



Softmax OP

For an input Vector $x \in \mathbb{R}^N$ and grab only first β .

1. $m(x) := \max_i x_i$; end the largest score in this Block β
(we need this to subtract to in order to prevent overflow)

2. exponentiated Differences vector: $t(x) := [e^{x_1 - m(x)} \dots e^{x_\beta - m(x)}]$
 $x_i - m(x) \leq 0$, so safely in the range of $[0, 1]$

3. sum of exponents (Denominator) $\ell(x) := \sum_i t(x)_i$

4. $\text{softmax}(x) := \frac{t(x)}{\ell(x)}$ Local softmax (invariant for softmax for now)

5. For vectors $x^1, x^2 \in \mathbb{R}^\beta$, decompose the softmax of the concat $x = [x^1, x^2] \in \mathbb{R}^{2\beta}$ as

$m(x) = m([x^1, x^2]) = \max(m(x^1), m(x^2))$; find the max between B1's local max and B2's local max

$$t(x) = [e^{m(x^1) - m(x)} t(x^1) \quad e^{m(x^2) - m(x)} t(x^2)]$$

$$e^{m(x^1) - m(x)} \cdot e^{x_i^1 - m(x^1)}$$

$\Downarrow$

$$e^{x_i^1 - m(x)}$$

$$6. \ell(x) = \ell([x^1, x^2]) = e^{m(x^1) - m(x)} \ell(x^1) + e^{m(x^2) - m(x)} \ell(x^2), \quad \text{softmax}(x) = \frac{t(x)}{\ell(x)}$$

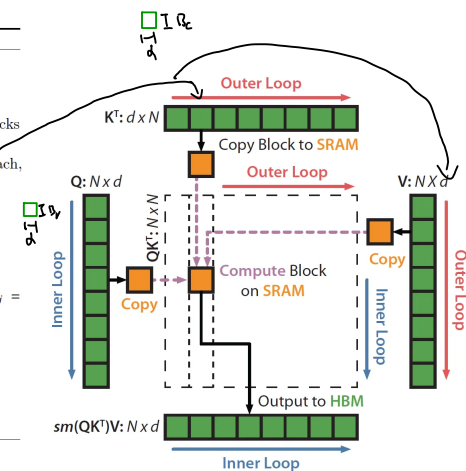
logic continues to $\frac{N}{\beta}$ of blocks with dim β

Alg1 assumes Batchsize of 1.

Algorithm 1 FLASHATTENTION

Require: Matrices $Q, K, V \in \mathbb{R}^{N \times d}$ in HBM, on-chip SRAM of size M .

- 1: Set block sizes $B_c = \lceil \frac{M}{d} \rceil$, $B_r = \min(\lceil \frac{M}{d} \rceil, d)$.
- 2: Initialize $O = (0)_{N \times d} \in \mathbb{R}^{N \times d}$, $\ell = (0)_{N \times 1} \in \mathbb{R}^N$, $m = (-\infty)_{N \times 1} \in \mathbb{R}^N$ in HBM.
- 3: Divide Q into $T_r = \lceil \frac{N}{B_r} \rceil$ blocks $Q_1, \dots, Q_{T_r}$ of size $B_r \times d$ each, and divide K, V into $T_c = \lceil \frac{N}{B_c} \rceil$ blocks $K_1, \dots, K_{T_c}$ and $V_1, \dots, V_{T_c}$, of size $B_c \times d$ each.
- 4: Divide O into T_r blocks $O_1, \dots, O_{T_r}$ of size $B_r \times d$ each, divide ℓ into T_r blocks $\ell_1, \dots, \ell_{T_r}$ of size B_r each, divide m into T_r blocks $m_1, \dots, m_{T_r}$ of size B_r each.
- 5: for $1 \leq j \leq T_c$ do
- 6: Load K_j, V_j from HBM to on-chip SRAM.
- 7: for $1 \leq i \leq T_r$ do
- 8: Load Q_i, O_i, ℓ_i, m_i from HBM to on-chip SRAM.
- 9: On chip, compute $S_{ij} = Q_i K_j^T \in \mathbb{R}^{B_r \times B_c}$.
- 10: On chip, compute $\tilde{m}_{ij} = \text{rowmax}(S_{ij}) \in \mathbb{R}^{B_r}$, $\tilde{P}_{ij} = \exp(S_{ij} - \tilde{m}_{ij}) \in \mathbb{R}^{B_r \times B_c}$ (pointwise), $\tilde{\ell}_{ij} = \text{rowsum}(\tilde{P}_{ij}) \in \mathbb{R}^{B_r}$.
- 11: On chip, compute $m_i^{\text{new}} = \max(m_i, \tilde{m}_{ij}) \in \mathbb{R}^{B_r}$, $\ell_i^{\text{new}} = e^{m_i - m_i^{\text{new}}} \ell_i + e^{\tilde{m}_{ij} - m_i^{\text{new}}} \tilde{\ell}_{ij} \in \mathbb{R}^{B_r}$.
- 12: Write $O_i \leftarrow \text{diag}(\ell_i^{\text{new}})^{-1} (\text{diag}(\ell_i) e^{m_i - m_i^{\text{new}}} O_i + e^{\tilde{m}_{ij} - m_i^{\text{new}}} \tilde{P}_{ij} V_j)$ to HBM.
- 13: Write $\ell_i \leftarrow \ell_i^{\text{new}}$, $m_i \leftarrow m_i^{\text{new}}$ to HBM.
- 14: end for
- 15: end for
- 16: Return O .



