

Appendix 1: Algorithms and FLOPS calculation

In this section, we announced our way to count the number of FLOPS. For brevity, we denoted M (N) as the number of output (input) capsules gathered from all locations over channels at the output (input) layer. In this way, a *votings* tensor has a size of $Q = M \times N \times D$, where D is the size of a capsule. We present routing algorithms and count their calculations as followed.

Algorithm 1: Attention-based routing (consider only 1 iteration)

Input: Votings v : size $M \times N \times D$ i^{th} global capsules: g^i : size $M \times D$ Output $(i + 1)^{th}$ global capsules: g^i : size $M \times D$ Activation probability for global capsules: <i>prob</i> size M		
$\forall m, n$	$a_{m,n} = \sum_k^D v_{m,n,k} \times g_{m,k}^i$	(1)
$\forall m, n$	$r_{m,n} = \frac{\exp(a_{m,n})}{\sum_k \exp(a_{k,n})}$	(2)
$\forall m$	$g_{m,:}^{i+1} = \sum_n^N r_{m,n} v_{m,n,:}$	(3)
$\forall m$	$g_{m,:}^{i+1} = \text{squash}(g_{m,:}^{i+1})$	(4)
$\forall m$	$prob_m = \ g_m\ _2$	(5)

The number of FLOPS can be counted by each line

(1): For each m, n , it needs $2 \times D$ calculations to compute the cosine similarity between a global capsule g_m^i and a voting capsule $v_{m,n}$, so there are $M \times N \times 2D$ calculations.

(2): First, the summation $\sum_k \exp(a_{k,n})$ requires $2M$ calculations if we assume that $\exp()$ is 1 calculation. Then, the softmax requires $M \times N$ calculations for the divisions. Therefore, there are $2M + M \times N$ calculations in total.

(3): To derive a new global capsule g_m^{i+1} , we sum up all N capsules weighted by the coefficients $r_{m,n}$, and the multiplication between N vectors and a scalar take $N \times D$ calculations. The summation among N capsules with D -dim takes $N \times D$ calculations, so we have $M \times N \times 2D$ calculations in total.

(4): $\text{squash}(g_m) = \frac{\|g_m\|_2^2}{1 + \|g_m\|_2^2} \frac{g_m}{\|g_m\|_2}$ working on $D - \text{dim}$ vector requires $3D$ calculations, $2D$ for computing the norm and D for scalar multiplication, so this step consumes $3 \times M \times D$ calculations

(5): the activation probability of M capsules requires $2 \times M \times D$ calculations for computing M norms.

As a result, one iteration in attention-based routing costs $4 \times M \times N \times D + M \times N + 5 \times M \times D + 2M$ calculations approximately. However, the most increasing term is $4 \times M \times N \times D = 4Q$.

Algorithm 2: Fuzzy-based routing (consider only 1 iteration)

Input: Votings v : size $M \times N \times D$ i^{th} global capsules: g^i : size $M \times D$ Output $(i + 1)^{th}$ global capsules: g^i : size $M \times D$ Activation probability for global capsules: <i>prob</i> size M		
$\forall m, n$	$d_{m,n} = \left(\ v_{m,n,:} - g_{m,:}^i\ _2 \right)^{\frac{2}{m_f - 1}}$	(1)
$\forall m, n$	$f_{m,n} = \frac{1}{\sum_k^M \frac{d_{m,n}}{d_{k,n}}}$	(2)
$\forall m, n$	$f_{m,n} = (f_{m,n})^{m_f}$	(3)
$\forall m, n$	$r_{m,n} = \frac{f_{m,n}}{\sum_k f_{m,k}}$	(4)
$\forall m$	$g_{m,:}^{i+1} = \sum_n^N r_{m,n} v_{m,n,:}$	(5)
$\forall m$	$\sigma_m^2 = \sum_n^N r_{m,n} \ v_{m,n,:} - g_{m,:}^{i+1}\ _2^2$	(6)
$\forall m$	$prob_m = \text{sigmoid}(\lambda(\beta_m - 0.5 * \ln(\sigma_m^2)))$	(7)

(1): For each m, n , it needs $2 \times D$ calculations to compute the Euclidean distance between a global capsule g_m^i and a voting capsule $v_{m,n}$, so there are $M \times N \times 2D$ calculations. Also, ones need another $M \times N \times D$ for the point-wise power, resulting in $3M \times N \times D$ calculations.

