Implemented Equations

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Binary CRBM 1

$V \in \mathbb{R}^{m imes k}$	(1)
$W \in \mathbb{R}^{m \times n}$	(2)
$b \in \mathbb{R}^n$	(3)
$c \in \mathbb{R}^m$	(4)
$y \in \mathbb{R}^k$	(5)
$x \in \mathbb{R}^n$	(6)
$z \in \mathbb{R}^m$	(7)

Binary up:

$$z = c + V \cdot y + W \cdot x$$

$$z = \text{sigm}(z)$$

$$z = \text{set randomly to true false, based on } uniform() > z$$
(10)

 $z = \text{set randomly to true false, based on } uniform() > z_i$ (10)

Algorithm 1 Binary up

Input: $x \in \mathbb{R}^n$ Input: $y \in \mathbb{R}^k$ **Output:** $z \in \mathbb{R}^m$

1: $z \leftarrow c + V \cdot y + W \cdot x$

2: $z \leftarrow \operatorname{sigm}(z)$

3: $z \leftarrow \text{set randomly to true false, based on } uniform() > z_i$

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Algorithm 2 Binary down

```
Input: z \in \mathbb{R}^m
Output: x \in \mathbb{R}^n

1: x \leftarrow b + W^T \cdot z

2: x \leftarrow \text{sigm}(x)

3: x \leftarrow \text{set randomly to true false, based on } uniform() > x_i
```

Algorithm 3 Control update

```
Input: x \in \mathbb{R}^n
Input: y \in \mathbb{R}^k
Input: u \in \mathbb{N}
Output: x \in \mathbb{R}^n
Output: z \in \mathbb{R}^m
1: z \leftarrow \text{binary\_up}(y, x)
2: for i from 1 to u do
3: x \leftarrow \text{binary\_down}(z)
4: z \leftarrow \text{binary\_up}(y, x)
5: end for
```

Algorithm 4 Learn update

```
Input: x \in \mathbb{R}^n

Input: y \in \mathbb{R}^k

Input: u \in \mathbb{N}

Output: x \in \mathbb{R}^n

Output: z \in \mathbb{R}^m

1: z \leftarrow \text{binary\_up}(y, x)

2: for i from 1 to u - 1 do

3: x \leftarrow \text{binary\_down}(z)

4: z \leftarrow \text{binary\_up}(y, x)

5: end for

6: x \leftarrow \text{binary\_down}(z)
```

Algorithm 5 Initialise b

```
Input: n \in \mathbb{N}
                                                                                                         ▶ Nr. of input units
                                                                                                  ▷ Nr. of rows in the data
Input: t \in \mathbb{N}
 1: for i from 1 to n do
 2:
         s \leftarrow \operatorname{columnsum}(\hat{S}(i))
                                                                                                  ▷ Only 1's in the column
         if s = t then
 3:
              p \leftarrow t-1/t
 4:
 5:
         else if s == 0 then
                                                                                                  ▷ Only 0's in the column
              p \leftarrow 1/t
 6:
 7:
         else
 8:
              p \leftarrow s/t
         end if
 9:
10: end for
```

```
Algorithm 6 Training
Input: CRBM
Input: S \in \mathbb{R}^{t \times |S|}
Input: A \in \mathbb{R}^{t \times |A|}
Input: bins \in \mathbb{N}
Input: batchsize \in \mathbb{N}
Input: numepochs \in \mathbb{N}
Output: x \in \mathbb{R}^n
Output: z \in \mathbb{R}^m
Output: m \in \mathbb{R}
                                                                                                                                             ▶ Momentum parameter
Output: \gamma \in \mathbb{R}
                                                                                                                                                                ▶ Weight cost
  1: N \leftarrow \lceil \log_2(\text{bins}) \rceil
  2: \hat{S} \leftarrow \text{binarise}(S)
  3: \hat{A} \leftarrow \text{binarise}(A)
  4: rbm.W \leftarrow \mathcal{N}(0, 0.01)
  5: rbm.V \leftarrow \mathcal{N}(0, 0.01)
  6: rbm.c \leftarrow 0
  7: rbm.b \leftarrow initialise based on data (see Alg. 5
  8: for i in 1 to numepochs do
  9:
             r \leftarrow \text{rand}(t - \text{batchsize})
             indices \leftarrow [r, r + \text{batchsize}]
10:
              \tilde{S} \leftarrow \hat{S}(\mathsf{rows} = \mathsf{indices})
11:
             \tilde{A} \leftarrow \hat{A}(\text{rows} = \text{indices})
12:
             z \leftarrow \text{binary\_up}(\tilde{S}, \tilde{A})
13:
             \bar{A} \leftarrow \text{random action matrix} \in [0, 1]^{|A|}
14:
             (\bar{A}, Z) \leftarrow \text{learn}(s, \bar{A})
15:
             \mathbb{E}\{b\} \leftarrow \operatorname{avg}(\tilde{A}) - \operatorname{avg}(\bar{A})
16:
             \mathbb{E}\{c\} \leftarrow \operatorname{avg}(\tilde{z}) - \operatorname{avg}(\bar{Z})
17:
             \mathbb{E}\{W\} \leftarrow z^T a - Z^T \bar{A}/\text{batchsize}
18:
19:
             \mathbb{E}\{V\} \leftarrow z^T s - Z^T \bar{s}/\text{batchsize}
             b \leftarrow b + \alpha \mathbb{E}\{b\}
20:
             c \leftarrow c + \alpha \mathbb{E}\{c\}
21:
              W \leftarrow W + \alpha \mathbb{E}\{W\}
22:
23:
              V \leftarrow V + \alpha \mathbb{E}\{V\}
             if m > 0 then
24:
                    b \leftarrow b + \alpha \cdot m \cdot b^m
25:
                    c \leftarrow c + \alpha \cdot m \cdot c^m
26:
                    W \leftarrow W + \alpha \cdot m \cdot W^m
27:
                     V \leftarrow V + \alpha \cdot m \cdot V^m
28:
29:
             end if
             if \gamma > 0 then
30:
31:
                    W \leftarrow (1 - \gamma)W
                    V \leftarrow (1 - \gamma)V
32:
             end if
33:
             b^m \leftarrow \mathbb{E}(b)
34:
             c^m \leftarrow \mathbb{E}(c)
35:
36:
              W^m \leftarrow \mathbb{E}(W)
              V^m \leftarrow \mathbb{E}(V)
37:
38: end for
```