1. $\text{ceil}(\frac{N}{\beta})$, because we need to allocate $Q, K, V, O \in \mathbb{R}^{d \times d}$
2. In HBM: $O \in \mathbb{R}^{N \times d}$ (output matrix)
 $\ell \in \mathbb{R}^N$ (normalizer, initialized to all 0)
 $m \in \mathbb{R}^N$ (row max vector, initialized to $-\infty$)

3. Divide Q into $T_r = \lceil \frac{N}{B_r} \rceil$ blocks $Q_1, \dots, Q_{T_r}$
 $Q_i \in \mathbb{R}^{B_r \times d}$
Divide K, V into $T_c = \lceil \frac{N}{B_c} \rceil$ each of $K_i, V_i \in \mathbb{R}^{B_c \times d}$

4. Divide O into T_r blocks $O_1, \dots, O_{T_r}$ of size $O_i \in \mathbb{R}^{B_r \times d}$
Divide ℓ into T_r blocks $\ell_1, \dots, \ell_{T_r}$ of size B_r , also m as well

Side: we can load ℓ and m into SRAM cuz they are so small

B_r (row Block Size)
2nd two buckets, we are still allocating space for every Q, K, V, O
 B_c (col Block size), but if B_c is bigger than d , K and V take advantage of B_c ($B_c \cdot d + B_c \cdot d$) whereas Q and O fall back to d to save resources
 $\hookrightarrow$ for QK^T

Total SRAM: $2 \cdot (\frac{B_r}{\beta} \cdot d) + 2(B_c \cdot d) + (B_r \cdot B_c)$

if $B_c < d$, then B_c which means: $4B_c \cdot d + B_c \cdot B_c$

9. $S_{ij} = Q_i K_j^T \in \mathbb{R}^{B_r \times B_c}$
we don't need the $N \times N$ since they are made up of tiny grid, so we only need to maintain one grid at a time

11. $m_i^{new} = \max_{j=1} (m_i; \tilde{m}_{ij}) \in \mathbb{R}^{B \times B}$

$\{ \cdot \} = e$ $\tilde{l}_i + e^{\tilde{m}_{ij} - m_i^{new}} \tilde{l}_j \in \mathbb{R}^{B \times B}$

$B_r \times d$
max for each row in the small grid but also the bigger

rescaled sum from past green blocks $i=2 \rightarrow$

rescaled sum from new yellow $j=3$

12. $O_i \leftarrow \text{diag}(\tilde{l}_i)^{new} (\text{diag}(\tilde{l}_i)^{m_i - m_i^{new}} O_i + e^{\tilde{m}_{ij} - m_i^{new}} \tilde{P}_{ij} V_j)$

green yellow

Green:

