

A Algorithm for Multi-step Selective-Scan-Update kernel

Algorithm 1 Fused multi-step State-Update kernel. This Algorithm is simplified with the only input being x, A, B, C, Δ, h in reference to the implementation of [10].

Notation $Z^* \leftarrow Z$: load Z into shared memory for computation. $Z \leftarrow Z^*$: Storing Z out to High-Bandwidth-Memory (HBM). Any variable with * means that this variable is in shared memory, everything else is in HBM.

Shapes B - Batch, L - Sequence Length, H - Num heads, P - State Dimension , D - Head Dimension, G - Num groups

Input: $h_0 : (B, H, D, P)$ - initial state, $x : (B, L, H, D)$ $A : (H, D, P)$, $B : (B, L, G, P)$, $C : (B, L, G, P)$, $\Delta : (B, L, H, D)$

Output: $y : (B, L, H, D)$ - Output, $\hat{h} : (B, H, D, P)$ - updated state

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1:  $h^* \leftarrow h_0$ 
2:  $i \leftarrow 0$ 
3: while  $i < L$  do
4:    $x^* \leftarrow x[:, i, :, :]$ 
5:    $\Delta^* \leftarrow \Delta[:, i, :, :]$ 
6:    $B^* \leftarrow B[:, i, :, :]$ 
7:    $C^* \leftarrow C[:, i, :, :]$ 
8:    $A^* \leftarrow A$ 
9:    $\bar{A}^* \leftarrow \exp(\Delta^* \times A^*)$ 
10:   $\bar{B}^* \leftarrow \Delta^* \times B^*$ 
11:   $h^* \leftarrow h^* \times \bar{A}^* + \bar{B}^* \times x^*$             $\triangleright$  State is not stored out to HBM at this point
12:   $y^* \leftarrow C^* \times h^*$ 
13:   $y[:, i, :, :] \leftarrow y^*$ 
14: end while
15:  $\hat{h} \leftarrow h^*$ 

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B Algorithm for Activation Replay

Line 1, 3, 5, 8 are performed by the original Mamba-2 block to get the output and update the states. The added operations are the extra *Conv1d_update* and *SSM_update* at line 2 and 5 to attain the updated states using the correct cached activations. Line 4 and 7 saves the activation of the current iteration for the use of next iteration.

Algorithm 2 Abstracted Mamba-2 block with Activation Replay. The two functions are simplified abstraction of what is actually used in the implementation.

Notation $A[a:b]$: a -th position to b -th position of A in the sequence length dimension

function $\text{Conv1d_update}(x, B, C, M)$:

- return** (x', B', C', M') ▷ Performs 1d convolution on x , B , C , and update conv state M

function $\text{SSM_update}(x, B, C, dt, S)$:

- return** (y, S') ▷ Performs SSM update with x , B , C , dt to get output y and update SSM state S

Input: E : token embedding, S : SSM state, M : Conv state, c : number of correct tokens

- $(x_{\text{cache}}^{\text{pre_conv}}, B_{\text{cache}}^{\text{pre_conv}}, C_{\text{cache}}^{\text{pre_conv}})$: Cached activation for Conv states from last forward pass
- $(x_{\text{cache}}^{\text{post_conv}}, B_{\text{cache}}^{\text{post_conv}}, dt_{\text{cache}})$: Cached activation for SSM states from last forward pass

Output: \hat{E} : New token embedding, \hat{S} : SSM state after the i -th correct token, \hat{M} : Conv state after the i -th correct token

- 1: $x^{\text{pre_conv}}, B^{\text{pre_conv}}, C^{\text{pre_conv}}, dt \leftarrow \text{in_projection}(E)$
- 2: $(_, _, _, \hat{M}) \leftarrow \text{Conv1d_update}(x_{\text{cache}}^{\text{pre_conv}}[0:c], B_{\text{cache}}^{\text{pre_conv}}[0:c], C_{\text{cache}}^{\text{pre_conv}}[0:c], M)$ ▷ Getting updated Conv state
- 3: $(x^{\text{post_conv}}, B^{\text{post_conv}}, C^{\text{post_conv}}, _) \leftarrow \text{Conv1d_update}(x^{\text{pre_conv}}, B^{\text{pre_conv}}, C^{\text{pre_conv}}, \hat{M})$ ▷ Actual Conv1d update
- 4: $(x_{\text{cache}}^{\text{pre_conv}}, B_{\text{cache}}^{\text{pre_conv}}, C_{\text{cache}}^{\text{pre_conv}}) \leftarrow (x^{\text{pre_conv}}, B^{\text{pre_conv}}, C^{\text{pre_conv}})$ ▷ Saving activations for next forward pass
- 5: $(_, \hat{S}) \leftarrow \text{SSM_update}(x_{\text{cache}}^{\text{post_conv}}[0:c], B_{\text{cache}}^{\text{post_conv}}[0:c], dt_{\text{cache}}[0:c], S)$ ▷ Getting the updated state using the correctly predicted tokens
- 6: $(y, _) \leftarrow \text{SSM_update}(x^{\text{post_conv}}, B^{\text{post_conv}}, C^{\text{post_conv}}, dt, \hat{S})$ ▷ Actual SSM update
- 7: $(x_{\text{cache}}^{\text{post_conv}}, B_{\text{cache}}^{\text{post_conv}}, dt_{\text{cache}}) \leftarrow (x^{\text{post_conv}}, B^{\text{post_conv}}, dt)$ ▷ Saving activations for next forward pass
- 8: $\hat{E} \leftarrow \text{out_projection}(y)$
