

$$17. y_i = \beta_0 + \beta_1 x_i + e_i \quad i=1, \dots, n$$

$$e_i \stackrel{iid}{\sim} N(0, \sigma^2)$$

$$\text{Var}(\hat{\beta}_1) = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\text{Var}(\hat{\beta}_0) = \frac{\sigma^2 \sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}$$

检验 $H_0: \beta_0 = 0$ 统计量及分布

由LSE,

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (x_i - \bar{x}) y_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\text{Var}(\hat{\beta}_1) = \frac{\sum_{i=1}^n (x_i - \bar{x})^2 \text{Var} y_i}{\left(\sum_{i=1}^n (x_i - \bar{x})^2 \right)^2} = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = \frac{\sum_{i=1}^n y_i}{n} - \frac{\sum_{i=1}^n (x_i - \bar{x}) \bar{x} y_i}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$= \sum_{i=1}^n \left(\frac{1}{n} - \frac{\bar{x} (x_i - \bar{x})}{\sum_{j=1}^n (x_j - \bar{x})^2} \right) y_i$$

$$\text{Var}(\hat{\beta}_0) = \sum_{i=1}^n \left(\frac{1}{n} - \frac{\bar{x} (x_i - \bar{x})}{\sum_{j=1}^n (x_j - \bar{x})^2} \right)^2 \sigma^2$$

$$= \frac{\sigma^2 \sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\text{因此 } \hat{\beta}_0 \sim N\left(\beta_0, \frac{\sigma^2 \sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}\right)$$

$$\hat{\sigma}^2 = \frac{1}{n-2} SSE = \frac{1}{n-2} \sum_{i=1}^n (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2 \quad H_0: \beta_0 = 0 \text{ 下有}$$

$$\frac{\hat{\beta}_0}{\hat{\sigma} \sqrt{\frac{\sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}}} \sim t_{n-2}$$

$$z \sim t_{n-2}$$

$$P(|z| > \lambda) = \alpha$$

$$\text{当 } |\hat{\beta}_0| > \lambda \hat{\sigma} \sqrt{\frac{\sum_{i=1}^n x_i^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}}$$

拒绝假设

$$18. \hat{\beta} = (X^T X)^{-1} X^T y = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{pmatrix}$$

