An interesting yet simple algorithmic approach to playing the popular 'Wordle' game recently bought by the NYT.

If nothing else, you will learn a lot about 5 letter words.

You can play the game here - https://www.powerlanguage.co.uk/wordle/

First, let us get a list of all possible 5 letter words.

I found this one on github - https://github.com/charlesreid1/five-letter-words/blob/master/sgb-words.txt

Corpus has 5757 words.

Preview: ['imply', 'didot', 'agars', 'aloft', 'vigil', 'spout', 'pasta', 'prawn', 'molly', 'snoop', 'hapax', 'tamer', 'umbel', 'yokes', 'bores', 'reign', 'taros', 'ba wdy', 'naiad', 'temps', 'older', 'halve', 'corer', 'youse', 'snack', 'manic', 'bebu g', 'color', 'adobe', 'weals', 'rubes', 'didos', 'gangs', 'sugar', 'hitch', 'feuar', 'sides', 'chile', 'plots', 'venue', 'crept', 'grata', 'mange', 'velds', 'worst', 'ha ppy', 'zests', 'sines', 'cupid', 'dotes', 'poohs', 'taxol', 'tided', 'rusks', 'shor e', 'sprig', 'inept', 'washy', 'truss', 'serfs', 'bikes', 'whipt', 'darer', 'molar', 'bough', 'scarp', 'phyla', 'hills', 'saxes', 'faddy', 'brawn', 'hyper', 'dicks', 'lo cks', 'gazed', 'feast', 'inode', 'bolas', 'gouts', 'polly', 'sagas', 'manse', 'dowe l', 'wheat', 'falls', 'graze', 'whiff', 'duped', 'filmy', 'bocci', 'mints', 'blunt', 'games', 'rends', 'scion', 'intro', 'might', 'clogs', 'pasts', 'frees']

The essense of the algorithm -

We will use simple probabilities to make guesses which eliminate maximum number of potential next guesses

Approach 1 - Don't care about positions

We find the frequency of occourence of each letter in our list of words. We only check if it appears somewhere in the word, not at any particular position.

```
In [13]:
# Dictionary to store the frequencies
frequencies = {}

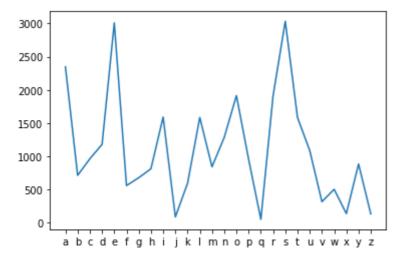
for word in words:
    for letter in word:
        # If letter already exists, increment
        if letter in frequencies:
```

```
frequencies[letter] += 1
# Else, set count to 1
else:
    frequencies[letter] = 1

print(frequencies)
```

```
{'w': 505, 'h': 814, 'i': 1592, 'c': 964, 't': 1585, 'e': 3009, 'r': 1910, 'a': 234 8, 'b': 715, 'o': 1915, 'u': 1089, 'l': 1586, 'd': 1181, 's': 3033, 'f': 561, 'g': 6 79, 'n': 1285, 'k': 596, 'y': 886, 'p': 955, 'v': 318, 'm': 843, 'q': 53, 'j': 89, 'x': 139, 'z': 135}
```

The above dictionary means that the letter 's' occours in 3033 words out of our 5757 words, 'w' occours in 505, etc. So, we guess 'S'



We generate a score based on how 'disruptive' each letter in the word is Which letters should we guess?

We guess the words which divide the list the most

Think about it -

- If we guess letters which which are in only 1% (no such actual case but imagine) of the words, there's a high chance we get a GREY and we have eliminated only 1% of possible options
- If we guess letters which are in 99% (no such actual case but imagine) of the words, there's a high chance we get a YELLOW and we have eliminated only 1% of possible options
- So we give higher scores to letters whose frequencies are closest to half (50%) of the number of total words!

```
In [26]: # Generating distances from halfway as TOTAL NO. OF WORDS - FREQUENCY OF WORDS IN WH
```

```
scores_algorithm_1 = {}
for letter in frequencies:
    scores_algorithm_1[letter] = len(words) - frequencies[letter]

# Sorting letters based on their distance from the half
print({k: v for k, v in sorted(scores_algorithm_1.items(), key=lambda item: item[1])

{'s': 2724, 'e': 2748, 'a': 3409, 'o': 3842, 'r': 3847, 'i': 4165, 'l': 4171, 't': 4
172, 'n': 4472, 'd': 4576, 'u': 4668, 'c': 4793, 'p': 4802, 'y': 4871, 'm': 4914,
'h': 4943, 'b': 5042, 'g': 5078, 'k': 5161, 'f': 5196, 'w': 5252, 'v': 5439, 'x': 56
18, 'z': 5622, 'j': 5668, 'q': 5704}
```

Does this mean 'seaor' is the best guess according to Algorithm 1?

