

## Contents

<b>1</b>	<b>Illustration of Hoeffding's Inequality</b>	<b>2</b>
1.1	Plot for Empirical Frequency, Markov bound, chebychev bound and hoeffding bound . . . . .	2
1.2	Comparison of Hoeffding bound with other plots . . . . .	2
1.3	Comparison of Exact Probability for $\alpha = 0.95, 1.0$ with Hoeffding bound . . .	2
<b>2</b>	<b>The effect of scale (range) and normalisation of random variables in Hoeffding's inequality</b>	<b>3</b>
<b>3</b>	<b>Distribution of Student's Grades</b>	<b>4</b>
3.1	Markov's Inequality . . . . .	4
3.2	Chebychev Inequality . . . . .	4
3.3	Hoeffding inequality . . . . .	5
3.4	non-vacuous value of $z$ . . . . .	6
<b>4</b>	<b>The Airline Question</b>	<b>6</b>
4.1	Bound the probability that the number of people that show up for a flight will be larger than the number of seats . . . . .	6
4.2	. . . . .	7
<b>5</b>	<b>Logistic Regression</b>	<b>8</b>
5.1	Cross-entropy error measure . . . . .	8
5.2	Logistic regression loss gradient . . . . .	8
5.3	Logistic regression implementation . . . . .	8
5.4	Iris flower data . . . . .	8

# 1 Illustration of Hoeffding's Inequality

## 1.1 Plot for Empirical Frequency, Markov bound, chebychev bound and hoeffding bound

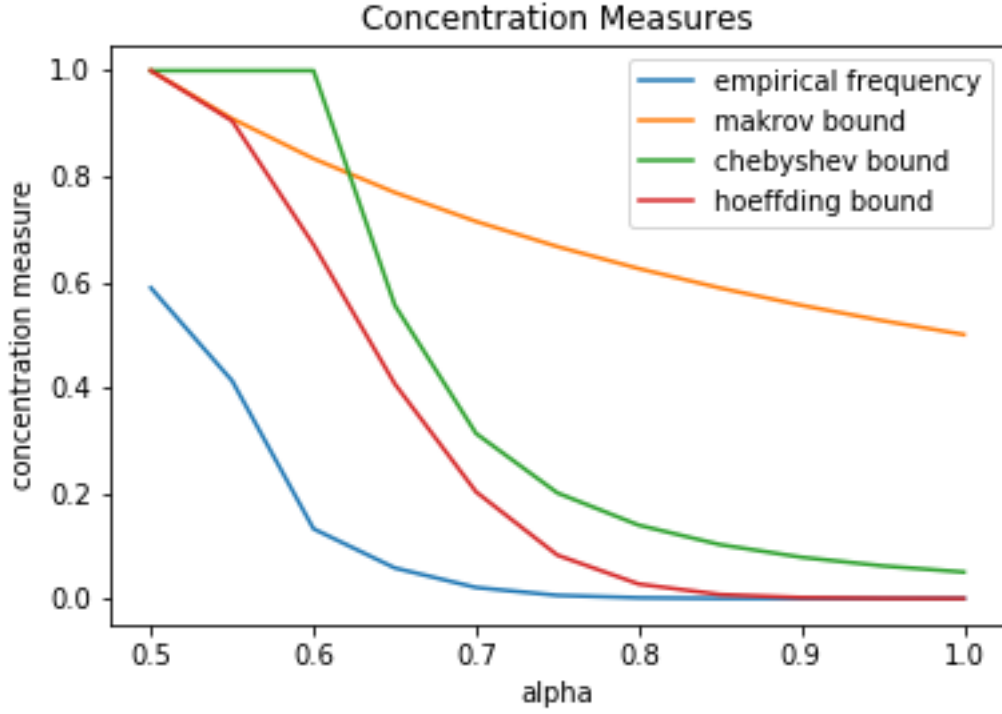


Figure 1: Empirical Frequency, Markov bound, chebychev bound and hoeffding bound

## 1.2 Comparison of Hoeffding bound with other plots

Hoeffding plot gives a better and tighter bound than both markov and chebychev bounds. For  $\alpha \geq 0.85$ , it gives a very tight bound over the empirical frequency and coincides with the empirical frequency plot.

