

Example using the Skui Cream example

$y$  : sales in district

$x_1$  : size of district

$x_2$  : per capita income of district

one can compute:

$$p = 3 \quad B = (X^T X)^{-1} X^T Y = \begin{pmatrix} 3.4526 \\ 0.4960 \\ 0.0092 \end{pmatrix}$$
$$n = 15$$

$$MSE = \frac{1}{p-1} (Y^T Y - B^T X^T Y) = 26922.4$$

$$MSE = 4.74$$

$$F^* = \frac{MSE}{MSE} = 5680$$

Assuming  $\alpha$  at 0.05 and assuming the  $\epsilon_i$  are independent  $N(0, \sigma^2)$ , we

require

$$F(0.95, 2, 12) = 3.89$$

Since  $F^*$  exceeds 3.89 we conclude

$C_2$ : Sales are related to population and income.

But is this relation useful for predictions,

well

$$R^2 = \frac{SSR}{SSTO} = 0.9989$$

So when the independent variables  $x_1$  and  $x_2$  are considered, the variation in sales is

"99.9% explained".

The F-test for deciding if

$\beta_1 = \beta_2 = \dots = \beta_{p-1} = 0$  or not  
in the model

$$y_i = \beta_0 + \beta_1 x_{i1} + \dots + \beta_{p-1} x_{i,p-1} + \varepsilon_i \quad (*)$$

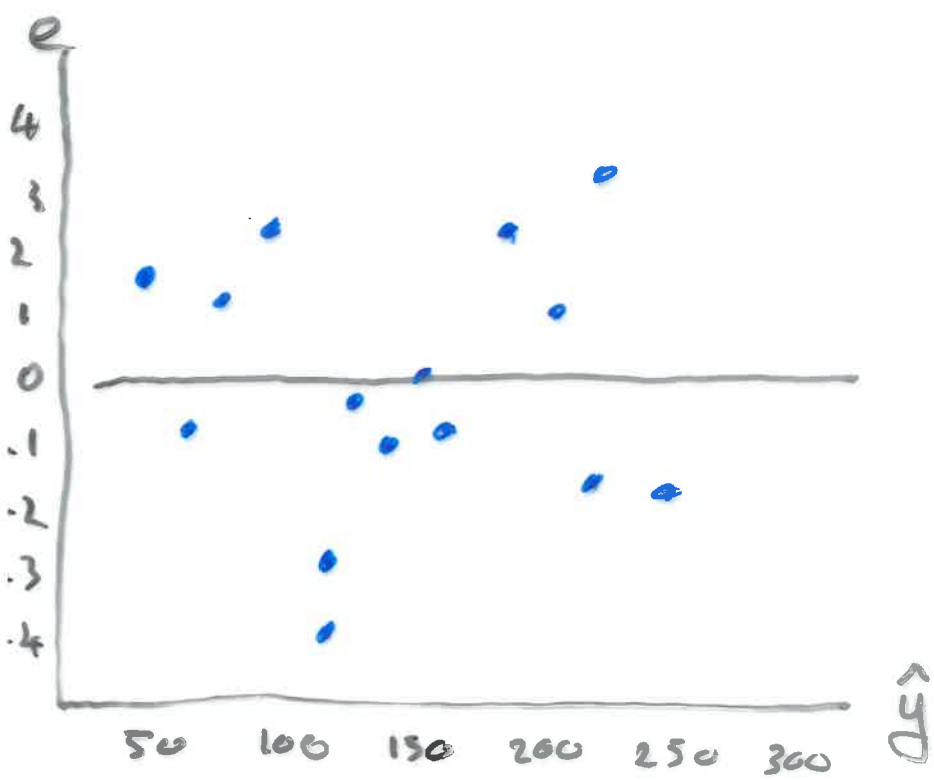
assumes that the  $\varepsilon_i$  are independent  
 $N(0, \sigma^2)$ .

To test these assumptions, in the  
skin cream example, we can plot

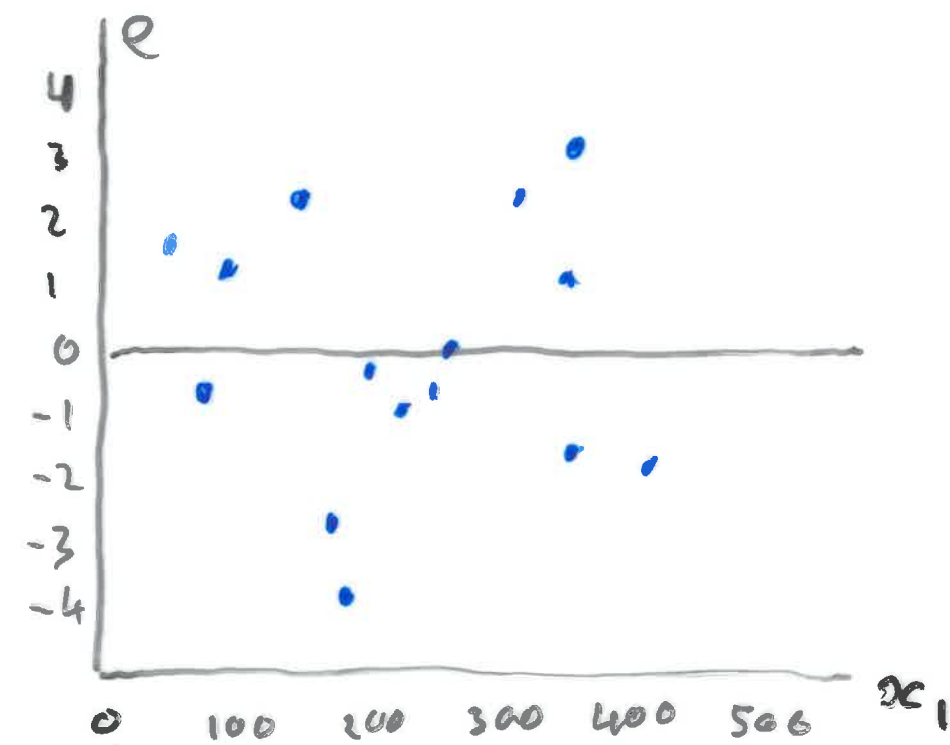
- i)  $\hat{y}_i$  against  $\varepsilon_i$
- ii)  $x_{i1}$  against  $\varepsilon_i$
- iii)  $x_{i2}$  against  $\varepsilon_i$ .