- (2): First, the summation $\sum_k^M \frac{1}{d_{k,n}}$ requires $2M$ calculations, so the term $\frac{1}{\sum_k \left(\frac{d_{m,n}}{d_{k,n}}\right)}$ consumes $2M + 2$ calculations. Totally, ones need $(2M + 2) \times N$ to derive $M \times N f_{m,n}$.
- (3): The point-wise power requires $M \times N$ calculations
- (4): First, the summation $\sum_k^N f_{m,k}$ requires N calculations, and the quotient $\frac{f_{m,n}}{\sum_k f_{m,k}}$ needs $M \times N$ calculations, so there are $N + M \times N$ calculations this step
- (5): As the algorithm above, it uses $M \times N \times 2D$ calculations to derive new global capsules
- (6): This step is the most consumption in the algorithm when it uses $M \times N \times 2D$ calculations to compute the distances and $2M \times N$ more calculations to perform the summation; thus there are $M \times N \times 2D + 2M \times N$ calculations.
- (7): there are 5 operations in total applied on m scalars, so it costs $5M$ calculations.
- As a result, one iteration in fuzzy-based routing costs $7M \times N \times D + 6M \times N + 2N + 5M$ calculations approximately. However, the most increasing term is $7 \times M \times N \times D = 7Q$.

Algorithm 3: EM-based routing (consider only 1 iteration)

Input: Votings v : size $M \times N \times D$ Activation probability at the previous iteration a : size M Routing coefficients at the previous iteration r : size $M \times N$ Output: Capsules at a higher level μ : size $M \times D$ Activation probability at a higher level a : size M	
M – step	
$\forall m, n \quad r_{m,n} = r_{m,n} \times a_n$	(1)
$\forall m, n \quad r_{m,n} = \frac{r_{m,n}}{\sum_k^N r_{m,k}}$	(2)
$\forall m \quad \mu_{m,:} = \sum_n^N r_{m,n} v_{m,n,:}$	(3)
$\forall m \quad \sigma_{m,d}^2 = \frac{1}{N} \sum_n^N r_{m,n} \times (v_{m,n,d} - \mu_{m,d})^2$	(4)
$\forall m \quad cost_{m,:} = (\beta_u + \log(\sigma_{m,:})) \sum_n^N r_{m,n}$	(5)
$\forall m \quad a_m = \text{sigmoid}(\lambda(\beta_a - \sum_d^D cost_{m,d}))$	(6)
E – step	
$\forall m, n \quad p_{m,n} = \frac{1}{\sqrt{\prod_d^D 2\pi\sigma_{m,d}^2}} \exp\left(-\frac{\sum_d^D (v_{m,n,d} - \mu_{m,d})^2}{2(\sigma_{m,d})^2}\right)$	(7)
$\forall n \quad r_{m,n} = \frac{a_m p_{m,n}}{\sum_k^M a_k p_{k,n}}$	(8)

- (1): This step obviously requires $M \times N$ calculations since $r_{m,n}$ and a_n are both scalars
- (2): The summation $\sum_k^N r_{m,k}$ takes N calculations, and the quotient $\frac{r_{m,n}}{\sum_k^N r_{m,k}}$ takes 1 calculation, resulting in $M \times (N + 1)$ calculations in total.
- (3): As the algorithms above, this step requires $M \times N \times 2D$ calculations
- (4): For each m, d , ones need $3N$ calculations to compute the summation, so there are $M \times D \times 3N$ calculations.
- (5): First, the summation $\sum_n^N r_{m,n}$ costs N calculations, then $3D$ calculations are needed to perform operations on the $D - \dim$ vector σ_m . Therefore, this step uses $M \times (N + 3D)$ operations.
- (6): Likewise, the $\sum_d^D cost_{m,d}$ requires D calculations, and ones need 3 more calculations to compute a_m . Since we need to compute m activation probabilities, this step requires $M \times (N + 3)$ calculations.
- (7): Firstly, the term $\frac{\sum_d^D (v_{m,n,d} - \mu_{m,d})^2}{2(\sigma_{m,d})^2}$ uses $3D + 1$ calculations. Secondly, the terms $\frac{1}{\sqrt{\prod_d^D 2\pi\sigma_{m,d}^2}}$ requires $2D + 2$ calculations. Totally, this step takes $M \times N \times (5D + 3)$ calculations.
- (8): The summation $\sum_k^M a_k p_{k,n}$ uses $2M$ calculations and the quotient $\frac{a_m p_{m,n}}{\sum_k^M a_k p_{k,n}}$ takes 2 more calculations, resulting in $N \times (2M + 2)$.
- As a result, one iteration in EM-based routing costs $10M \times N \times D + 8M \times N + 3M \times D + 3M + 2N$ calculations approximately. However, the most increasing term is $10 \times M \times N \times D = 10Q$.