Extra credit – Tried to Improve Prediction Acuuracy

ATTEMPT -1 : Alternate similarity metrics

Implemented Jacard and Dice Similarity (Bad Accuracies)

Other possible similarity measures

$$\text{sim}_{\text{cosine}}(\vec{v},\vec{w}) = \frac{\vec{v}\cdot\vec{w}}{|\vec{v}||\vec{w}|} = \frac{\sum_{i=1}^{N} v_i \times w_i}{\sqrt{\sum_{i=1}^{N} v_i^2}\sqrt{\sum_{i=1}^{N} w_i^2}}$$

$$\text{sim}_{\text{Jaccard}}(\vec{v},\vec{w}) = \frac{\sum_{i=1}^{N} \min(v_i,w_i)}{\sum_{i=1}^{N} \max(v_i,w_i)}$$

$$\text{sim}_{\text{Dice}}(\vec{v},\vec{w}) = \frac{2\times\sum_{i=1}^{N} \min(v_i,w_i)}{\sum_{i=1}^{N} (v_i+w_i)}$$

$$\text{sim}_{\text{JS}}(\vec{v}||\vec{w}) = D(\vec{v}|\frac{\vec{v}+\vec{w}}{2}) + D(\vec{w}|\frac{\vec{v}+\vec{w}}{2})$$

```
def cossim_jacard(v1,v2):
    numerator = np.sum(np.minimum(v1,v2))
    denominator = np.sum(np.maximum(v1,v2))
    JacSim = numerator/denominator
    return JacSim

def cossim_dice(v1,v2):
    numerator = np.sum(np.minimum(v1,v2))
    denominator = np.sum(v1+v2)
    DiceSim = (2*numerator)/denominator
    return DiceSim
```

Poor performance using Jacard similarity – not helpful

Group	Best-1 Acc	Best-5 Acc	Best-10 Acc
capital	0.000	0.000	0.000
currency	0.000	0.000	0.000
city-in-state	0.000	0.000	0.000
family	0.032	0.038	0.115
adjective-to-adverb	0.000	0.000	0.000
comparative	0.000	0.000	0.005
superlative	0.000	0.000	0.000
nationality-adjective	0.081	0.209	0.233

Poor performance using Dice similarity – not helpful

Group	Best-1 Acc	Best-5 Acc	Best-10 Acc
capital	0.000	0.000	0.000
currency	0.000	0.000	0.000
city-in-state	0.000	0.000	0.000
family	0.000	0.000	0.000
adjective-to-adverb	0.000	0.000	0.000
comparative	0.000	0.000	0.000
superlative	0.000	0.000	0.000
nationality-adjective	0.000	0.000	0.000

ATTEMPT - 2 : Alternate pre trained vectors

Used pre trained word vectors from https://nlp.stanford.edu/projects/glove/

Download the vectors for words from Wikipedia 2014 and Gigaword5(822 MB)

System lacked the processing power(process kept terminating abruptly)

ATTEMPT – 3 : Stemming words

Tried to add the context of word stem for word vectors. Logic: words having the same stem are more likely to have the same semantic meaning. However, this technique does not generate the words that can be derived or inflected from the stemmed forms. Hence did not improve prediction accuracy as expected.