

CH5440: MULTIVARIATE DATA ANALYSIS ASSIGNMENT-1

① a) i) y known, b unknown

$$\text{objective} : \min_{\alpha, b, u_i} \sum_i \frac{(y_i - \alpha \hat{u}_i - b)^2}{\sigma_\varepsilon^2} + \frac{(u_i - \hat{u}_i)^2}{\sigma_b^2}$$

$$\begin{aligned} &\text{multiply by } \sigma_\varepsilon^2 \\ &\text{divide by } \sigma_\varepsilon^2 : \min_{\alpha, b, u_i} \frac{1}{\sigma_\varepsilon^2} \sum_i (y_i - \alpha \hat{u}_i - b)^2 \\ &\quad + \frac{\sigma_\varepsilon^2}{\sigma_b^2} (u_i - \hat{u}_i)^2 \end{aligned}$$

$$\begin{aligned} &\text{this is equivalent} \\ &\text{to} : \min_{\alpha, b, u_i} \sum_{i=1}^N (y_i - \alpha \hat{u}_i - b)^2 \\ &\quad + g(u_i - \hat{u}_i)^2 \\ &\quad - E \text{ (say)} \end{aligned}$$

$$\frac{\partial E}{\partial \alpha} = 0 \Rightarrow 2 \sum_{+i} \hat{u}_i (y_i - \alpha \hat{u}_i - b) = 0$$

$$\Rightarrow \sum_{i=1}^N \hat{u}_i (y_i - \alpha \hat{u}_i - b) = 0 \quad \text{--- (1)}$$

$$\frac{\partial E}{\partial b} = 0 \Rightarrow 2 \sum_{+i} (y_i - \alpha \hat{u}_i - b) = 0$$

$$\Rightarrow a \sum_{+i} \hat{u}_i + b \sum_{+i} 1 = \sum_{+i} y_i$$

$$\Rightarrow a \cdot \sum_{+i} \hat{u}_i + Nb = \sum_{+i} y_i \quad \text{--- (2)}$$

$$\frac{\partial E}{\partial \hat{u}_i} = 0 + i \Rightarrow \text{not soln.} \\ a(y_i - \alpha \hat{u}_i - b) + g(u_i - \hat{u}_i) = 0$$

$$\Rightarrow \boxed{\hat{u}_i = \frac{ay_i - ab + gu_i}{a^2 + g}} \quad (3)$$

Substitute (3) in (2),

$$Nb = \sum y_i - \alpha a \sum \left(\frac{ay_i - ab + gu_i}{a^2 + g} \right)$$

$$\Rightarrow N(a^2 + g)b = (a^2 + g) \sum y_i \\ + (-a^2 \sum y_i + a^2 b + \\ - agu_i)$$

$$\Rightarrow Nb = g \sum_{i=1} y_i - ga \sum_{i=1} u_i$$

$$\Rightarrow b = \frac{\sum y_i}{N} - \frac{ga \sum u_i}{N}$$

$$\Rightarrow \boxed{b = \bar{y} - a \bar{u}} \quad (4)$$

Substitute (3) in (1)

$$\sum \hat{u}_i (y_i - \alpha \left(\frac{ay_i - ab + gu_i}{a^2 + g} \right) - b) = 0$$

$$\Rightarrow \sum \hat{u}_i (ay_i + gy_i - ay_i + ab - agu_i - a^2 b \\ - bg) = 0$$

$$\Rightarrow \sum \hat{u}_i (gy_i - (ag u_i) - g \bar{y} - a^2 \bar{u}) = 0$$

(Substituting b value from (4))

$$\Rightarrow \sum_{+i} \hat{u}_i (y_i - \cancel{\alpha u_i})$$

cancelling $\cancel{\alpha}$,

$$\Rightarrow \sum_{+i} \hat{u}_i ((y_i - \bar{y}) - \alpha(u_i - \bar{u})) = 0$$

Substituting value of \hat{u}_i from ③,

$$\sum \left(\frac{\alpha y_i - \alpha \bar{y} + \beta u_i}{\alpha^2 + \beta} \right) ((y_i - \bar{y}) - \alpha(u_i - \bar{u})) = 0$$

