

# Introduction to Deep Neural Networks

## GPU computing perspective

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### Abstract

Some abstract text

**Keywords:** Deep Learning, Deep Neural Networks, GPU computing

## **1. Getting Started**

### **1.1 Basics in machine learning**

### **1.2 Preprocessing**

### **1.3 Gradient-based optimization**

## **2. Classification**

### **2.1 What is classification?**

### **2.2 Logistic regression and softmax regression**

### **2.3 Linear support vector machine**

### 3. Multilayer Perceptron

## 4. Auto-encoders

## 5. Convolutional Neural Networks

## 6. Long-short Term Memory

## 7. General Discussion

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## Appendix A.

In this appendix we prove the following theorem from Section 6.2:

**Theorem** *Let  $u, v, w$  be discrete variables such that  $v, w$  do not co-occur with  $u$  (i.e.,  $u \neq 0 \Rightarrow v = w = 0$  in a given dataset  $\mathcal{D}$ ). Let  $N_{v0}, N_{w0}$  be the number of data points for which  $v = 0, w = 0$  respectively, and let  $I_{uv}, I_{uw}$  be the respective empirical mutual information values based on the sample  $\mathcal{D}$ . Then*

$$N_{v0} > N_{w0} \Rightarrow I_{uv} \leq I_{uw}$$

*with equality only if  $u$  is identically 0.* ■

**Proof.** We use the notation:

$$P_v(i) = \frac{N_v^i}{N}, \quad i \neq 0; \quad P_{v0} \equiv P_v(0) = 1 - \sum_{i \neq 0} P_v(i).$$

These values represent the (empirical) probabilities of  $v$  taking value  $i \neq 0$  and 0 respectively. Entropies will be denoted by  $H$ . We aim to show that  $\frac{\partial I_{uv}}{\partial P_{v0}} < 0, \dots$

*Remainder omitted in this sample. See <http://www.jmlr.org/papers/> for full paper.*

## References

- C. K. Chow and C. N. Liu. Approximating discrete probability distributions with dependence trees. *IEEE Transactions on Information Theory*, IT-14(3):462–467, 1968.