## 1.3 Comparison of Exact Probability for $\alpha = 0.95, 1.0$ with Hoeffding bound

$$P\left(\frac{1}{20} \sum_{i=1}^{20} X_i \geq 0.95\right) = 21 * \left(\frac{1}{20}\right)^{20} = 2.0027161e - 25 \quad (1)$$

$$P\left(\frac{1}{20} \sum_{i=1}^{20} X_i \geq 0.1\right) = \left(\frac{1}{20}\right)^{20} = 9.5367432e - 27 \quad (2)$$

Hoeffding bounds for  $\alpha = 0.95, 1.0$  are  $0.00030353913807886244$ ,  $4.539992976248405e - 05$  respectively.

Both are very very small and close to zero.

## 2 The effect of scale (range) and normalisation of random variables in Hoeffding's inequality

Corollary 2.5 : Let  $X_1, \dots, X_n$  be independent random variables, such that  $X_i \in [0, 1]$ ,  $E[X_i] = \mu$  for all  $i$ , then for every  $\epsilon > 0$ :

$$P\left(\frac{1}{n} \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{-2n\epsilon^2} \quad (3)$$

We need to prove above corollary.

Let  $P(X_i) = p$  and  $X_i \in [0, 1]$

So,  $\mu = E[X_i] = 0 * p + 1 * p = p$

$$E\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n E[X_i] = n * p$$

From the Hoeffding's inequality we have

$$P\left(\left|\sum_{i=1}^n X_i - E\left[\sum_{i=1}^n X_i\right]\right| \geq \epsilon'\right) \leq e^{\frac{-2\epsilon'^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (4)$$

$$P\left(\left|\sum_{i=1}^n X_i - n * p\right| \geq \epsilon'\right) \leq e^{\frac{-2\epsilon'^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (5)$$

$$P\left(\frac{1}{n} \left|\sum_{i=1}^n X_i - n * p\right| \geq \frac{1}{n} * \epsilon'\right) \leq e^{\frac{-2\epsilon'^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (6)$$

$$P\left(\frac{1}{n} * \sum_{i=1}^n X_i - p \geq \frac{1}{n} * \epsilon'\right) \leq e^{\frac{-2\epsilon'^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (7)$$

$$P\left(\frac{1}{n} * \sum_{i=1}^n X_i - \mu \geq \frac{1}{n} * \epsilon'\right) \leq e^{\frac{-2\epsilon'^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (8)$$

Let  $\epsilon' = n * \epsilon$

$$P\left(\frac{1}{n} * \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{\frac{-2 * (n * \epsilon)^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (9)$$

$X_i \in [0, 1]$

So,  $a_i = 0$  and  $b_i = 1$

$$\sum_{i=1}^n (b_i - a_i)^2 = n$$

$$P\left(\frac{1}{n} * \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{\frac{-2 * (n * \epsilon)^2}{n}} \quad (10)$$

$$P\left(\frac{1}{n} * \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{-2 * n * (\epsilon)^2} \quad (11)$$

Hence, proved

### 3 Distribution of Student's Grades

Let  $X_1, X_2, \dots, X_7$  be the scores in the assignment.

$$\hat{Z} = \frac{1}{7} * \sum_{i=1}^7 X_i$$

$$E[X_i] = p \text{ for all } i$$

For  $p = 50$  and  $\delta = 0.05$

We need to find maximal value of  $z$  such that  $P(\hat{Z} \leq z) \leq \delta$