Substituting b from ④

$$\Rightarrow \sum (\alpha y_i - \alpha(\bar{y} - \alpha \bar{u}) + \beta u_i) \left((y_i - \bar{y}) - \cancel{\alpha(u_i - \bar{u})} \right) = 0$$

$$\Rightarrow \sum (\alpha(y_i - \bar{y}) + \alpha^2 \bar{u} + \beta u_i) ((y_i - \bar{y}) - \cancel{\alpha(u_i - \bar{u})}) = 0$$

$$\begin{aligned} \Rightarrow \sum_{+i} & \alpha(y_i - \bar{y})^2 - \cancel{\alpha^2(y_i - \bar{y})(u_i - \bar{u})} \\ & + \cancel{\alpha^2 \bar{u}(y_i - \bar{y})} - \cancel{\alpha^3 \bar{u}(u_i - \bar{u})} \\ & + \beta u_i(y_i - \bar{y}) - \cancel{\beta \alpha u_i(u_i - \bar{u})} = 0 \end{aligned}$$

$$\begin{aligned} \alpha^3 \text{ term: } - \sum \bar{u}(u_i - \bar{u}) &= \bar{u} (\sum u_i - N\bar{u}) \cancel{(\sum 1)} \\ &= \bar{u} (\sum u_i - N\bar{u}) = 0 \end{aligned}$$

$$\begin{aligned} \alpha^2 \text{ term: } \sum \bar{u}(y_i - \bar{y}) - (y_i - \bar{y})(u_i - \bar{u}) &= \bar{u} (\sum y_i - N\bar{y}) - (\sum y_i u_i + N\bar{y}\bar{u} - \bar{y} \sum u_i \\ &\quad - \bar{u} \sum y_i) \end{aligned}$$

$$= - \left(\sum y_i u_i + N \bar{y} \bar{u} - N \bar{y} \bar{u} - \alpha \bar{y} \bar{u} \right)$$

$$= - \left(\sum y_i u_i - N \bar{y} \bar{u} \right) \quad (\because \sum_{i=1}^N \alpha_i = \bar{\alpha})$$

$$= - N S_{yu} \quad (\text{definition of } S_{yu}) \rightarrow E(YU)$$

a term: $\sum (y_i - \bar{y})^2 - g u_i (u_i - \bar{u}) - E(Y)E(u)$

$$= N S_{yy} - \left[g \sum_{i=1}^N (u_i^2) + g u_i - g \bar{u} \sum u_i \right]$$

$$= N S_{yy} - g \left[\sum (u_i^2) - N (\bar{u})^2 \right]$$

$$= N S_{yy} - g N S_{uu} = N (S_{yy} - g S_{uu})$$

constant : $\sum g u_i (y_i - \bar{y}) = g \left(\sum u_i y_i - \sum u_i \bar{y} \right)$

$$= N g \left(\frac{\sum (u_i y_i)}{N} - \bar{u} \bar{y} \right)$$

$$= N g S_{yu} \quad (\text{definition of } S_{yu})$$

$$\downarrow \\ E(YU) - E(Y)E(U)$$

So, ⑤ becomes:

$$- N S_{yu} a^2 + (N S_{yy} - g N S_{uu}) a + N g S_{yu} = 0$$

$$\Rightarrow S_{yu} a^2 + (g S_{uu} - S_{yy}) a + g S_{yu} = 0$$

$$\Rightarrow a = \frac{(S_{yy} - g S_{uu}) S_{yy} - g S_{yu} + \sqrt{g^2 S_{yu}^2}}{2 S_{yu}}$$

Applying Quadratic Formula,

$$a = \frac{sy_y - f s_{uu} \pm \sqrt{(sy_y - f s_{uu})^2 + 4f^2 s_y u}}{2 s_y u} \quad \text{--- (6)}$$

For minimum 2nd order condition: $\frac{d^2 E}{da^2} > 0$

$$\frac{d^2 E}{da^2} = 2 s_y u a + (f s_{uu} - sy_y)$$

Subst + root

$$\frac{2 s_y u}{2 s_y u} \left(sy_y - f s_{uu} \pm \sqrt{(sy_y - f s_{uu})^2 + 4f^2 s_y u} \right) + (f s_{uu} - sy_y)$$

$$= \sqrt{(sy_y - f s_{uu})^2 + 4f^2 s_y u} > 0 \quad \text{--- (7)}$$

other root will give $- \sqrt{(sy_y - f s_{uu})^2 + 4f^2 s_y u}$
that's not a minimum

$$\therefore \hat{a} = \frac{sy_y - f s_{uu} + \sqrt{(sy_y - f s_{uu})^2 + 4f^2 s_y u}}{2 s_y u} \quad \text{--- (8)}$$

ii) If b is known to be 0.

$$\text{objective func} \quad \min_{a_i, b_i, \hat{t}_i} \sum \frac{(y_i - a_i t_i)^2}{\sigma_e^2} + \frac{(u_i - \hat{t}_i)^2}{\sigma_g^2}$$

$$\frac{\partial E}{\partial \hat{u}_i} = 0 \Rightarrow -a(y_i - a\hat{u}_i) + g(u_i - \hat{u}_i) = 0$$