$\text{diag}(\tilde{l}_i)^{m_i - m_i^{new}} O_i$
 $O_i \in \mathbb{R}^{B \times d}$: current val of O_i for previous $j=1, 2$
 $\text{diag}(\tilde{l}_i)$: $B_r \times B_r$ diagonal matrix
 $\text{diag}(\tilde{l}_i) O_i$ undivides $O_i \rightarrow$ goes back to Σ

dim d
 $O_i[k] = [O_{i,1}, \dots, O_{i,d}]$
 $O_i[k] = \frac{1}{\tilde{l}_i[k]} [\text{row } O_i, \dots, \text{row } O_d]$
 $\text{diag}(\tilde{l}_i)[k] \cdot O_i[k] = [\text{row } O_i, \dots, \text{row } O_d]$
 $\text{rdw}_c = \sum_{\text{past keys } p} e^{S_{kp} - m_i} \cdot V_{pc}$

Yellow:

$e^{\tilde{m}_{ij} - m_i^{new}} \tilde{P}_{ij} V_j$

$\left[e^{m_i - m_i^{new}} \sum_p e^{S_{kp} - m_i} \cdot V_{p1}, \dots, e^{m_i - m_i^{new}} \sum_p e^{S_{kp} - m_i} \cdot V_{pd} \right]$

$\tilde{P}_{ij} \in \mathbb{R}^{B_r \times B_c}$
 $\downarrow$ score $\alpha: k_j^T$
 $\tilde{P}_{ij}[k] = [e^{S_{k,1} - \tilde{m}_{ij}}, e^{S_{k,2} - \tilde{m}_{ij}}, \dots, e^{S_{k,B_c} - \tilde{m}_{ij}}]$
 $V_j \in \mathbb{R}^{B_c \times d}$
 $\tilde{P}_{ij} V_j[k] = \left[\sum_{q=1}^{B_c} e^{S_{kq} - \tilde{m}_{ij}} V_{q1}, \sum_{q=1}^{B_c} e^{S_{kq} - \tilde{m}_{ij}} V_{q2}, \dots \right]$
 $e^{\tilde{m}_{ij} - m_i^{new}} \left(\sum_{q=1}^{B_c} e^{S_{kq} - \tilde{m}_{ij}} V_{q1} \right) = \sum_{q=1}^{B_c} e^{S_{kq} - m_i^{new}} V_{q1}$

Green + Yellow

Part A[k] = $\left[\sum_{\text{past } p} e^{S_{kp} - m_i^{new}} V_{p1}, \dots, \sum_{\text{past } p} e^{S_{kp} - m_i^{new}} V_{pd} \right]$
 Part B[k] = $\left[\sum_q e^{S_{kq} - m_i^{new}} V_{q1}, \dots, \sum_q e^{S_{kq} - m_i^{new}} V_{qd} \right]$
 $A+B = \sum_{\text{past } q} e^{S_{k,q11} - m_i^{new}} V_{q11} \in \mathbb{R}^{B_r \times d}$

Final $\left[\frac{1}{t_1}, \dots, \frac{1}{t_{\text{row}}} \right] \times (\text{part A} + \text{part B}) = O_i^{new}$
 $B_r \times B_r \quad B_r \times d \quad B_r \times d$

Complexity Analysis

$Q, K, V, O \in \mathbb{R}^{N \times d}$, ℓ and $m(N)$ in HBM

Total HBM mem $\approx 4Nd + 2N$

Space complexity $= O(N)$ note d is much smaller than N

IO (HBM Access) complexity: $O(N^2 d^2 M^{-1})$

1. Num of Blocks $T = \frac{N}{B} = \frac{N}{M/d} = \frac{NM}{d}$

2. Total iter: $T_r \times T_c \approx \frac{16N^2 d^2}{M^2}$

3. Data transferred per iter: $B \cdot 4d = M$

4. Total HBM Access = Total iter $\times$ data per iter $= \frac{16N^2 d^2}{M}$