigram_phraser[preprocess(item).split()])
        txt_df['Processed Comments'] = processed_list
        txt_df['Bi-grams'] = bigram_list
        txt df.head()
```

# Out[5]:

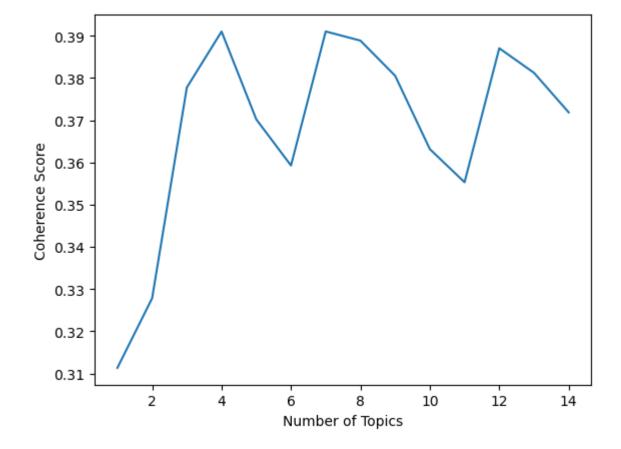
	index	Unnamed: 0	User ID	Comments	Upvotes	Comments Simplified	Processed Comments	Bi-grams
0	0	0	jbe3ase	This seems to address a lot of the feedback so	250	this seems to address a lot of the feedback so	seem address lot feedback far really nice chan	[seem, address, lot, feedback, far, really, ni
1	1	1	jbe3kcn	Cruor's Vow would work with all combat styles 	97	cruor's vow would work with all combat styles 	cruors vow would work combat style yeah ill up	[cruors vow, would work, combat style, yeah, i
2	2	2	jbe8g6l	During the initial proposal I was like: hmm, f	6	during the initial proposal i was like hmm fin	initial proposal like hmm fine whatever iterat	[initial proposal, like, hmm, fine, whatever,
3	3	3	jbfo38h	Given the power of the book, I don't think we	1	given the power of the book i dont think we ca	give power book dont think conclude itll feel	[give, power, book, dont think, conclude, itll
4	4	4	jbe3exk	For Glacies' Vow, is the bonus magic damage de	72	for glacies' vow is the bonus magic damage dea	glacies vow bonus magic damage dealt secondary	[glacies vow, bonus magic, damage dealt, secon

```
In [6]: # We now aim to find the optimal # of topics
        import gensim
        from gensim.models import CoherenceModel
        # Create dictionary for id and words
        # print(nyt_df['Processed Headlines'].type())
        # add in column that is processed split words
        split words = []
        for item in txt df['Processed Comments']:
            split words.append(item.split())