$\Rightarrow \boxed{\hat{u}_i = \frac{ay_i + g u_i}{a^2 + g}} \quad \textcircled{9}$

$$\frac{\partial E}{\partial a} = 0 \Rightarrow \sum \hat{u}_i (a y_i - a \hat{u}_i + y_i) = 0 \quad \textcircled{10}$$

Subst. ⑨ in ⑩,

$$\sum \hat{u}_i \left(-a \left(\frac{ay_i + gu_i}{a^2 + g} \right) + y_i \right) = 0$$

$$\Rightarrow \sum \hat{u}_i (g)(y_i - ay_i) = 0$$

Subst. from ⑨ again,

$$\sum \frac{ay_i + gu_i}{a^2 + g} (y_i - ay_i) = 0$$

$$\Rightarrow \sum ay_i^2 - ay_i u_i + gy_i u_i - a^2 y_i^2 = 0$$

$$\Rightarrow a^2(\sum y_i u_i) + a(g \sum u_i^2 - \sum y_i^2) - gy_i u_i \sum y_i u_i = 0$$

These coeff can be termed as respective variances around zero.

Solving for a & ensuring $\frac{\partial E}{\partial a} > 0$ as in previous case,

$$\hat{a} = \frac{\left(\sum y_i^2 - g \sum x_i^2 \right) + \sqrt{\left(\sum y_i^2 - a \sum x_i^2 \right)^2 + 4g \left(\sum x_i y_i \right)^2}}{2 \sum x_i y_i}$$

$$b) \hat{a} = \frac{S_{yy} - S_{\text{sum}} + \sqrt{(S_{yy} - S_{\text{sum}})^2 + 4S^2 y_u}}{2S_{yu}}$$

$$\text{let } \hat{a} = \frac{S_{yy} + \sqrt{(S_{yy} - 0)^2 + 0}}{2S_{yu}}$$

$$S \rightarrow 0$$

$$\Rightarrow \hat{a}_{\text{IOLS}} = \frac{S_{yy}}{S_{yu}}$$

$$\hat{b}_{\text{IOLS}} = \bar{y} - \hat{a}_{\text{IOLS}} \bar{u} = \bar{y} - \hat{a}_{\text{IOLS}} \bar{u}$$

$$\hat{u}_{\text{IOLS}} = \frac{\hat{a}_i y_i - \hat{a} \hat{b}_{\text{IOLS}}}{\hat{a}^2} = \frac{y_i - \hat{b}_{\text{IOLS}}}{\hat{a}_{\text{IOLS}}}$$

$$\hat{y}_{i, \text{IOLS}} = y_i \quad (\text{Subst } f=0 \text{ in (3)})$$

$$\begin{aligned} \hat{a}_{\text{OLS}} &= \frac{S_{yy} - S_{\text{sum}} + \sqrt{(S_{yy} - S_{\text{sum}})^2 + 4S^2 y_u}}{2S_{yu}} \\ &\text{let } S \rightarrow \infty \end{aligned}$$

$$\begin{aligned} &= \frac{\hat{a} + \frac{S_{yy} - S_{\text{sum}} + \sqrt{(S_{yy} - S_{\text{sum}})^2 + 4S^2 y_u}}{2S_{yu}}}{S} \\ &\quad S \rightarrow \infty \end{aligned}$$

$$\text{let } \xi_g = \frac{1}{S}$$

$$\Rightarrow \hat{a}_{\text{OLS}} = \frac{\hat{a} + \frac{\xi_g S_{yy} - S_{\text{sum}} + \sqrt{(\xi_g S_{yy} - S_{\text{sum}})^2 + 4\xi_g^2 y_u}}{2\xi_g S_{yu}}}{\xi_g} \quad \xi_g \rightarrow 0$$

* OLS:

Recall the quadratic eqn derived in (a),

$$a^2 \text{Sym} + (\text{Sum} - \text{Syy})a - \text{Syy} = 0$$

Divide by $\frac{1}{8}$,

$$a^2 \left(\frac{\text{Sym}}{\frac{1}{8}} \right) + \left(\text{Sum} - \frac{\text{Syy}}{\frac{1}{8}} \right) a - \text{Syy} = 0$$