Well yes... but the game only allows valid words. So we have to use these letter scores to generate word scores for all valid words

Lesser the score, better the word.

```
In [118...
          # Score based on overall frequency
          word_scores_algorithm_1 = {}
          for word in words:
              scores = scores algorithm 1.copy()
              word score = 0
              for letter in word:
                  word_score += scores[letter]
                  0.00
                  THIS PART IS IMPORTANT!
                  Currently the algorithm just tries to fit in as much high scoring letters as
                  But we want to penalize repetition of letters because repeated letters give
                   So, once we have used a letter, we set its score to the max score value to p
                  scores[letter] = len(words)
              word_scores_algorithm_1[word] = word_score
          # Sorting the scored words
          _sorted_scores = {k: v for k, v in sorted(word_scores_algorithm_1.items(), key=lambd
          list(_sorted_scores.items())[:15]
Out[118... [('arose', 16570),
          ('raise', 16893),
           ('arise', 16893),
           ('aloes', 16894),
           ('stoae', 16895),
           ('laser', 16899),
           ('earls', 16899),
           ('reals', 16899),
           ('tears', 16900),
           ('rates', 16900),
           ('stare', 16900),
           ('aster', 16900),
           ('tares', 16900),
           ('snare', 17200),
           ('earns', 17200)]
In [117...
          Let's just normalize the above scores to be between 1-100
          We will use the simplest, most common way to normalize - min/max normalization
          max score algorithm 1 = word scores algorithm 1[max(word scores algorithm 1, key=wor
          min score algorithm 1 = word scores algorithm 1[min(word scores algorithm 1, key=wor
          normalized_scores_algorithm_1 = {}
```

Results tell us 'arose' is the best word to guess

And this is perfect because it has the same letters as 'seaor' but is a valid word. So our scoring logic worked really well!

Pros of algorithm 1:

Penalizes Repetitions

Cons of algorithm 1:

Does not care about position of letters within the word

So, we make algorithm 2 to take this into account

```
In [48]:
          We find the frequency - not overall, but in each position
          position_to_letter_mapping = {
              0: '1st Letter',
              1 : '2nd Letter'
              2 : '3rd Letter'
              3 : '4th Letter',
              4 : '5th Letter'
          }
          frequency_at_positions = {}
          for position in range(5):
              frequency_at_positions[position] = {}
          for word in words:
              for position in range(5):
                  if word[position] in frequency_at_positions[position]:
                       frequency_at_positions[position][word[position]] += 1
```

```
frequency of letters as the 1st Letter:
{'w': 228, 't': 376, 'a': 296, 'o': 108, 'c': 440, 'f': 318, 'r': 268, 'y': 47, 'p':
386, 's': 724, 'g': 279, 'e': 129, 'n': 118, 'u': 75, 'm': 298, 'h': 239, 'b': 432,
'l': 271, 'k': 91, 'v': 109, 'q': 39, 'd': 311, 'i': 74, 'j': 73, 'z': 24, 'x': 4}
frequency of letters as the 2nd Letter:
{'h': 271, 'b': 32, 'o': 911, 't': 122, 'r': 456, 'i': 673, 'a': 930, 'f': 12, 'e':
660, 'l': 360, 'g': 24, 'v': 27, 'm': 71, 'n': 168, 's': 40, 'u': 534, 'p': 113,
'd': 43, 'c': 82, 'q': 10, 'w': 81, 'y': 65, 'x': 33, 'k': 29, 'z': 6, 'j': 4}
frequency of letters as the 3rd Letter:
{'i': 516, 'e': 397, 'o': 484, 'u': 313, 'h': 39, 'r': 475, 't': 280, 'g': 139, 'a':
605, 'v': 121, 'd': 178, 'l': 388, 'k': 90, 'n': 410, 'p': 169, 'm': 209, 's': 248,
'b': 128, 'w': 98, 'y': 68, 'c': 184, 'j': 8, 'x': 67, 'f': 87, 'z': 52, 'q': 4}
frequency of letters as the 4th Letter:
{'c': 210, 'r': 310, 'i': 284, 'u': 154, 'l': 365, 's': 257, 'e': 1228, 'd': 218,
't': 447, 'h': 73, 'n': 386, 'a': 339, 'o': 262, 'g': 176, 'v': 61, 'w': 70, 'm': 18
8, 'k': 243, 'b': 99, 'p': 196, 'f': 100, 'y': 41, 'z': 41, 'j': 4, 'x': 5}
frequency of letters as the 5th Letter:
{'h': 192, 'e': 595, 'r': 401, 't': 360, 'd': 431, 's': 1764, 'k': 143, 'n': 203,
'l': 202, 'y': 665, 'w': 28, 'g': 61, 'c': 48, 'p': 91, 'o': 150, 'f': 44, 'a': 178,
'm': 77, 'b': 24, 'x': 30, 'i': 45, 'u': 13, 'z': 12}
```