#### 3.1 Markov's Inequality

Let  $Q = 100 - \hat{Z}$

From Markov's Inequality,

$$P(Q \geq \epsilon) \leq \frac{E[Q]}{\epsilon} \quad (12)$$

$$P(100 - \hat{Z} \geq \epsilon) \leq \frac{E[100 - \hat{Z}]}{\epsilon} \quad (13)$$

$$P(\hat{Z} \leq 100 - \epsilon) \leq \frac{100 - E[\hat{Z}]}{\epsilon} \quad (14)$$

$$P(\hat{Z} \leq 100 - \epsilon) \leq \frac{100 - E[\frac{1}{7} * \sum_{i=1}^7 X_i]}{\epsilon} \quad (15)$$

$$P(\hat{Z} \leq 100 - \epsilon) \leq \frac{100 - \frac{1}{7} * \sum_{i=1}^7 E[X_i]}{\epsilon} \quad (16)$$

$$P(\hat{Z} \leq 100 - \epsilon) \leq \frac{100 - p}{\epsilon} \quad (17)$$

Let  $z = 100 - \epsilon$

$$P(\hat{Z} \leq z) \leq \frac{100 - p}{\epsilon} \quad (18)$$

Thus,

$$\frac{100 - p}{100 - z} \leq 0.05 \quad (19)$$

$$\frac{50}{100 - z} \leq 0.05 \quad (20)$$

$$z \leq -900 \quad (21)$$

#### 3.2 Chebychev Inequality

From chebychev Inequality,

$$P(|Z - E[Z]| \geq \epsilon) \leq \frac{Var[Z]}{\epsilon^2} \quad (22)$$

$$(23)$$

We have

$$P(\hat{Z} \leq z) \leq 0.05 \quad (24)$$

$$P(\hat{Z} - p \leq z - p) \leq 0.05 \quad (25)$$

$$P(|\hat{Z} - p| \geq p - z) \leq 0.05 \quad (26)$$

By using chebychev inequality, we can also say that

$$P(|\hat{Z} - p| \geq p - z) \leq \frac{Var[\hat{Z}]}{(p - z)^2} \quad (27)$$

By using last two equations we can conclude that

$$\frac{Var[\hat{Z}]}{(p - z)^2} \leq 0.05 \quad (28)$$

$$\frac{Var[\hat{Z}]}{(p - z)^2} \leq 0.05 \quad (29)$$

$$Var[\hat{Z}] \leq 0.05 * (p - z)^2 \quad (30)$$

The variance of a random variable  $X \in [a, b]$  is maximized when  $X = a$  with probability  $1/2$  and  $X = b$  with probability  $1/2$ . In other words, let  $Y$  be a random variable, such that  $P(Y = a) = 1/2$  and  $P(Y = b) = 1/2$ , then for any random variable  $X \in [a, b]$  we have  $Var[X] \leq Var[Y]$ .

$a = 0, b = 100, \mu = 50$

$$Var[Y] = \frac{(0-50)^2 + (100-50)^2}{2} = 2500$$

$$Var[\hat{Z}] = Var[\frac{1}{7} \sum_{i=1}^7 X_i] = \frac{1}{7^2} \sum_{i=1}^7 Var[X_i] = \frac{1}{7} Var[X_i] \leq \frac{1}{7} Var[Y_i] \leq \frac{2500}{7} \quad (31)$$

From last two equations we can say that

$$\frac{2500}{7} \leq 0.05 * (p - z)^2 \quad (32)$$

$$\frac{2500}{7 * 0.05} \leq (50 - z)^2 \quad (33)$$

$$\sqrt{\frac{2500}{7 * 0.05}} \leq 50 - z \quad (34)$$

$$z \leq 50 + \sqrt{\frac{2500}{7 * 0.05}} \quad (35)$$

### 3.3 Hoeffding inequality

From Hoeffding inequality, where  $X_i \in [a_i, b_i]$

$$P(|\sum_{i=1}^n X_i - E[\sum_{i=1}^n X_i]| \geq \epsilon) \leq e^{\frac{-2\epsilon^2}{\sum_{i=1}^n (b_i - a_i)^2}} \quad (36)$$