Let $\xi = \frac{1}{8}$, $a^2 (\xi \text{Sym})$

$$a^2 (\xi \text{Sym}) + (\text{Sum} - \xi \text{Syy}) a - \text{Syy} = 0$$

$$\begin{aligned} \lim_{\xi \rightarrow 0} & \Rightarrow \lim_{\xi \rightarrow 0} \Rightarrow \text{Let } (\xi \text{Sym}) a^2 \\ & \xi \rightarrow 0 + (\text{Sum} - \xi \text{Syy}) a \\ & - \text{Syy} = 0 \\ \Rightarrow a(\text{Sum}) - \text{Syy} & = 0 \end{aligned}$$

$$\Rightarrow \hat{a}_{OLS} = \frac{\text{Sym}}{\text{Sum}}$$

$$\hat{b}_{OLS} = \bar{y} - \hat{a}_{OLS} \bar{x}$$

$$\hat{u}_{i, OLS} = u_i \Rightarrow \hat{u}_{i, OLS} = \hat{u}_i$$

$$\begin{aligned} \hat{u}_{i, OLS} &= u_i \Rightarrow \hat{u}_{i, OLS} = \hat{a}_{OLS} \hat{x}_i + \hat{b}_{OLS} \\ \hat{y}_{i, OLS} &= \end{aligned}$$

WTLS Solutions (derived in (a)) :

$$\hat{a}_{WTLS} = \frac{S_{yy} - g S_{uu} + \sqrt{(S_{yy} - g S_{uu})^2 + 4g^2 S_{uu}}}{2S_{yuu}}$$

$$\hat{b}_{WTLS} = \bar{y} - \hat{a}_{WTLS} \bar{u}$$

$$\hat{a}_{WTLS} = \hat{a}$$

$$\hat{u}_{i, WTLS} = \alpha \cdot \text{df} \cdot \frac{\left(\hat{a}_{WTLS} (y_i - \hat{b}_{WTLS}) + g u_i \right)}{(\alpha^2 + g)}$$

$$\hat{y}_{i, WTLS} = \hat{a}_{WTLS} \hat{u}_{WTLS} + \hat{b}_{WTLS}$$

- ② a) Here, I consider CO₂ concentration as input variable and temperature deviation as output variable

This is because CO₂ concentration measurements are likely to have more errors. And, from an engineering viewpoint, CO₂ emission causes temperature deviation

N = 31 samples.

Perform Computing sample statistics using MATLAB we get,

$$\bar{y} = \frac{\sum y_i}{N} = 0.8512$$

$$\bar{u} = \frac{\sum u_i}{N} = 369.4197$$

$$s_{yy} = \frac{1}{N} \sum (y_i - \bar{y})^2 = 0.0875$$

$$s_{yu} = \frac{1}{N} \sum (y_i - \bar{y})(u_i - \bar{u}) = 4.192$$

$$s_{uu} = \frac{1}{N} \sum (u_i - \bar{u})^2 = 255.74$$

$$\hat{\alpha}_{OLS} = \frac{s_{yu}}{s_{uu}} = 0.01639$$

$$\hat{\beta}_{OLS} = \bar{y} - \hat{\alpha} \bar{u} = -5.2044$$

$$\text{OLS estimate of CO}_2 \text{ concentration} = \frac{3.6 - \hat{\beta}_{OLS}}{\hat{\alpha}_{OLS}}$$

regd. for 3.6°F rise

$$= 537.01675 \text{ ppm}$$

$$\hat{\alpha}_{TLS} = \frac{s_{yy} - s_{uu} + \sqrt{(s_{yy} - s_{uu})^2 + 4s_{yu}^2}}{2s_{yu}} = 0.016393$$

$$\hat{\beta}_{TLS} = \bar{y} - \hat{\alpha}_{TLS} \bar{u} = -5.20482$$

$$\text{TLS estimate of CO}_2 \text{ concentration} = \frac{3.6 - \hat{\beta}_{TLS}}{\hat{\alpha}_{TLS}}$$

regd. for 3.6°F rise

$$= 537.0952 \text{ ppm}$$

b) Now, we keep time^(year) as input variable u
and CO₂ as output variable y.