        txt_df['Processed Comments List'] = split_words
        # We used a bag of words model ~ look to incorporate TF IDF in the future
        # id2word ~ associate ID to words in the headlines, corpus should be the headl
        # times they appear?
        id2word1 = gensim.corpora.Dictionary(txt df['Bi-grams'])
        corpus1 = [id2word1.doc2bow(text) for text in txt df['Bi-grams']]
        # Calculate optimal number of topics
        number of topics = []
        coherence_score = []
        lda_list = []
        for i in range(1, 15, 1):
            lda model = gensim.models.ldamodel.LdaModel(corpus = corpus1, id2word = id
                                                         num topics = i)
            lda list.append(lda model)
            coherence_model_lda = CoherenceModel(model = lda_model, texts = txt_df['Bi
                                                 coherence = 'c v')
            coherence lda = coherence model lda.get coherence()
            number of topics.append(i)
            coherence_score.append(coherence_lda)
        # store in a df
        topic_coherence = pd.DataFrame(list(zip(number_of_topics, coherence_score)), d
        topic coherence
```

#### Out[6]:

	Number of Topics	Coherence Score
0	1	0.311344
1	2	0.327839
2	3	0.377772
3	4	0.391026
4	5	0.370221
5	6	0.359285
6	7	0.391044
7	8	0.388873
8	9	0.380548
9	10	0.363165
10	11	0.355321
11	12	0.387064
12	13	0.381246
13	14	0.371869

```
In [7]: # graph coherence score:
    x_ax = topic_coherence['Number of Topics']
    y_ax = topic_coherence['Coherence Score']
    plt.figure()
    plt.plot(x_ax, y_ax)
    xl=plt.xlabel('Number of Topics')
    yp=plt.ylabel('Coherence Score')
    # significant trough at k = 5
```



In [8]: # we are using 4 topics
all\_topic\_dists = [lda\_list[3].get\_document\_topics(d) for d in corpus1]
 txt\_df['Dominant Topic'] = [max(d, key = lambda i: i[1])[0] for d in all\_topic
 txt\_df.head()

### Out[8]:

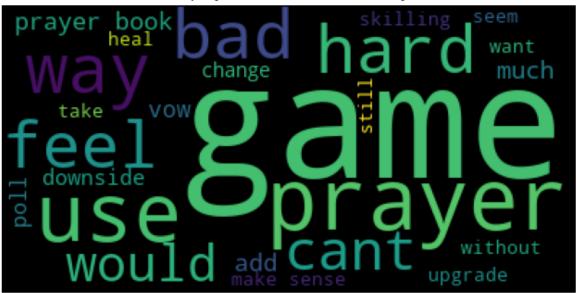
	index	Unnamed: 0	User ID	Comments	Upvotes	Comments Simplified	Processed Comments	Bi-grams	Processed Comments Lis
0	0	0	jbe3ase	This seems to address a lot of the feedback so	250	this seems to address a lot of the feedback so	seem address lot feedback far really nice chan	[seem, address, lot, feedback, far, really, ni	[seem address lot feedback far, really ni
1	1	1	jbe3kcn	Cruor's Vow would work with all combat styles	97	cruor's vow would work with all combat styles	cruors vow would work combat style yeah ill up	[cruors vow, would work, combat style, yeah, i	[cruors vow would work combat style yeah
2	2	2	jbe8g6l	During the initial proposal I was like: hmm, f	6	during the initial proposal i was like hmm fin	initial proposal like hmm fine whatever iterat	[initial proposal, like, hmm, fine, whatever,	[initial proposal like, hmm fine whatever,
3	3	3	jbfo38h	Given the power of the book, I don't think we	1	given the power of the book i dont think we ca	give power book dont think conclude itll feel	[give, power, book, dont think, conclude, itll	[give power book, dont think conclude itl
4	4	4	jbe3exk	For Glacies' Vow, is the bonus magic damage de	72	for glacies' vow is the bonus magic damage dea	glacies vow bonus magic damage dealt secondary	[glacies vow, bonus magic, damage dealt, secon	[glacies vow bonus magic damage dealt, se
4									<b>&gt;</b>

```
In [9]: # sentiment scores
        from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
        def sentiment scores(sentence):
            # Create a SentimentIntensityAnalyzer object.
            sid_obj = SentimentIntensityAnalyzer()