$$(37)$$

We have

$$P(\hat{Z} \leq z) \leq 0.05 \quad (38)$$

$$P(\hat{Z} - p \leq z - p) \leq 0.05 \quad (39)$$

$$P(|\hat{Z} - p| \geq p - z) \leq 0.05 \quad (40)$$

Let  $\epsilon = p - z$

From Hoeffding inequality, we can make the following inference

here  $[a_i, b_i] = [0, 100]$

$$e^{\frac{-2\epsilon^2}{\sum_{i=1}^7 (100-0)^2}} \leq 0.05 \quad (41)$$

$$\frac{-2\epsilon^2}{7 * 100^2} \leq \ln 0.05 \quad (42)$$

$$\frac{-2(50 - z)^2}{7 * 100^2} \leq \ln 0.05 \quad (43)$$

### 3.4 non-vacuous value of z

Markov inequality gives non-vacuous value. In this case, it gives bound smaller than 0.

## 4 The Airline Question

### 4.1 Bound the probability that the number of people that show up for a flight will be larger than the number of seats

$P(\text{the person will not show up}) = 0.05$  Let  $X_i$  be the random variable that the  $i^{th}$  person shows up.

$X_i = 1$ , when the  $i^{th}$  person shows up else 0.

$\mu = E[X_i] = 0.95 * 1 + 0.05 * 0 = 0.95$

Let Z be the total number of people showed up

$Z = \sum_{i=1}^{100} X_i$

$$P(Z \geq 100) = P\left(\frac{Z}{100} \geq 1\right) \quad (44)$$

$$= P\left(\frac{Z}{100} - \mu \geq 1 - \mu\right) \quad (45)$$

We can compare the above equation with the following hoeffding inequality

$$P\left(\frac{1}{n} \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{-2n\epsilon^2} \quad (46)$$

Let  $\epsilon = 1 - \mu$   
 So, we can say that

$$P\left(\frac{Z}{100} - \mu \geq 1 - \mu\right) \leq e^{-2*100*(1-\mu)^2} \quad (47)$$

$$\leq e^{-0.5} \quad (48)$$

## 4.2

Event 1: In the sample of 10000 passengers, where each passenger shows up with probability  $p$ , we observe 95 percent shows up

Event 2 : In the sample of 100 passengers, where each passenger shows up with probability  $p$ , everybody shows up

$$P(\text{Event 2}) = p^{100}$$

Let  $X_i$  be the random variable that the  $i^{th}$  person shows up.

$X_i = 1$ , when the  $i^{th}$  person shows up else 0.

$$\mu = E[X_i] = p * 1 + (1 - p) * 0 = p$$

Let  $Z$  be the total number of people showed up

$$Z = \sum_{i=1}^{10000} X_i$$

$$P(\text{Event1}) = P(Z \geq 0.95 * 10000) \quad (49)$$

$$= P\left(\frac{Z}{10000} \geq 0.95\right) \quad (50)$$

$$= P\left(\frac{Z}{100} - \mu \geq 0.95 - \mu\right) \quad (51)$$

We can compare the above equation with the following hoeffding inequality

$$P\left(\frac{1}{n} \sum_{i=1}^n X_i - \mu \geq \epsilon\right) \leq e^{-2n\epsilon^2} \quad (52)$$

Let  $\epsilon = 0.95 - \mu$

So, we can say that

$$P\left(\frac{Z}{10000} - \mu \geq 0.95 - \mu\right) \leq e^{-2*10000*(0.95-\mu)^2} \quad (53)$$

$$\leq e^{-2*10000*(0.95-p)^2} \quad (54)$$

$$P(\text{Event 1 and Event 2 happens simultaneously}) = P(\text{Event1}) P(\text{Event2})$$

$$P(\text{Event1})P(\text{Event2}) \leq p^{100}e^{-2*10000*(0.95-p)^2} \quad (55)$$

## **5 Logistic Regression**

### **5.1 Cross-entropy error measure**

### **5.2 Logistic regression loss gradient**

### **5.3 Logistic regression implementation**

### **5.4 Iris flower data**