See next slide

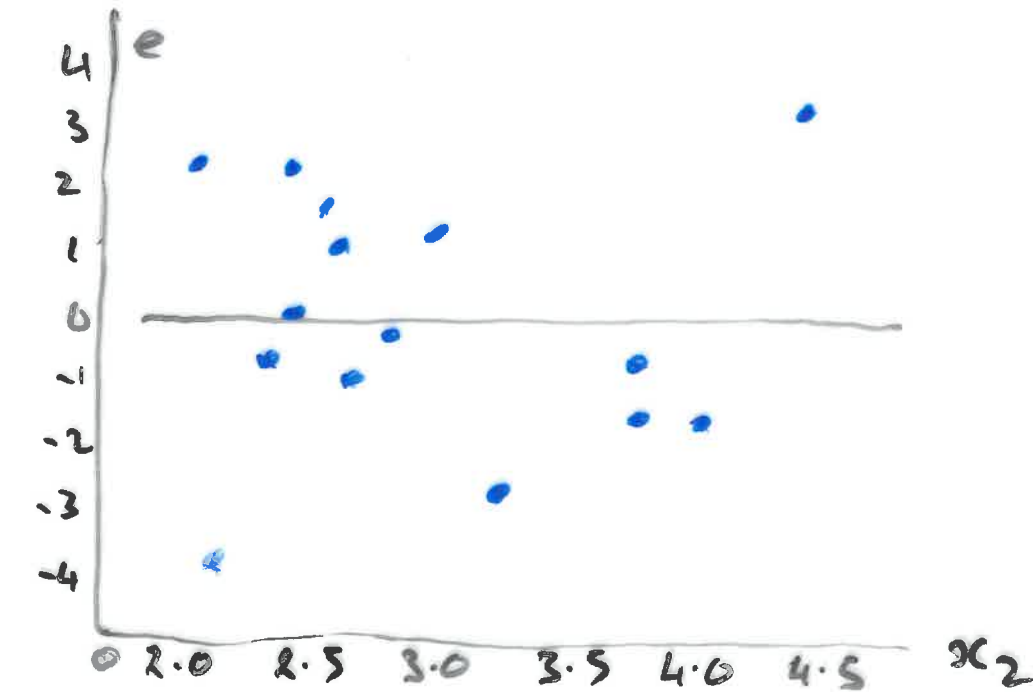
There appears to be no systematic  
deviation, the residuals seem to  
be independent and not depend on  
the level of  $\hat{y}$  or the values of  
 $x_{i1}, x_{i2}$ . So it seems ok to  
accept that  $\varepsilon_i$  are independent  
and  $N(0, \sigma^2)$ .



$y$  vs  $e$



$x_1$  vs  $e$



$x_2$  vs  $e$

Defn The estimated covariance  
matrix for  $(T)$  is

$$S^2(B) = \text{MSE} (X^t X)^{-1}$$

$$= \begin{pmatrix} S^2(b_0) & S(b_0, b_1) & \dots & S(b_0, b_{p-1}) \\ S(b_1, b_0) & S^2(b_1) & \dots & \\ & & \ddots & \\ & & & S^2(b_{p-1}) \end{pmatrix}$$

We only need  $S^2(b_0), S^2(b_1), \dots$ .

Theorem Assume  $\epsilon_i$  are independent  
 $N(0, \sigma^2)$  the quantity

$$\frac{b_k - \beta_k}{S(b_k)}$$

follows a  $t$ -distribution with  
 $n-p$  degrees of freedom.

So, if  $q \leq p$  parameters  $\beta_k$  are to be estimated jointly, the confidence intervals with family coefficient  $1-\alpha$  are:

$$b_k - Ts(b_k) \leq \beta_k \leq b_k + Ts(b_k)$$

where

$$T = t\left(1 - \frac{\alpha}{2q}, n-p\right).$$

Example Continuing with skin cream sales, it is desired to estimate  $\beta_1$  and  $\beta_2$  jointly with a family confidence coefficient of 0.90.

$$s^2(B) = \text{MSE}(X^t X)^{-1} = \begin{pmatrix} 5.9021 & \lambda & + \\ + & .000036656 & + \\ \lambda & + & .000000937 \end{pmatrix}$$

$$s^2(b_1) = .000036656, \quad s(b_1) = .006054$$

$$s^2(b_2) = .000000937, \quad s(b_2) = .0009681$$

$$T = t\left(1 - \frac{0.10}{2 \times 2}, 12\right) = t(0.975, 12) = 2.179$$

So

$$0.4961 - (2.179)(.006054) \leq \beta_1 \leq 0.4961 + (2.179)(.006054)$$

or

$$0.483 \leq \beta_1 \leq 0.509$$

and similarly

$$0.0071 \leq \beta_2 \leq 0.0113$$