

---

**Algorithm 1:** *ComputeAlpha*

---

**Data:**  $out_{m \times n}$  (result of softmax), where  $m = \bar{W}_{\text{unpadded}}/4$ ,  $n = |\hat{A}|$ ,  
 $label$  (encoded by alphabet),  
 $T, L, bl = 0$  (blank index)  
**begin**  
     $a = \text{zeros}(T, L)$   
     $a[0][0] = out_{\text{unpadded}}[0][bl]$   
     $a[0][1] = out_{\text{unpadded}}[0][label[0]]$   
     $c = a[0][0] + a[0][1]$   
    **if**  $c > 0$  **then**  
         $a[0][0] = \frac{a[0][0]}{c}$   
         $a[0][1] = \frac{a[0][1]}{c}$   
    **for**  $t := 1$  **to**  $T$  **do**  
         $start = \max(0, L - 2 \times (T - t))$   
         $end = \min(2 \times t + 2, L)$   
        **for**  $s := start$  **to**  $L$  **do**  
             $i = \text{floor}(\frac{s-1}{2})$   
             $a[t][s] = a[t-1][s]$   
            **if**  $s > 0$  **then**  
                 $a[t][s] = a[t][s] + a[t-1][s-1]$   
            **if**  $s \bmod 2 = 0$  **then**  
                 $a[t][s] = a[t][s] \times out_{\text{unpadded}}[t][bl]$   
            **else if**  $s = 1$  **or**  $label[i] = label[i-1]$  **then**  
                 $a[t][s] = a[t][s] \times out_{\text{unpadded}}[t][label[i]]$   
            **else**  
                 $a[t][s] = (a[t][s] + a[t-1][s-2]) \times out_{\text{unpadded}}[t][label[i]]$   
         $c = \sum_{i=start}^{end} a[t][i]$   
        **if**  $c > 0$  **then**  
            **for**  $s := start$  **to**  $end$  **do**  
                 $a[t][s] = \frac{a[t][s]}{c}$   
    **return**  $a$

---

---

**Algorithm 2:** CTC Loss and **softmax** gradient computation

---

**Data:**  $out_{M \times N}$  (result of softmax), where  $M = \bar{W}_{\text{padded}}/4$ ,  $N = |\hat{A}|$ ,  
 $label$  (encoded by alphabet),  
 $w = \bar{W}_{\text{unpadded}}/4$ ,  
 $bl = 0$  (blank index)  
**begin**  
     $L = 2 \times \text{len}(label) + 1$   
     $T = w$   
     $out_{\text{unpadded}} = \text{zeros}(T, N)$   
    **for**  $t := 0$  **to**  $T$  **do**  
        **for**  $n := 0$  **to**  $N$  **do**  
             $out_{\text{unpadded}}[t][n] = out_{\text{padded}}[t][n]$   
     $a = \text{ComputeAlpha}(out_{\text{unpadded}}, label, T, L, bl)$   
     $out_{\text{unpadded}}^{\text{flipped}} = \text{flipud}(\text{fliplr}(out_{\text{unpadded}}))$   
     $label_{\text{reversed}} = \text{reverse}(label)$   
     $b = \text{ComputeAlpha}(out_{\text{unpadded}}^{\text{flipped}}, label_{\text{reversed}}, T, L, bl)$   
     $b = \text{flipud}(\text{fliplr}(b))$   
     $ab = a * b$   
     $lab = \text{zeros}(T, N)$   
    **for**  $s := 0$  **to**  $L$  **do**  
        **if**  $s \bmod 2 = 0$  **then**  
            **for**  $t := 0$  **to**  $T$  **do**  
                 $ab[t][s] = \frac{ab[t][s]}{out_{\text{unpadded}}[t][bl]}$   
                 $lab[t][bl] = lab[t][bl] + ab[t][s]$   
        **else**  
            **for**  $t := 0$  **to**  $T$  **do**  
                 $i = \text{floor}(\frac{s-1}{2})$   
                 $ab[t][s] = \frac{ab[t][s]}{out_{\text{unpadded}}[t][label[i]]}$   
                 $lab[t][i] = lab[t][i] + ab[t][s]$   
     $lh = \text{zeros}(T)$   
    **for**  $t := 0$  **to**  $T$  **do**  
         $lh[t] = \sum_{s=1}^L ab[t][s]$   
     $loss = -\sum_{t=1}^T \ln lh[t]$   
     $softmaxGrad = \text{zeros}(M, N)$   
    **for**  $t := 0$  **to**  $T$  **do**  
        **for**  $n := 0$  **to**  $N$  **do**  
             $softmaxGrad[t][n] = out_{\text{unpadded}}[t][n] - \frac{lab[t][n]}{out_{\text{unpadded}}[t][n] * lh[t]}$   
    **return**  $loss, softmaxGrad$ 

---