Estimation Candidates

 How do you test whether a given set of real-valued observations are drawn from a Gaussian distribution?

I use P-value approach. we already know the target Gaussian distribution, noted as $\mathcal{N} \sim (\mu_0, \sigma_0^2)$, and a given set of real-valued observations as $(x_1, x_2 \dots x_n)$.

For set of observations, note its mean is μ , its standard deviation is σ , and number of observations is n.

1) Test for mean. (Student's t-test)

 $H_0: \mu = \mu_0$, $H_1: \mu \neq \mu_0$, and confidence level is 95%.

We can derive that confidence interval is

$$U = \left(\mu_0 - 1.96 \frac{\sigma}{\sqrt{n}}, \, \mu_0 + 1.96 \frac{\sigma}{\sqrt{n}}\right)$$

if μ is in confidence interval U, that we can say H_0 is true that its mean is μ_0 . Otherwise, this set of observations are not from this Gaussian distribution.

2) Test for standard deviation. (Chi-squared test)

 $H_0: \sigma = \sigma_0, H_1: \sigma \neq \sigma_0$, and confidence level is 95%.

Note that

$$s^2 = \frac{\sum_{i=1}^{n} (x_i - \mu_0)^2}{n - 1}$$

Test statistics is that

$$\chi^2 = \frac{(n-1)\cdot s^2}{\sigma_0^2}$$

We know that when confidence level is 95%, confidence interval for χ^2 is U'

$$U' = (12.40, 39.36)$$

If χ^2 is in the confidence interval U', that we can say H_0 is true that its standard deviation is σ_0 . Otherwise, this set of observations are not from this Gaussian distribution.

2. Given a streaming time series of real-valued, scalar observations, describe an online algorithm for determining a linear Gaussian model of these observations as a function of time step.

Notations: at time step n, we have mean μ_n , standard deviation σ_n and input at step n is x_n .

$$\mu_n = \frac{\sum_{i=1}^n x_i}{n}$$

$$\sigma_n^2 = \frac{\sum_{i=1}^n (x_i - \mu_n)^2}{n - 1}$$

Therefore, at time step (n-1), we have

$$\mu_{n-1} = \frac{\sum_{i=1}^{n-1} x_i}{n-1}$$

$$\sigma_{n-1}^2 = \frac{\sum_{i=1}^{n-1} (x_i - \mu_{n-1})^2}{n-2}$$

For mean at time step n, we have

$$\mu_n = \frac{\sum_{i=1}^n x_i}{n}$$

$$\mu_n = \frac{n-1}{n} \cdot \frac{\sum_{i=1}^{n-1} x_i}{n-1} + \frac{1}{n} x_n$$

$$\mu_n = \frac{n-1}{n} \mu_{n-1} + \frac{1}{n} x_n$$

For standard deviation at step n, we have

$$\sigma_n^2 = \frac{\sum_{i=1}^n (x_i - \mu_n)^2}{n-1}$$

$$\sigma_n^2 = \frac{1}{n-1} \sum_{i=1}^n ((x_i - \mu_{n-1}) + (\mu_{n-1} - \mu_n))^2$$

$$\sigma_n^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \mu_{n-1})^2 + 2(x_i - \mu_{n-1})(\mu_{n-1} - \mu_n) + (\mu_{n-1} - \mu_n)^2$$

$$\sigma_n^2 = \frac{1}{n-1} \left(\sum_{i=1}^{n-1} (x_i - \mu_{n-1})^2 + (x_n - \mu_{n-1})^2 + \sum_{i=1}^n 2(x_i - \mu_{n-1})(\mu_{n-1} - \mu_n) + \sum_{i=1}^n (\mu_{n-1} - \mu_n)^2 \right)$$

$$\sigma_n^2 = \frac{1}{n-1} \left((n-2)\sigma_{n-1}^2 + n^2(\mu_n - \mu_{n-1})^2 + 2(n \cdot \mu_n - n \cdot \mu_{n-1})(\mu_{n-1} - \mu_n) + n(\mu_{n-1} - \mu_n)^2 \right)$$

$$\sigma_n^2 = \frac{1}{n-1} \left((n-2)\sigma_{n-1}^2 + n^2(\mu_n - \mu_{n-1})^2 - 2n(\mu_{n-1} - \mu_n)^2 + n(\mu_{n-1} - \mu_n)^2 \right)$$

$$\sigma_n^2 = \frac{1}{n-1} \left((n-2)\sigma_{n-1}^2 + n(n-1)(\mu_n - \mu_{n-1})^2 \right)$$

$$\sigma_n^2 = \frac{1}{n-1} \left((n-2)\sigma_{n-1}^2 + n(n-1)(\mu_n - \mu_{n-1})^2 \right)$$

Therefore, we have

$$\mu_n = \frac{n-1}{n} \mu_{n-1} + \frac{1}{n} x_n$$

$$\sigma_n^2 = \frac{n-2}{n-1} \sigma_{n-1}^2 + n(\mu_n - \mu_{n-1})^2$$

We can write the algorithm in Java as following:

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 \begin{array}{l} public \ static \ void \ main(String[] \ args) \{ \\ double \ u = 0, \ r = 0, \ n = 0; \\ Scanner \ scan = new \ Scanner(System.in); \\ double \ x1 = Double.parseDouble(scan.nextLine()); \\ double \ x2 = Double.parseDouble(scan.nextLine()); \\ n=2; \\ u = (x1+x2)/2; \\ r = Math.sqrt((x1-u)*(x1-u)+(x2-u)*(x2-u)) \\ while(in.hasNextLine()) \{ \\ //time \ step \\ n+=1; \\ double \ x_n = Double.parseDouble(scan.nextLine()); \\ double \ u_n = (n-1)/n^*u+1/n^*x_n; \\ double \ r_n = Math.sqrt((n-2)/(n-1)^*r^*r+n^*(u-u_n)^*(u-u_n)); \\ u = u_n; \\ r = r_n; \\ \} \\ System.out.println("The \ mean \ of \ this \ linear \ Gaussian \ model \ is "+u); \\ System.out.println("The \ standard \ deviation \ of \ this \ linear \ Gaussian \ model \ is "+r); \\ \} \end{aligned}
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