$$\hat{\beta}_0 = -11.3 \quad \hat{\beta}_1 = 36.95$$

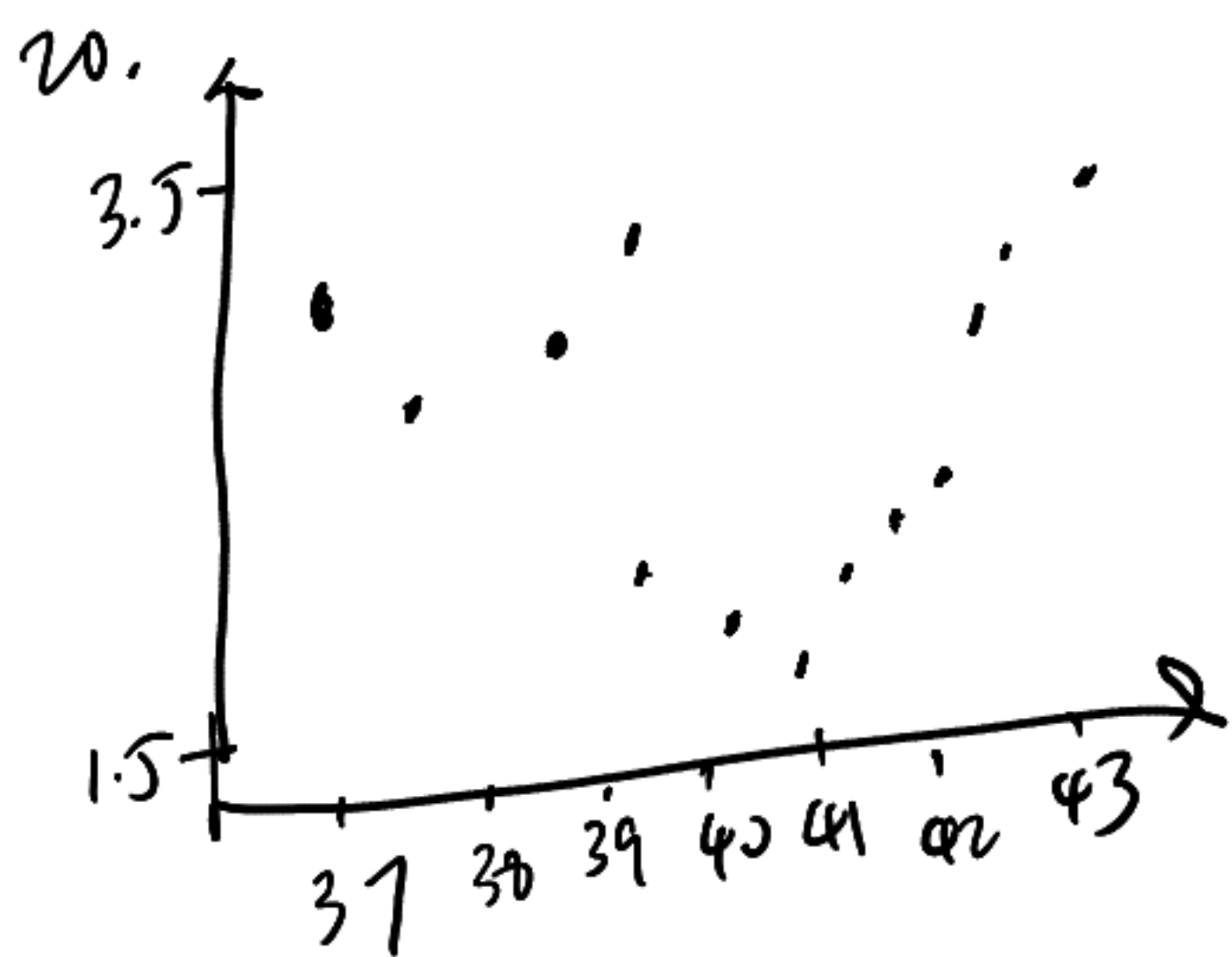
$$X = [x_1 \dots x_n]^T$$

$$y = [y_1 \dots y_n]^T$$

$$T = \lambda \hat{\sigma} \sqrt{\frac{\sum_{i=1}^n (x_i)^2}{n \sum_{i=1}^n (x_i - \bar{x})^2}} = 41.277$$

$$|\hat{\beta}_0| < T \quad \text{不拒绝 } H_0$$

$$\text{即认为 } y = \beta_1 x, \beta_0 = 0$$



从图上看有非线性

可以用二次函数或分段线性函数

二次函数拟合结果:

