

# Restricted Boltzmann Machine

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Here is the Python code for the Restricted Boltzmann Machine:

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4
5 def sample_patterns_equal_prob(batch_size, patterns):
6     pattern_indices = np.random.choice(len(
7         patterns), size=batch_size, p=[0.25, 0.25,
8         0.25, 0.25])
9     mini_batch = np.array([patterns[idx] for idx
10         in pattern_indices])
11     return mini_batch
12
13
14 def p_boltzmann(b):
15     denominator = 1+np.exp(-2*b)
16     return 1/denominator
17
18
19 def compute_delta_w(eta, b_i_h_0, v_0, b_i_h, v, w):
20     delta_w = w.copy()
21     for m in range(0, len(w)):
22         for n in range(0, len(w[0])):
23             term1 = np.tanh(b_i_h_0[m])*v_0[n]
24             term2 = np.tanh(b_i_h[m])*v[n]
25             delta_w[m][n] = eta*(term1 - term2)
26     return delta_w
27
28
29 def compute_delta_theta_v(eta, v_0, v, theta_v):
30     delta_theta_v = np.zeros(len(theta_v))
31     for n in range(0, len(theta_v)):
32         delta_theta_v[n] = -eta*(v_0[n]-v[n])
33     return delta_theta_v
```

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31
32
33 def compute_delta_theta_h(eta, b_i_h_0, b_i_h, theta_h
    ):
34     delta_theta_h = np.zeros(len(theta_h))
35     for m in range(0, len(theta_h)):
36         term1 = np.tanh(b_i_h_0[m])
37         term2 = np.tanh(b_i_h[m])
38         delta_theta_h[m] = -eta*(term1-term2)
39     return delta_theta_h
40
41
42
43 def train_RBM(M, eta, k, epochs, batch_size): #, patterns)
44
45     PD = np.array([0.25, 0, 0, 0.25, 0, 0.25, 0.25,
46                    0])
47
48     ### Init ###
49     #patterns that have a 1/4 probability to be
50     #sampled, used for training
51     patterns = np.array([[ -1, -1, -1],
52                          [ -1, 1, 1],
53                          [ 1, -1, 1],
54                          [ 1, 1, -1]])
55
56     variance = 1/ np.maximum(3, M)
57     std = np.sqrt(variance)
58     w = np.random.normal(0, std, size=(M, 3))
59
60     theta_v = np.zeros(3)
61     theta_h = np.zeros(M)
62     v = np.zeros(3)
63     h = np.zeros(M)
64     #####
65
66     energies = []
67     all_samples = []
68
69     for epoch in range(0, epochs):
70
71         ### Init weights and thresholds ###
72         delta_w = np.zeros((M, 3))
73         delta_theta_v = np.zeros(3)

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74     delta_theta_h = np.zeros(M)
75     #####
76
77     ### Sample from patterns with equal prob ###
78     mini_batch = sample_patterns_equal_prob(
79         batch_size, patterns)
80     #####
81
82     for sample in mini_batch:
83         all_samples.append(sample)
84
85     for mu, pattern in enumerate(mini_batch):
86         v = pattern.copy()
87         v_0 = v.copy()
88         b_i_h_0 = np.zeros(M)
89
90         ### Calculate  $b_i^h(0)$  and update all
91         hidden neurons  $h_i(0)$  ###
92         for i in range(0, len(b_i_h_0)):
93             b_i_h_0[i] = np.dot(w[i], v_0) - theta_h[
94                 i]
95             r = np.random.rand()
96             p_B = p_boltzmann(b_i_h_0[i])
97             if(r < p_B):
98                 h[i] = 1
99             else:
100                 h[i] = -1
101             #
102             #####
103
104         b_i_h = b_i_h_0.copy()
105         b_j_v = np.zeros(3)
106         for step in range(0, k):
107             ### update visible neurons ###
108             for j in range(0, 3):
109                 b_j_v[j] = np.dot(h, w[:, j]) -
110                     theta_v[j]
111                 p_B_v = p_boltzmann(b_j_v[j])
112                 r = np.random.rand()
113                 if(r < p_B_v):
114                     v[j] = 1
115                 else:
116                     v[j] = -1
117             #####
118
119             ### update hidden neurons ###

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```

114         for i in range(0,M):
115             b_i_h[i] = np.dot(w[i],v)-theta_h[
116                 i]
117             p_B_h = p_boltzmann(b_i_h[i])
118             r = np.random.rand()
119             if(r < p_B_h):
120                 h[i] = 1