We know that there is no error in the input, i.e.,
year. So we choose to go with an OLS
estimator

Computing Sample Statistics in MATLAB,

$$\bar{y} = 369.4197 \text{ ppm} \quad \bar{u} = 1999$$

$$s_{yy} = 285.7407 \quad s_{uu} = 142.5342 \quad s_{uy} = 80$$

$$\hat{\alpha}_{OLS} = \frac{s_{uy}}{s_{uu}} = \underline{1.7817}$$

$$\hat{\beta}_{OLS} = \bar{y} - \hat{\alpha}\bar{u} = \underline{-3192.153}$$

i) Finding time taken to reach the OLS estimate
of concentration

$$t = \frac{537.105 - \hat{\beta}}{\hat{\alpha}} = \boxed{2093.1179}$$

ii) Finding time taken to reach the LS estimate of
concentration

$$t = \frac{537.0952 - \hat{\beta}}{\hat{\alpha}} = \boxed{2093.111}$$

Both the estimates are similar. Conservatively, we
can say by 2093 year $\frac{1}{2}$ (72 more years) we will
reach the minimum level
of CO₂ to
cause 3.6 °F rise

③ a) Since we are performing OLS as well as Regs:
 choice of input-output doesn't matter for this analysis

Set input, u as EP method
 output, y as CP method.

Sample statistics. [All computations done in MATLAB]

$$Syu = 0.9232 \quad Syu = 0.9241 \quad \text{Sum} = 0.9311$$

$$\bar{y} = 1.9505 \quad \bar{u} = 2.0155$$

Our aim to see whether the 95% confidence interval of $\hat{\alpha}$ contains 1 (so that $y = u$ in this model comes)

$$95\% \text{ CI: } \hat{\alpha} \pm 2.16 \sigma_{\hat{\alpha}}$$

$$\therefore \sigma_{\hat{\alpha}}^2 \text{ is unknown, } \sigma_{\hat{\alpha}}^2 = \frac{1}{N-2} \sum (y_i - \hat{\alpha} u_i - \hat{\beta})^2$$

OLS Assumption: EP method is free of error, CP method has error

$$\hat{\alpha}_{OLS} = \frac{Syu}{Sum} = 0.9924$$

$$\hat{\beta}_{OLS} = -0.0497$$

$$95\% \text{ confident interval of } \hat{\alpha} : [0.9776, 1.0073]$$

C-I contains 1! $\hat{\beta}$ close to 0!

DOLS

Assumption : EP has error, CF method has no error.

$$\hat{\alpha}_{DOLS}^1 = \frac{Syy}{yu} = \boxed{0.9991}$$

$$\hat{\beta}_{DOLS}^1 = \bar{y} - \hat{\alpha}_{DOLS}^1 \bar{u} = \boxed{-0.0564}$$

95% confidence interval for $\hat{\alpha}_{DOLS}^1$: $[0.9809, 1.0074]$

C.I. contains 1! $\hat{\beta}$ close to 0!

for $\hat{\alpha}$

TLS

Assumption: both EP & CF method has error.

$$\hat{\alpha}_{TLS}^1 = \frac{Syy - Suu + \sqrt{(Syy - Suu)^2 + 4S^2yu}}{2Syy} = \boxed{0.9957} \\ - \boxed{0.016893}$$

$$\hat{\beta}_{TLS}^1 = \frac{\alpha}{y} - \hat{\alpha}_{TLS}^1 \bar{u} = \boxed{-0.0564}$$

95% confidence interval for $\hat{\alpha}_{TLS}^1 = [0.9809, 1.00106]$

C.I. contains $\hat{\alpha} = 1$! $\hat{\beta}$ close to 0

All 3 approaches have $\hat{\alpha}$ close to 1 and $\hat{\beta}$ close to 0

So we can conclude that the new method (CF) is
a good sg substitute for the established method
(EP)

b) OLS : Input is error free \rightarrow EP meant
as error free

\Rightarrow truth = u_i =
 \Rightarrow estimate = y_i = 2.31 mg/L

EOLS
Output is error free \rightarrow CF is error free

\Rightarrow Estimate = y_i = 2.20 mg/L

TLS Approximating $\alpha = 1$, $\beta = 0$,

Assuming $\alpha = 1$,

estimate $\hat{y}_i = \hat{u}_i = \frac{\alpha(y_i - \beta)}{\alpha^2 + 1} + u_i$

$$= \frac{y_i + u_i}{2} = \frac{2.31 + 2.2}{2}$$

\Rightarrow estimate of phytic acid = 2.255 mg/L

④ The objective is to find if the defect is aligned across vertical or horizontal axis.

So by performing linear regression,

- If we get equation as $y = \beta$ it is parallel to x-axis
- if we get equation as $x = \beta$ it is parallel to y-axis

Essentially we can do 2 regression for each data:

$$y = \alpha_1 x + \beta_1 \epsilon \quad x = \alpha_2 y + \beta_2$$

and check if α_1 or α_2 is small enough / close to zero

Since swapping x & y is equal to does we need not do that separately. Just

need to do OLS and to cover TLS also.