$$\hat{y} = 0.156x^2 - 12.62x + 257.07$$

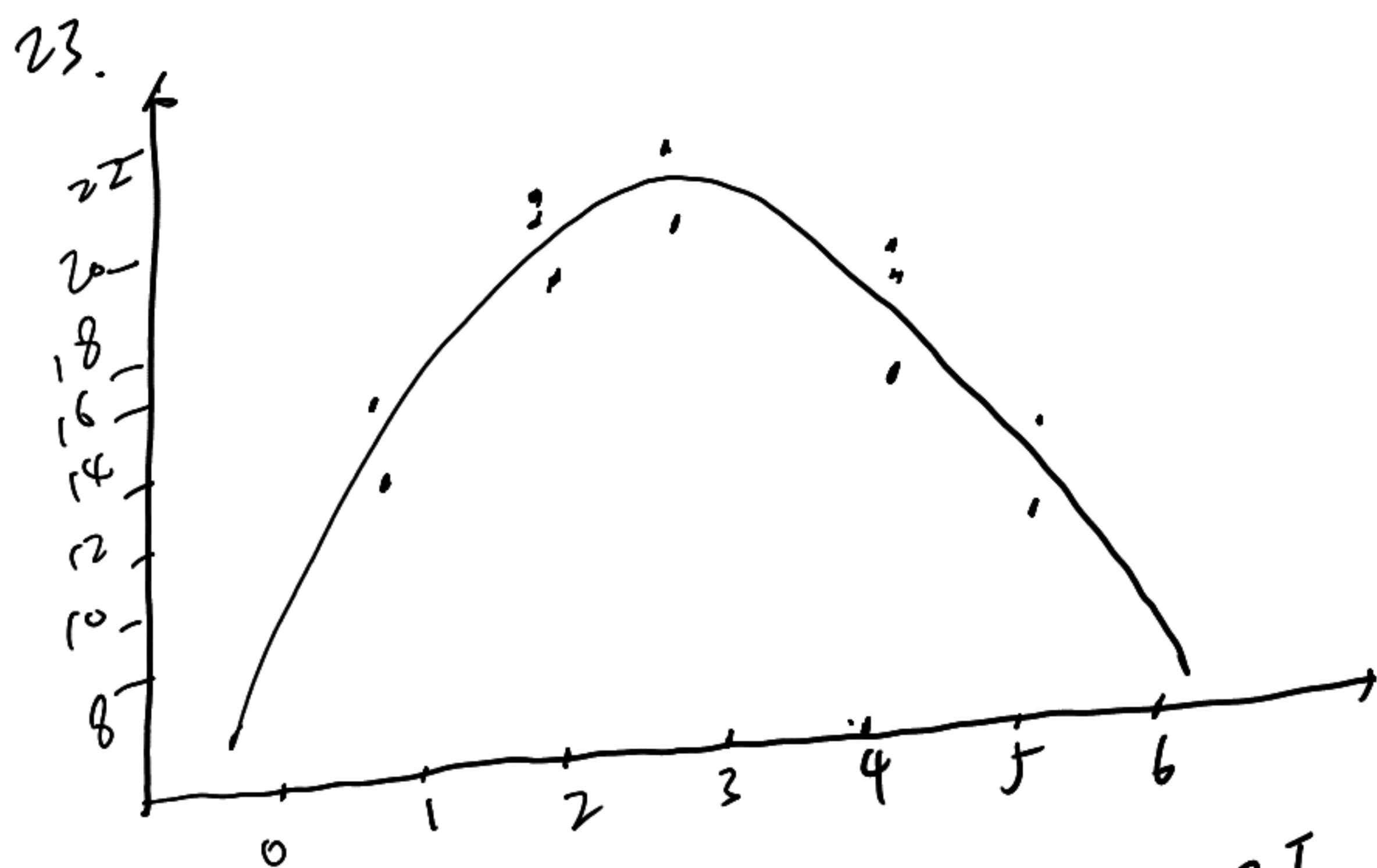
$$\text{对假设 } \beta_0 = \beta_1 = \beta_2 = 0$$

$$\text{有 } F \text{ statistic} = 0.13$$

用分段线性拟合

$$\hat{y} = \begin{cases} -26.92 - 0.63x & x \leq 40 \\ -18.485 + 0.495x & x > 40 \end{cases}$$

效果上这种拟合更好



设 $X = [x_1 \dots x_{12}]^T$ $y = [y_1 \dots y_{12}]^T$

一次拟合 $X^{(1)} = \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_{12} \end{pmatrix}$ $r(X^{(1)})=2$ 拟合结果 $\hat{y} = 0.07x + 18.4$

二次拟合 $X^{(2)} = \begin{pmatrix} 1 & x_1 & x_1^2 \\ \vdots & \vdots & \vdots \\ 1 & x_{12} & x_{12}^2 \end{pmatrix}$ $r(X^{(2)})=3$ $\hat{y} = -1.5x^2 + 9.09x + 7.627$

三次拟合 $X^{(3)} = \begin{pmatrix} 1 & x_1 & x_1^2 & x_1^3 \\ \vdots & \vdots & \vdots & \vdots \\ 1 & x_{12} & x_{12}^2 & x_{12}^3 \end{pmatrix}$ $r(X^{(3)})=4$ $\hat{y} = -0.063x^3 - 0.935x^2 + 7.589x + 8.7301$

分别计算 F statistic ($H_0: \beta_0 = \dots = \beta_{p-1} = 0$) $\alpha = 0.1$

1次: $F = 0.092$

临界值 3.285

2次: $F = 0.324$

临界值 2.923

3次: $F = 0.294$

临界值 2.9605

故选 2次拟合最合适