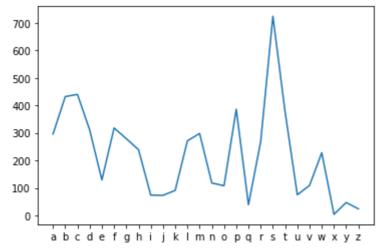
In [49]:

```
# Plotting the same
"""

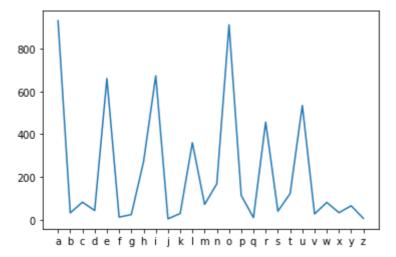
Visualizing the data above just for the lulz
"""

for position in range(5):
    print('frequency of letters as the {}:'.format(position_to_letter_mapping[positilists = sorted(frequency_at_positions[position].items()) # sort alphabetically
    x, y = zip(*lists)
    plt.plot(x, y)
    plt.show()
```

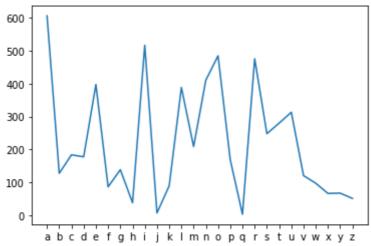
frequency of letters as the 1st Letter:



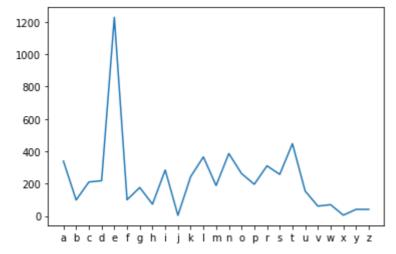
frequency of letters as the 2nd Letter:



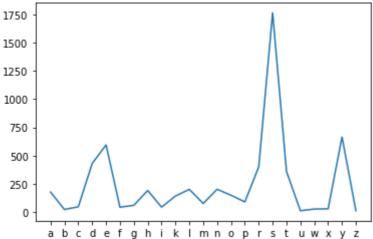
frequency of letters as the 3rd Letter:



frequency of letters as the 4th Letter:



