

# Assignment 1

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This is an example of how to cite: Darwiche (2013) proposed a new approach for inference in Bayesian networks (BNs) (Pearl, 1988) based on partial differentiation called *Arithmetic Circuits* (Darwiche, 2013).

This template is organized as follows. In Section 1, ...

## 1. Example Section

A *joint probability distribution* (JPD) is function  $P$  on the Cartesian product  $V$  of the variable domains such that the following two conditions hold:

$$0 \leq P(v) \leq 1.0 \text{ for each configuration } v \in V; \text{ and} \quad (1)$$

$$\sum_{v \in V} P(v) = 1.0 \quad (2)$$

Referring to Equation 2 here.

Table 1: Table (a) provides the prior probability of variable  $A$  and Table (b) provides the conditional probability of  $B$  given  $A$ .

(a)		(b)		
$A$	$P(A)$	$A$	$B$	$P(B A)$
1	0.3	1	1	0.1
0	0.7	1	0	0.9
		0	1	0.8
		0	0	0.2

Referring to Table 1 (a) here.

### 1.1 Example Subsection

This is how you use math items inline:  $(\theta_a \star \lambda_a)$ ,  $(\theta_{ab} \star \lambda_b)$ ,  $(\theta_{a\bar{b}} \star \lambda_{\bar{b}})$ ,  $(\lambda_b \star \theta_{a\bar{b}})$ ,  $(\lambda_{\bar{b}} \star \theta_{a\bar{b}})$ , and  $(\lambda_{\bar{a}} \star \theta_{\bar{a}})$ .

Referring to Figure 1.

```

1 import numpy as np
2 from keras.models import Sequential
3 from keras.layers.core import Dense, Dropout, Layer, Activation
4 import time
5 import tensorflow as tf
6
7 f = open("results.csv", "w")
8
9
10 INPUT_SIZE = 10
11 OUTPUT_SIZE = INPUT_SIZE
12 nb_class = 3
13
14 batch_size = 128
15 nb_epoch = 40
16
17 np.random.seed(123)
18
19 X_train = np.random.rand(INPUT_SIZE, nb_class)
20 Y_train = np.random.rand(OUTPUT_SIZE, nb_class)
21
22 X_test = np.random.rand(INPUT_SIZE)
23 Y_test = np.random.rand(OUTPUT_SIZE)
24
25 for i in range(1,51):
26     start_time = time.time()
27
28     model = Sequential()
29     model.add(Dense(INPUT_SIZE, input_shape=(nb_class,)))
30     model.add(Activation('linear'))
31     model.add(Dense(OUTPUT_SIZE))
32     model.add(Activation('linear'))
33     model.compile(loss='categorical_crossentropy', optimizer='rmsprop')
34
35     final_time = time.time()
36     diff_time = final_time - start_time
37
38     f.write(str(i)+", "+str(diff_time)+" ,"+str("\n"))
39
40
41 f.close()

```

Table 2: Comparison of parallel and serial solutions for ACs with 50 runs each

Solution	Time average	Standard deviation
Serial	0.1610	0.0536
Parallel	0.0434	0.0082

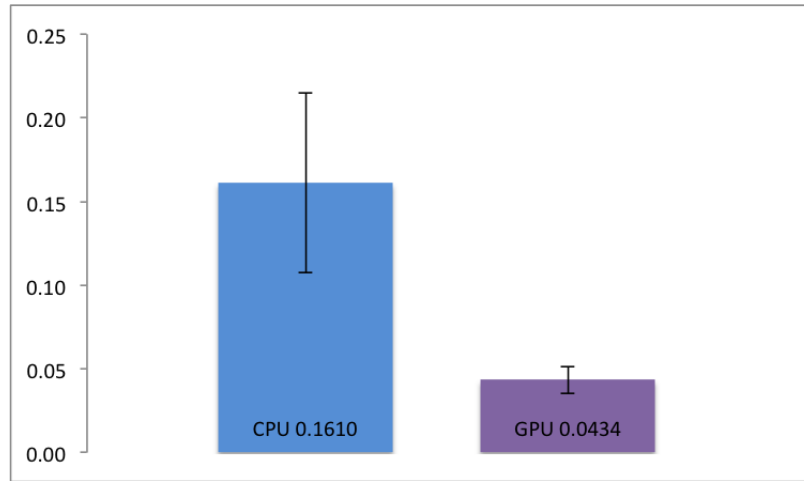


Figure 1: Comparison between CPU and GPU implementation of AC compiling.

## 2. Conclusion

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## References

- Adnan Darwiche. A differential approach to inference in bayesian networks. *Proceedings of the Sixteenth Conference on Uncertainty in Artificial Intelligence (UAI2000)*, 2013.
- J. Pearl. *Probabilistic Reasoning in Intelligent Systems: Networks of Plausible Inference*. Morgan Kaufmann, 1988.