121             else:
122                 h[i] = -1
123             #####
124             ### calculate delta_w ###
125             for m in range(0,M):
126                 for n in range(0,3):
127                     delta_w[m][n] += eta* ( np.tanh(
128                         b_i_h_0[m])*v_0[n] - np.tanh(
129                             b_i_h[m])*v[n] )
130                     #####
131                     ### calculate delta_theta_v ###
132                     for n in range(0,3):
133                         delta_theta_v[n] -= eta*(v_0[n]-v[n])
134                     #####
135                     ## calculate delta_theta_h ###
136                     for m in range(0, M):
137                         delta_theta_h[m] -= eta*(np.tanh(
138                             b_i_h_0[m])-np.tanh(b_i_h[m]))
139                     #####
140                     ### update values ###
141                     w += delta_w
142                     theta_h += delta_theta_h
143                     theta_v += delta_theta_v
144                     #####
145                     ### Monitor energy function ###
146                     H = compute_energy_function(w, h, v , theta_v,
147                         theta_h)
148                     energies.append(H)
149
150     tmp_freq = compute_frequencies(all_samples)
151     print("training distribution: ", tmp_freq)
152
153     ### Monitor energy function ###
154     epoch_range = np.arange(epochs)
155     plt.plot(epoch_range,energies,marker='o', color='

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        'green', markersize=1, linestyle='--')
155 plt.title(f'Energy of training, M={M}')
156 plt.xlabel('Epochs')
157 plt.ylabel('Energy')
158 plt.grid()
159 plt.tight_layout()
160 plt.show()
161
162 return w, theta_h, theta_v
163
164 def D_kl_bound(M):
165     #number of inputs
166     N=3
167     expression = 3 - int(np.log2(M+1)) - ((M+1)/(2**
        int(np.log2(M+1))))
168     if M < (2**(N-1) - 1):
169         return np.log(2)*expression
170
171     elif M >= (2**(N-1) - 1):
172         return np.log(2)*0
173
174
175 def D_kl(PD,PB):
176     sum= 0
177     for mu in range (0,len(PD)):
178         if(PB[mu] > 0 and PD[mu] > 0):
179             #print("PB^mu: ", PB[mu], ", ", "PD^mu: ",
        PD[mu])
180             sum+= PD[mu]* np.log(PD[mu]/PB[mu])
181     return sum
182
183 #convert binary number to decimal index, then
    increment.
184 def compute_frequencies(samples):
185     n_patterns = 2**len(samples[0])
186     PB = np.zeros(n_patterns)
187
188     len_non_filled = 0
189
190     for pattern in samples:
191         #in case samples array is not full.
192         if(np.array_equal(pattern, [0,0,0])):
193             continue
194         else:
195             len_non_filled +=1
196

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```

197         #convert to 0 and 1 bits
198         binary_pattern = []
199         for bit in pattern:
200             if bit == 1:
201                 binary_pattern.append(1)
202             else:
203                 binary_pattern.append(0)
204
205         # convert binary pattern to decimal index
206         idx = 0
207         n_bits = len(samples[0])
208         for i,bit in enumerate(binary_pattern):
209             idx += bit * (2**(n_bits-i-1))
210         #print("Index: ", idx)
211         PB[idx] +=1
212     #normalize
213     PB = PB/len_non_filled
214     return PB
215
216
217 def compute_energy_function(w, h, v , theta_v, theta_h
218 ):
219     term1_sum1 = 0
220     for i in range(0,len(h)):
221         term1_sum2 = 0
222         for j in range(0,len(v)):
223             term1_sum2 += w[i][j]*h[i]*v[j]
224         term1_sum1 += term1_sum2
225
226     term2_sum = 0
227     for j in range(0,len(theta_v)):
228         term2_sum += theta_v[j]*v[j]
229
230     term3_sum = 0
231     for i in range(0,len(theta_h)):
232         term3_sum += theta_h[i]*h[i]
233
234     return -term1_sum1 + term2_sum + term3_sum
235
236 def main():
237     ### Init ###
238     # pattern = 0.25 for index 0,3,5,6, used when
239     # sampling
240     all_patterns = np.array([[ -1, -1, -1],
241                             [ -1, -1, 1],