In addition, we can check using TLS. Also, value of β doesn't matter in terms of deciding defect orientation

Defect - ①

for $y = \alpha_1 x + \beta_1$

$$\bar{y} = 824.044 \quad \bar{x} = 3774.6$$

$$s_{yy} = 267.0202$$

$$s_{xy} = 171.9279 \quad \text{Sum} \\ = 1276.5$$

$$\text{Model 1: } y = \alpha_1 x + \beta_1$$

$$\alpha_{1, \text{OLS}} = \frac{s_{yy}}{s_{xx}} = \frac{0.0092}{s_{yy}} \quad \text{P}_{1, \text{O.S.}}$$

$$\alpha_{1, \text{TLS}} = \frac{s_{yy} - s_{xy} + \sqrt{(s_{yy} - s_{xy})^2 + 4s_{xy}^2}}{2s_{xy}} \\ = \underline{0.0093}$$

Both OLS & TLS suggest, the line is of the form

$$y = \underline{\beta}$$

$$\text{Model 2: } \underline{y - x = \alpha_2 y + \beta_2}$$

~~Note that actual slope is inverse of this~~

So we have

successfully identified the defect α along x -axis

So \therefore Defect 1 is due to corrosion of steel reinforcement bars

Defect 2

$$2\bar{x} = 2143.9 \quad \bar{y} = 3190.2$$

$$s_{yy} = 1904.0 \quad s_{yy} = 1708.0$$

$$s_{xy} = 2153.9 \quad ; \text{ model ①: } y = \hat{\alpha}x + \hat{\beta}$$

$$\hat{\alpha}_{1,OLS} = \frac{S_{yy}}{S_{xx}} = 0.8092 \quad 0.793$$

$$\hat{\alpha}_{1,TLS} = \frac{S_{yy} - S_{xx} + \sqrt{(S_{yy}-S_{xx})^2 + 4S_{xy}^2}}{2S_{xy}} = 0.9295$$

Both $\hat{\alpha}$ estimates are close to 1. Trying with

model ② -

$$\text{Model ② : } x = \hat{\alpha} y + \hat{p}$$

Model ② :

$$\hat{\alpha}_{2,OLS} = \frac{S_{yy}}{S_{xx}} = 0.8971$$

(using my nomenclature as defined earlier)

$$\hat{\alpha}_{2,TLS} = \frac{S_{yy} - S_{xx} + \sqrt{(S_{yy}-S_{xx})^2 + 4S_{xy}^2}}{2S_{xy}} = 1.0758$$

Once again both are close to 1.

The equation is of the form $y = x$ and not
along vertical or horizontal axis
(instead oriented 45° along $y=x$)

\Rightarrow Defect 2 is not due to corrosion of steel reinforcement bars

Defect 2

$$\bar{y} = 3522.3 \quad \bar{x} = 2832.7 \quad s_{yy} = 3099.5$$

$$s_{yx} = 910.12 \quad s_{xx} = 5152.5$$

$$\text{Model ①: } y = \hat{\alpha} x + \hat{\beta}$$

$$\hat{\alpha}_{\text{OLS}} = 28.4298 \quad \hat{\alpha}_{\text{TLS}} = 28.4298$$

$$\text{Model ②: } x = \hat{\alpha} y + \hat{\beta}$$

$$\hat{\alpha}_{\text{OLS}} = 0.0294 \quad \hat{\alpha}_{\text{TLS}} = 0.0352$$

We see that $\hat{\alpha}_{\text{OLS}}$ & $\hat{\alpha}_{\text{TLS}}$ are close to zero.

