# PHDOpen Assigment2

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## 1 Single-head attention

#### 1.1

Attention weights can be interpreted as categorical probability distribution, because they are between 0 and 1, they sum up to 1 and higher weight means more important item.

#### 1.2

 $k_i \cdot q$  must be significantly greater then  $k_j \cdot q$ . In such a case  $\exp(k_i \cdot q) >> \exp(k_j \cdot q)$ , so  $\alpha_i >> \alpha_j$ 

#### 1.3

Let  $q = t(k_i + k_j)$ , where t >> 0. Then  $k_k \cdot q = 0$  for  $k \notin \{i, j\}$  and  $k_k \cdot q = t$  otherwise, so for large t  $\alpha_k \approx 0$  for  $k \notin \{i, k\}$  and  $\frac{1}{2}$  otherwise. Therefore:

$$c = \sum_{k=1}^{n} \alpha_k \cdot v_k = \sum_{k=1, k \notin \{i, j\}}^{n} 0 \cdot v_k + \alpha_i \cdot v_i + \alpha_j \cdot v_j = \frac{1}{2} \cdot v_i + \frac{1}{2} \cdot v_j = \frac{1}{2} \cdot (v_i + v_j)$$

#### 1.4

Let  $A = [s_1 s_2 \dots s_S]$ . According to this page  $P = A(A^T A)^{-1}A^T$  gives an orthogonal projection to the subspace formed by vectors  $(s_1, s_2, \dots, s_S)$ . Therefore  $c^T \cdot P = \frac{1}{2}v_i$  (because  $v_i$  belongs to subspace formed by vectors  $s_k$  and  $v_j$  belongs to subspace formed by vectors  $t_k$ , which is orthogonal to the first one), so, if  $M = 2 \cdot P = A(A^T A)^{-1}A^T$ , then  $c^T M = 2 \cdot c^T \cdot P = v_i$ .

#### 1.5

Let  $q = t(\mu_i + \mu_j)$ , where t >> 0.  $\epsilon << 1$ , so  $k_k \approx \mu_k$ . Therefore  $k_k \cdot q \approx 0$  for  $k \notin \{i, j\}$  and  $k_k \cdot q \approx t$  otherwise, so for large t  $\alpha_k \approx 0$  for  $k \notin \{i, k\}$  and  $\frac{1}{2}$  otherwise. Therefore:

$$c = \sum_{k=1}^{n} \alpha_k \cdot v_k = \sum_{k=1, k \notin \{i, j\}}^{n} 0 \cdot v_k + \alpha_i \cdot v_i + \alpha_j \cdot v_j = \frac{1}{2} \cdot v_i + \frac{1}{2} \cdot v_j = \frac{1}{2} \cdot (v_i + v_j)$$