

# HW1

1. 已知某工厂某批次水泥重量服从正态分布，总体方差为2.65公斤，从该工厂随机抽取 18 袋水泥，其平均重量为 24.9公斤，试求该工厂水泥平均重量的95%和99%的置信区间。

$$SE_X = \frac{\sigma}{\sqrt{n}} = \frac{\sqrt{2.65}}{\sqrt{18}} = 0.38$$

95%置信区间的显著性水平 $\alpha = 0.05$ ,  $Z_{\alpha/2} = 1.96$

95%置信区间为:  $24.9 - 1.96 \times 0.38 \leq u \leq 24.9 + 1.96 \times 0.38$ , 即[24.16, 25.64] (1)

99%置信区间的显著性水平 $\alpha = 0.01$ ,  $Z_{\alpha/2} = 2.58$

99%置信区间为:  $24.9 - 2.58 \times 0.38 \leq u \leq 24.9 + 2.58 \times 0.38$ , 即[23.92, 25.88]

2. 从某学校抽取20名高一学生，经测量，这 20名学生的平均身高为155cm，标准差为10cm，假设平均身高服从正态分布，试求该学校高一学生总平均身高的95%和99%的置信区间。

$$SE_X = \frac{S}{\sqrt{n-1}} = \frac{10}{\sqrt{19}} = 2.29$$

$t_{0.05/2(19)} = 2.093$ ,  $t_{0.01/2(19)} = 2.861$  (2)

95%置信区间为:  $155 - 2.093 \times 2.29 \leq u \leq 155 + 2.093 \times 2.29$ , 即[150.21, 159.79]

99%置信区间为:  $155 - 2.861 \times 2.29 \leq u \leq 155 + 2.861 \times 2.29$ , 即[148.45, 161.55]

3. 1.2 节P19，证明被抽样总体分布未知情况下的无偏估计。

**例** 设 $x_1, x_2, \dots, x_n$ 是均值为 $\mu$ 、方差为 $\sigma^2$ 的随机变量X的n个观测值的随机样本，证明：样本方差 $s^2$ 是总体方差 $\sigma^2$ 的一个无偏估计，其中：

- a) 被抽样总体为正态分布
- b) 被抽样总体的分布未知

设样本均值为 $\mu$

$$\begin{aligned}
 s^2 &= \frac{\sum (x_i - \mu)^2}{n - 1} \\
 &= \frac{\sum (x_i - x)^2 + 2(x - \mu) \times \sum (x_i - x) + n(\mu - x)^2}{n - 1} \\
 &= \frac{\sum (x_i - x)^2 + 2(x - \mu) \times (n(\mu - x)) + n(\mu - x)^2}{n - 1} \\
 &= \frac{\sum (x_i - x)^2 - n(x - \mu)^2}{n - 1}
 \end{aligned}$$

$$\begin{aligned}
 E[(x - \mu)^2] &= \text{Var}[x - \mu] + ([E(x - \mu)]^2) \\
 &= \text{Var}\left[\frac{\sum (x - x_i)}{n}\right] + \left(x - \frac{Ex_i}{n}\right)^2 \\
 &= \frac{\sum \text{Var}[x_i - x]}{n^2} + \left(x - \frac{nx}{n}\right)^2 \\
 &= \frac{n\sigma^2}{n^2} \\
 &= \frac{\sigma^2}{n} \\
 E[s^2] &= E\left[\frac{\sum (x_i - x)^2 - n(x - \mu)^2}{n - 1}\right] \\
 &= \frac{n \times (E[\frac{\sum (x_i - x)^2}{n}] - E[\mu - x]^2)}{n - 1} \\
 &= \frac{n \times \sigma^2 - n \times \frac{\sigma^2}{n}}{n - 1} \\
 &= \sigma^2
 \end{aligned} \tag{3}$$

#### 4. 证明Lemma 1.3.1

**Lemma 1.3.1.** Let  $P$  be the transition probability matrix for a connected Markov Chain. The  $n \times (n + 1)$  matrix  $A = [P - I, 1]$  obtained by augmenting the matrix  $P - I$  with an additional column of ones has rank  $n$ .

反证法, 若  $\text{rank}(A) < n$

则  $A\vec{x} = 0$  存在两个线性无关解

注意到  $\sum P_{ij} = 0$ , 因此  $(\vec{1}, 0)^T$  是方程的解

因此存在第二个解, 设为  $(\vec{y}, \alpha)^T$

$$\Rightarrow A \cdot (\vec{y}, \alpha)^T = 0 \Rightarrow (P - I, \vec{1}) \cdot (\vec{y}, \alpha)^T = 0$$

$$\Rightarrow (P - I) \cdot \vec{y} + \alpha \cdot \vec{1} = 0$$

对第  $i$  个:  $\sum_j P_{ij} \cdot y_j - y_i + \alpha = 0$

$$\Rightarrow y_i = \sum_j P_{ij} \cdot y_j + \alpha$$

由于  $(\vec{y}, \alpha)^T$  与  $(\vec{1}, 0)$  线性无关

因此  $y_i$  互不相等

$$\Rightarrow \exists i_m, i_n \text{ s.t. } \forall k \in N, k \leq n \quad y_{i_m} \leq y_k \leq y_{i_n} \text{ 且 } y_{i_m} > y_{i_n}$$

$$\Rightarrow y_{i_m} > \sum_j P_{imj} \cdot y_j, \quad y_{i_n} < \sum_j P_{inj} \cdot y_j$$

$$\Rightarrow 0 < \alpha < 0, \text{ 矛盾}$$

实验1: PageRank最高20份node:

```
(py) jyjs@192 HW % python pagerank.py
[('7237', np.float64(0.007402252556857881)), ('7498', np.float64(0.006691202515011874)), ('7339', np.float64(0.006377762845352468)), ('7162', np.float64(0.0031767569339574593)), ('7224', np.float64(0.0029302963203570952)), ('7435', np.float64(0.002750701330228721)), ('6519', np.float64(0.0027171361605782102)), ('7100', np.float64(0.0025939892857066867)), ('7595', np.float64(0.002575726801676199)), ('7199', np.float64(0.002575281129187989)), ('4811', np.float64(0.0024343041872750726)), ('6101', np.float64(0.0023085323564403337)), ('7536', np.float64(0.002238758251594308)), ('4785', np.float64(0.0022117014168385888)), ('7489', np.float64(0.0019699795701771067)), ('7587', np.float64(0.001849670865613301)), ('7488', np.float64(0.0018347441540418777)), ('6617', np.float64(0.0017652065393296442)), ('7226', np.float64(0.0017385075407141284)), ('7416', np.float64(0.0017249734913823893))]
```

实验2:

结果统计:

```
recall: 0.6818181818181818  
precision: 0.5284974093264249  
f1 score: 0.5954465849387041
```

实验代码在压缩包中