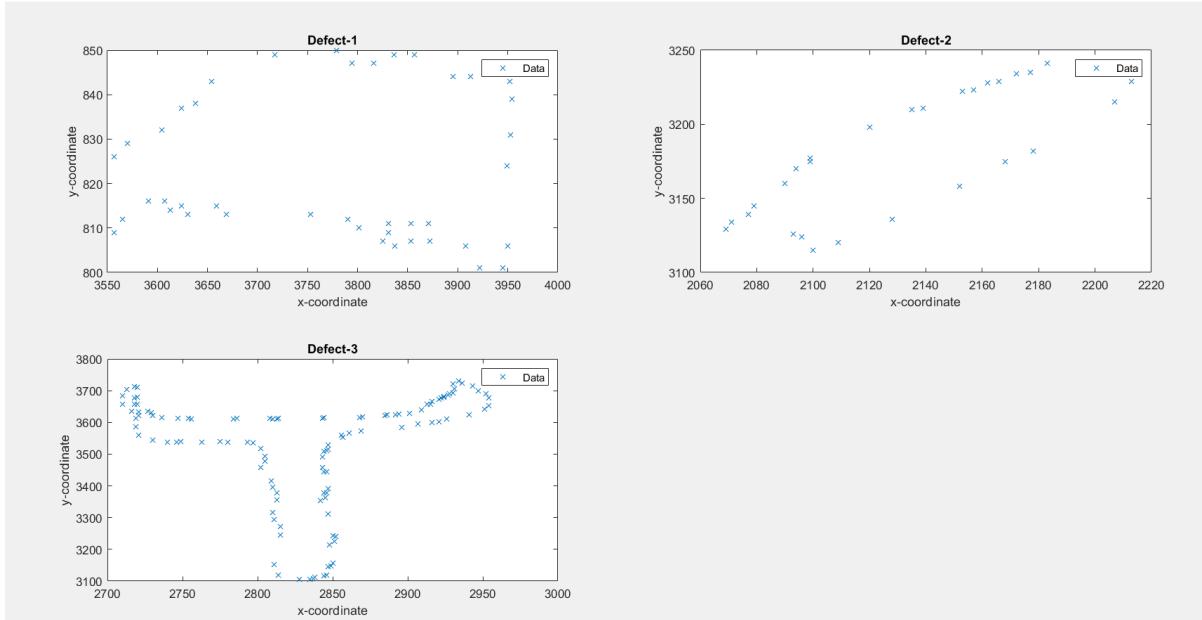
\Rightarrow the line is of $x = \hat{\beta}$. The defect is along the dy -axis

- Defect 3 is due to corrosion of steel reinforcement bars.

Conclusion: Defects 1 and 2 are due to corrosion of steel reinforcement bars

Plots

Question-4 plots



The above plots agree with the inferences concluded using OLS/TLS.

Defect-1: along horizontal axis

Defect-2: along $y = x$

Defect-3: along vertical axis

Code

OLS.m (function to perform OLS)

```
function [alpha, beta, uhat, yhat, s] = OLS(u,y)
N = length(u);
ybar = mean(y);
ubar = mean(u);
syy = var(y,1);
suu = var(u,1);
syu = 1/N*sum((y-ybar).* (u-ubar));
uhat = u;
alpha = syu/suu;
beta = ybar - alpha*ubar;
yhat = alpha*uhat + beta;
s =
struct('alpha',alpha,'beta',beta,'ybar',ybar,'ubar',ubar,
'syy',syy,'syu',syu,'suu',suu);
end
```

IOLS.m (function to perform IOLS)

```
function [alpha, beta, uhat, yhat, s] = IOLS(u,y)
    N = length(u);
    ybar = mean(y);
    ubar = mean(u);
    syy = var(y,1);
    suu = var(u,1);
    syu = 1/N*sum((y-ybar).* (u-ubar));
    yhat = y;
    alpha = syy/syu;
    beta = ybar - alpha*ubar;
    uhat = (yhat - beta)/alpha;
    s =
struct('alpha',alpha,'beta',beta,'ybar',ybar,'ubar',ubar,
'syy',syy,'syu',syu,'suu',suu);
end
```

TLS.m (function to perform TLS)

```
function [alpha, beta, uhat, yhat, s] = TLS(u,y)
    N = length(u);
    ybar = mean(y);
    ubar = mean(u);
    syy = var(y,1);
    suu = var(u,1);
    syu = 1/N*sum((y-ybar).* (u-ubar));
    alpha = (syy - suu + sqrt((syy-suu)^2 +
4*syu^2))/2/syu;
    beta = ybar - alpha*ubar;
    uhat = (alpha*(y-beta)+u)/(alpha^2+1);
    yhat = alpha*uhat + beta;
    s =
struct('alpha',alpha,'beta',beta,'ybar',ybar,'ubar',ubar,
'syy',syy,'syu',syu,'suu',suu);
end
```

Question-2

```
clear; close all;
%% Open data
A = readmatrix('CO2.csv');
time = A(:,1);
CO2 = A(:,2);
temp = A(:,3);
temp_cut = 3.6;
%% Predict maximum permissible level of CO2 - OLS
% y -> temperature deviation
% u -> CO2
```

```
u = CO2;
y = temp;
[alpha, beta, uhat, yhat, s] = OLS(u,y);
disp(s);
CO2_cut_OLS = (temp_cut-beta)/alpha;
%% Predict maximum permissible level of CO2 - TLS
[alpha_TLS, beta_TLS, uhat_TLS, yhat_TLS,s_TLS] =
TLS(u,y);
CO2_cut_TLS = (temp_cut-beta_TLS)/alpha_TLS;
%% Predict year
ut = time;
yt = CO2;
[alpha_t, beta_t, uhat_t, yhat_t, s_t] = OLS(ut,yt);
t_OLS = (CO2_cut_OLS-beta_t)/alpha_t;
tpred_OLS = ceil(t_OLS);
t_TLS = (CO2_cut_TLS-beta_t)/alpha_t;
tpred_TLS = ceil(t_TLS);
```