frequency of letters as the 5th Letter:



```
In [116...
          We know we have to score the words to get the best valid guess
          # score based on positional
          word_scores_algorithm_2 = {}
          for word in words:
              word_score = 0
               position = 0
               for letter in word:
                   scores = frequency_at_positions[position]
                   word_score += len(words) - scores[letter]
                   position += 1
               word_scores_algorithm_2[word] = word_score
           _sorted_scores = {k: v for k, v in sorted(word_scores_algorithm_2.items(), key=lambd
          list(_sorted_scores.items())[:15]
         [('sores', 23683),
Out[116...
           ('sales', 23751),
           ('soles', 23770),
           ('sates', 23859),
           ('sires', 23921),
           ('cares', 23948),
           ('bares', 23956),
           ('cores', 23967),
           ('bores', 23975),
           ('sines', 23986),
           ('sages', 24000),
           ('pares', 24002),
           ('tares', 24012),
           ('canes', 24013),
           ('saves', 24018)]
In [115...
          Let's just normalize the above scores as we did with algorithm 1
          max score algorithm 2 = word scores algorithm 2[max(word scores algorithm 2, key=wor
          min_score_algorithm_2 = word_scores_algorithm_2[min(word_scores_algorithm_2, key=word_scores_algorithm_2)
          normalized_scores_algorithm_2 = {}
          for word in words:
               normalized scores algorithm 2[word] = 100 - (word scores algorithm 2[word] - min
                           max_score_algorithm_2 - min_score_algorithm_2)
```

```
_sorted_scores = {k: v for k, v in sorted(normalized_scores_algorithm_2.items(), key
          list(_sorted_scores.items())[:15]
Out[115... [('sores', 100.0),
          ('sales', 99.9851170934559),
           ('soles', 99.98095863427446),
           ('sates', 99.96147953600351),
           ('sires', 99.94790982709564),
           ('cares', 99.94200043773255),
           ('bares', 99.94024950755089),
           ('cores', 99.9378419785511),
           ('bores', 99.93609104836945),
           ('sines', 99.93368351936967),
           ('sages', 99.93061939155176),
           ('pares', 99.93018165900635),
           ('tares', 99.92799299627927),
           ('canes', 99.92777413000657),
           ('saves', 99.92667979864304)]
```

So according to algorithm 2, 'sores' is the most disruptive word because it has letters which are most frequent (closest to half of the number of words) at that particular position

Pros of algorithm 2:

- Cares about position of letters within the word
 Cons of algorithm 2:
- Does not penalize repetitions

```
In [114...
          Since both have pros and cons, what do we do?
          Simple, take the average
          normalized_scores = {}
          for word in words:
              normalized_scores[word] = (normalized_scores_algorithm_1[word] + normalized_scor
           sorted scores = {k: v for k, v in sorted(normalized scores.items(), key=lambda item
          list(_sorted_scores.items())[:15]
Out[114... [('tares', 99.9469774161386),
          ('cares', 99.92195431891784),
           ('tales', 99.92074708912936),
           ('dares', 99.91902878006829),
           ('pares', 99.9155807727729),
           ('rates', 99.91381917582343),
           ('lores', 99.91112818170213),
           ('bares', 99.90823718286259),
           ('nares', 99.90327177423114),
           ('roles', 99.90127919943029),
           ('mares', 99.90017448348857),
           ('cores', 99.89754399082274),
           ('lanes', 99.8961922460838),
           ('dales', 99.89279845305906),
           ('hares', 99.89222231214666)]
```

Best starting word as per combined algorithm is 'tares'