```

```

241         [-1, 1, -1],
242         [-1, 1, 1],
243         [1, -1, -1],
244         [1, -1, 1],
245         [1, 1, -1],
246         [1, 1, 1]])
247 M_values = [1, 2, 4, 8]
248 d_kl_bound_values = []
249 d_kl_values = []
250 eta = 0.006
251 k = 10
252 batch_size = 100
253 epochs = 1500
254
255 # sample using the dynamincs in the CD_k algorithm
256 #
257 num_iterations = 10000
258 max_T = 10
259 print(f"Sampling: num_iterations={num_iterations},
260       T={max_T}")
261 #####
262
263 for M in M_values:
264     print("-----")
265     print(f"Training configuration: M={M} | eta={
266           eta} | k={k} | batch_size={batch_size} |
267           epochs={epochs}")
268
269     w, theta_h, theta_v = train_RBM(M,eta,k,epochs
270                                     ,batch_size)
271     print(f"\n w={w} | theta_h={theta_h} |
272           theta_v={theta_v}")
273
274     PD = np.array([0.25, 0, 0, 0.25, 0, 0.25,
275                   0.25, 0])
276
277     h = np.zeros(M)
278
279     samples = np.zeros((num_iterations,3))
280
281     for step in range(0,num_iterations):
282         v = all_patterns[np.random.randint(
283             all_patterns.shape[0])].copy()
284         b_j_v = np.zeros(len(v))
285         b_i_h = np.zeros(len(h))

```

```

279         for T in range(0,max_T):
280             ### update hidden neurons ###
281             for i in range(0,M):
282                 b_i_h[i] = np.dot(w[i],v)-theta_h[
                    i]
283                 p_B_h = p_boltzmann(b_i_h[i])
284                 r = np.random.rand()
285                 if(r < p_B_h):
286                     h[i] = 1
287                 else:
288                     h[i] = -1
289             #####
290
291             ### update visible neurons ###
292             for j in range(0,3):
293                 b_j_v[j] = np.dot(h,w[:,j]-theta_v
                    [j])
294                 p_B_v = p_boltzmann(b_j_v[j])
295                 r = np.random.rand()
296                 if(r < p_B_v):
297                     v[j] = 1
298                 else:
299                     v[j] = -1
300             #####
301
302
303         samples[step] = v
304
305         #Implement early stopping since i need
            different "num_iterations" for
            different M. modulo 10 for performance
            reasons
306         if(step % 10 == 0):
307             tmp = compute_frequencies(samples)
308             print(f"runtime PD: {tmp}, iteration:
                {step}")
309
310
311         print(f"Results:\nM={M}")
312         PB = compute_frequencies(samples)
313         print("PB: ", PB, "sum =", np.sum(PB))
314         print("PD: ", PD)
315
316         d_kl_bound = D_kl_bound(M)
317         d_kl_bound_values.append(d_kl_bound)
318

```



```

319         d_kl = D_kl(PD,PB)
320         d_kl_values.append(d_kl)
321         print(f"D_KL: {d_kl}, D_KL_bound: {d_kl_bound}
322               ")
323         print("-----")
324     print("D_KL: " ,d_kl_values," ", "D_KL_bound: ",
325           d_kl_bound_values)
326     plt.plot(M_values,d_kl_bound_values,marker='o',
327              color='green', markersize=4, linestyle='-',
328              label='Bound')
329     plt.plot(M_values,d_kl_values,marker='o', color='
330              orange', markersize=4, linestyle='--',label='
331              True value')
332     plt.title('Kullback-Leibler divergence bound vs
333              true value')
334     plt.xlabel('Number of hidden neurons (M)')
335     plt.ylabel('D_KL')
336     plt.grid()
337     plt.legend()
338     plt.tight_layout()
339     plt.show()
340
341 if __name__ == "__main__":
342     main()

```