- (a) It's intuitive that only when w equals to o, y_w wont be 0.
- (b) First, we have $\hat{y} = softmax(U^T v_c)$. Then

$$\begin{split} \frac{\partial J}{\partial v_c} &= \frac{\partial J}{\partial U^T v_c} \frac{\partial U^T v_c}{\partial v_c} \\ &= U^T (\hat{y} - y) \end{split}$$

(c) The first steps are the same as the last problem. We have $\hat{y} = softmax(U^T v_c)$. Then

$$\frac{\partial J}{\partial u_w} = \frac{\partial J}{\partial U^T v_c} \frac{\partial U^T v_c}{\partial u_w}$$
$$= v_c (\hat{y} - y)^T$$

(d) The process is as below,

$$\frac{\partial \sigma(x)}{\partial x} = \frac{e^x(e^x + 1) - e^x e^x}{\partial (e^x + 1)^2}$$
$$= \sigma(x)(1 - \sigma(x))$$

(e) The partial derivative for v_c is

$$(\sigma(u_o^T v_c) - 1)u_o + \sum_{k=1}^K (\sigma(-u_k^T v_c) - 1)u_k$$

The partial derivative for u_o is

$$(\sigma(u_o^T v_c) - 1)v_c$$

and for u_k is

$$(\sigma(-u_k^T v_c) - 1)v_c$$

This derivative is easier to compute because we don't need to compute the partial derivative for all words in vocabulary.

(f) (i) $\sum_{j \in window} \frac{\partial J(v_c, w_j, U)}{\partial U}$ (ii) $\sum_{j \in window} \frac{\partial J(v_c, w_j, U)}{\partial v_c}$

(ii)
$$\sum_{j \in window} \frac{\partial J(v_c, w_j, U)}{\partial v_c}$$

(iii) 0

After training, final loss is 9.367644.

The results of word vectors are as shown in the belowing picture.

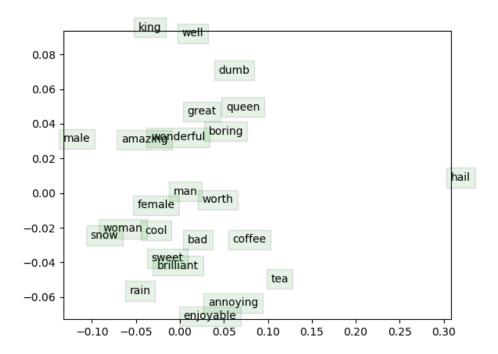


Figure 1: word vectors

We can see that

- 1. Some words like amazing, wonderful and boring are clustered well
- 2. Some words like hail, snow and rain are not.
- 3. It's interesting to see that the relative distance between male to king and female to queen is equal.