            # polarity scores method of SentimentIntensityAnalyzer
            # object gives a sentiment dictionary.
            # which contains pos, neg, neu, and compound scores.
            sentiment dict = sid obj.polarity scores(sentence)
            neg_score = sentiment_dict['neg']
            neu score = sentiment dict['neu']
            pos score = sentiment dict['pos']
            compound_score = sentiment_dict['compound']
            1 = [neg_score, neu_score, pos_score, compound_score]
            return 1
        neg_1 = []
        neu 1 = []
        pos 1 = []
        com 1 = []
        for item in txt df['Processed Comments']:
            scores_1 = sentiment_scores(item)
            neg_l.append(scores_l[0])
            neu l.append(scores l[1])
            pos 1.append(scores 1[2])
            com_l.append(scores_1[3])
        txt_df['Negative Score'] = neg_1
        txt_df['Neutral Score'] = neu_l
        txt_df['Positive Score'] = pos_1
        txt_df['Compound Score'] = com_1
        txt df.head()
```

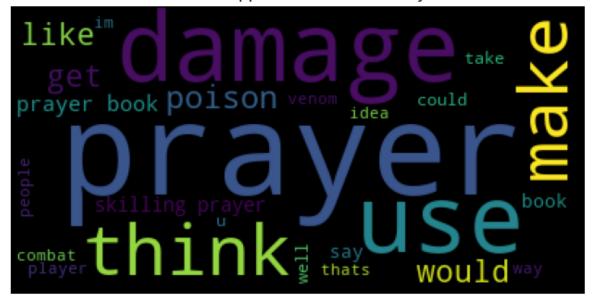
# Out[9]:

	index	Unnamed: 0	User ID	Comments	Upvotes	Comments Simplified	Processed Comments	Bi-grams	Processed Comments Lis
0	0	0	jbe3ase	This seems to address a lot of the feedback so	250	this seems to address a lot of the feedback so	seem address lot feedback far really nice chan	[seem, address, lot, feedback, far, really, ni	[seem address lot feedback far, really ni
1	1	1	jbe3kcn	Cruor's Vow would work with all combat styles	97	cruor's vow would work with all combat styles	cruors vow would work combat style yeah ill up	[cruors vow, would work, combat style, yeah, i	[cruors vow would work combat style yeah
2	2	2	jbe8g6l	During the initial proposal I was like: hmm, f	6	during the initial proposal i was like hmm fin	initial proposal like hmm fine whatever iterat	[initial proposal, like, hmm, fine, whatever,	[initial proposal like, hmm fine whatever,
3	3	3	jbfo38h	Given the power of the book, I don't think we	1	given the power of the book i dont think we ca	give power book dont think conclude itll feel	[give, power, book, dont think, conclude, itll	[give power book, dont think conclude itl
4	4	4	jbe3exk	For Glacies' Vow, is the bonus magic damage de	72	for glacies' vow is the bonus magic damage dea	glacies vow bonus magic damage dealt secondary	[glacies vow, bonus magic, damage dealt, secon	[glacies vow bonus magic damage dealt, se
4									<b>&gt;</b>

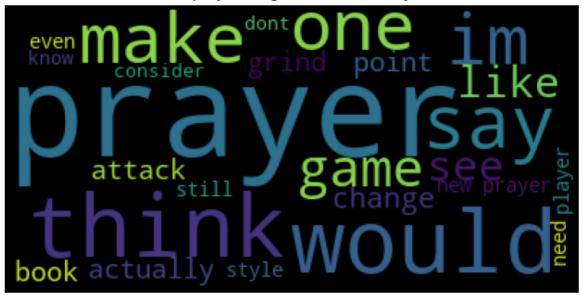
# Gameplay Downsides to New Prayers



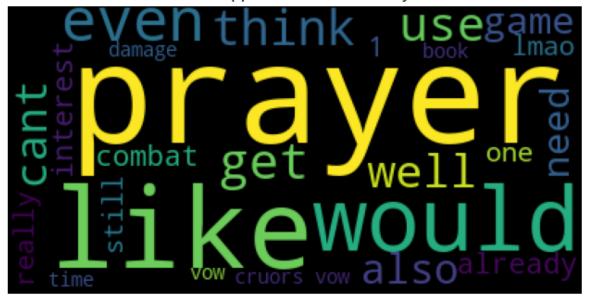
#### Defensive Applications for New Prayers



### Gameplay Changes with New Prayers



Combat Application for New Prayers



In [17]: # group by topics, and take the average of the compound scores
 aggregate\_sentiment\_df = txt\_df[['Dominant Topic', 'Negative Score', 'Neutral
 aggregate\_sentiment\_df.index = topic\_list
 aggregate\_sentiment\_df

### Out[17]:

	Negative Score	Neutral Score	Positive Score	Compound Score
Gameplay Downsides to New Prayers	0.125570	0.640641	0.215581	0.221394
Defensive Applications for New Prayers	0.134134	0.697675	0.168188	0.061254
Gameplay Changes with New Prayers	0.154706	0.646766	0.198525	0.105041
Combat Application for New Prayers	0.162159	0.642311	0.195533	0.049417

In [ ]: # perhaps look at the wordclouds?