Using this holy grail of information

So we did this for a list of all words, if we want to do it for a list of filtered words, all the logic remains the same

```
In [101...
          .....
          There is literally nothing new in this code
          It is just the code used above mushed into a single function so it can be called for
          def get best word(words):
              # ALGO 1
              scores_algorithm_1 = {}
              for letter in frequencies:
                  scores_algorithm_1[letter] = len(words) - frequencies[letter]
              word_scores_algorithm_1 = {}
              for word in words:
                  scores = scores_algorithm_1.copy()
                  word score = 0
                  for letter in word:
                      word_score += scores[letter]
                      scores[letter] = len(words)
                  word_scores_algorithm_1[word] = word_score
              max_score_algorithm_1 = word_scores_algorithm_1[max(word_scores_algorithm_1, key
              min_score_algorithm_1 = word_scores_algorithm_1[min(word_scores_algorithm_1, key
              normalized_scores_algorithm_1 = {}
              for word in words:
                  normalized_scores_algorithm_1[word] = round(100 - (word_scores_algorithm_1[w
                          max_score_algorithm_1 - min_score_algorithm_1), 3)
              sorted_scores_algorithm_1 = {k: v for k, v in
                                            sorted(normalized_scores_algorithm_1.items(), key=1
              print('Best guesses according to Algorithm 1:')
              print(list(sorted_scores_algorithm_1.items())[:5])
              # ALGO 2
              frequency_at_positions = {}
              for position in range(5):
                  frequency_at_positions[position] = {}
              for word in words:
                  for position in range(5):
                      if word[position] in frequency_at_positions[position]:
                          frequency_at_positions[position][word[position]] += 1
                      else:
                          frequency_at_positions[position][word[position]] = 1
              word_scores_algorithm_2 = {}
              for word in words:
                  word_score = 0
                  position = 0
                  for letter in word:
                      scores = frequency_at_positions[position]
                      word_score += len(words) - scores[letter]
                      position += 1
                  word_scores_algorithm_2[word] = word_score
              max score algorithm 2 = word scores algorithm 2[max(word scores algorithm 2, key
              min_score_algorithm_2 = word_scores_algorithm_2[min(word_scores_algorithm_2, key
              normalized_scores_algorithm_2 = {}
              for word in words:
                  normalized_scores_algorithm_2[word] = round(100 - (word_scores_algorithm_2[w
```

```
In [110...
          WE MAKE A SYSTEM TO PLAY THE GAME, i.e. FILTER OUT WORDS BASED ON THE RULES AND FIND
          0.00
          .....
          I could write some code to read the webpage and get the output
          But I am not going to do that because that's not the point
          Just manually update the ouput of each guess in the dictionary 'guesses' below
          .....
          GREEN - 2
          YELLOW - 1
          GRAY - 0
          .....
          Make the guess suggested by the algorithm and update the results as per the legend a
          You get it. Or atleast I hope you do.
          Keep 'guesses' empty when you start and just run this cell.
          Once you get the results, put the key as the word you guessed
          If the result is GREY GREEN GREY YELLOW YELLOW just put the value as 02011
          Repeat this for each guess you have to make!
          # Starting
          guesses = {
          }
          # After 3 guesses it will look like
          guesses = {
              'tares': '00100',
              'irony': '01100',
              'dolor': '01022'
          }
          def guess(guesses):
              filtered_words = words
              print('{} possible words remain!'.format(len(filtered_words)))
              get_best_word(filtered_words)
              for guess in guesses:
                  print()
                   for letter_index in range(len(guess)):
                       letter = guess[letter_index]
```

```
result = guesses[guess][letter_index]
   if result == '2':
        filtered_words = [word for word in filtered_words if word[letter_ind
   if result == '1':
        filtered words = [word for word in filtered words if (letter in word
   if result == '0':
        filtered_words = [word for word in filtered_words if letter not in w
print('{} possible words remain!'.format(len(filtered_words)))
if len(filtered_words) == 0:
   print('Filtered out all words. Something\'s fishy.')
elif len(filtered words) == 1:
   print('Guessed the word. It\'s \'{}\''.format(filtered words[0]))
elif len(filtered_words) == 2:
   print('Its either \'{}\' or \'{}\''.format(filtered_words[0], filtered_w
else:
   get_best_word(filtered_words)
```

Let's play today's wordle! (#236 when I'm making this)

```
In [111...
          YOU CAN LISTEN TO ALGORITHM 1 OR 2 AS WELL, BUT I'M GOING TO LISTEN TO THE COMBINED
          ALL OF THEM WILL WORK THOUGH
          # After 0 guesses
          guesses = {}
          # After 1 guesses
          guesses = {
               'tares': '02011'
          guess(guesses)
          # After 2 guesses
          "WE ALREADY DAMN FOUND THE WORD IN 2 GUESSES."
          # Okay it's smart and all but we were also lucky today
          # HOT DAMN
          0.00
          Wordle 236 2/6
          print('\nTips hat')
```

```
5757 possible words remain!
Best guesses according to Algorithm 1:
[('arose', 100.0), ('raise', 99.967), ('arise', 99.967), ('aloes', 99.967), ('tear s', 99.966)]
Best guesses according to Algorithm 2:
[('sores', 100.0), ('sales', 99.985), ('soles', 99.981), ('sates', 99.961), ('sire s', 99.948)]
Best guesses overall:
[('tares', 99.947), ('cares', 99.922), ('tales', 99.921), ('dares', 99.919), ('pare s', 99.916)]

10 possible words remain!
Best guesses according to Algorithm 1:
[('lapse', 100.0), ('sable', 99.859), ('false', 99.768), ('manse', 99.757), ('caus e', 99.713)]
Best guesses according to Algorithm 2:
```

```
[('pause', 100.0), ('cause', 99.857), ('masse', 99.857), ('passe', 99.857), ('false', 99.714)]

Best guesses overall:
[('pause', 99.853), ('cause', 99.785), ('lapse', 99.785), ('false', 99.741), ('manse', 99.736)]

Tips hat

In []:
```