Question-3

```
close all; clear;
%% Data
EP =
[1.98,2.31,3.29,3.56,1.23,1.57,2.05,0.66,0.31,2.82,0.13,3
.15,2.72,2.31,1.92,1.56,0.94,2.27,3.17,2.36]';
CF =
[1.87,2.2,3.15,3.42,1.1,1.41,1.84,0.68,0.27,2.8,0.14,3.2,
2.7,2.43,1.78,1.53,0.84,2.21,3.10,2.34]';
% u = EP, y = CF
u = EP;
y = CF;
N = length(u);
%% Part a)
%% OLS
[alpha_OLS, beta_OLS, uhat_OLS, yhat_OLS,s_OLS] =
OLS(u,y);
sigma_e_OLS = 1/(N-2)*sum((y-alpha_OLS*u-beta_OLS).^2);
CI_OLS = [alpha_OLS-
2.16*sigma_e_OLS,alpha_OLS+2.16*sigma_e_OLS];
%% IOLS
[alpha_IOLS, beta_IOLS, uhat_IOLS, yhat_IOLS,s_IOLS] =
IOLS(u,y);
sigma_e_IOLS = 1/(N-2)*sum((y-alpha_IOLS*u-
beta_IOLS).^2);
CI_IOLS = [alpha_IOLS-
2.16*sigma_e_IOLS,alpha_IOLS+2.16*sigma_e_IOLS];
%% TLS
```

```
[alpha_TLS, beta_TLS, uhat_TLS, yhat_TLS, s_TLS] =
TLS(u, y);
sigma_e_TLS = 1/(N-2)*sum((y-alpha_TLS*u-beta_TLS).^2);
CI_TLS = [alpha_TLS-
2.16*sigma_e_TLS, alpha_TLS+2.16*sigma_e_TLS];
%% Part b)
ui = 2.31;
yi = 2.20;
% OLS: u is perfect
OLS_pred = ui;
% IOLS: y is perfect
IOLS_pred = yi;
% TLS: both imperfect. doing a perpendicular projection
will give us an
% estimate for EP and CF. Since we need a single estimate
we will assume
% alpha = 1 and beta = 0. So yhat = uhat = (u+y)/2
TLS_pred = (ui+yi)/2;
```

Question-4

```
clear; close all;
%% Open data
A = readmatrix('defects_annotation_data.csv');
x1 = rem_NaN(A(:,1)); y1 = rem_NaN(A(:,2));
x2 = rem_NaN(A(:,4)); y2 = rem_NaN(A(:,5));
x3 = rem_NaN(A(:,7)); y3 = rem_NaN(A(:,8));
%% Defect-1
N = length(x1);
% TLS
[alpha_TLS, beta_TLS, uhat_TLS, yhat_TLS, s_TLS] =
TLS(x1, y1);
sigma_e_TLS = 1/(N-2)*sum((y1-alpha_TLS*x1-beta_TLS).^2);
CI_TLS = [alpha_TLS-
2.16*sigma_e_TLS, alpha_TLS+2.16*sigma_e_TLS];
% TLS - inverted
[alpha_TLS2, beta_TLS2, uhat_TLS2, yhat_TLS2, s_TLS2] =
TLS(y1, x1);
sigma_e_TLS2 = 1/(N-2)*sum((y1-alpha_TLS2*x1-
beta_TLS2).^2);
CI_TLS2 = [alpha_TLS2-
2.16*sigma_e_TLS2, alpha_TLS2+2.16*sigma_e_TLS2];
% OLS
[alpha_OLS, beta_OLS, uhat_OLS, yhat_OLS, s_OLS] =
OLS(x1, y1);
sigma_e_OLS = 1/(N-2)*sum((y1-alpha_OLS*x1-beta_OLS).^2);
```

```
CI_OLS = [alpha_OLS-
2.16*sigma_e_OLS, alpha_OLS+2.16*sigma_e_OLS];
% OLS - inverted
[alpha_OLS2, beta_OLS2, uhat_OLS2, yhat_OLS2, s_OLS2] =
OLS(y1,x1);
sigma_e_OLS2 = 1/(N-2)*sum((y1-alpha_OLS2*x1-
beta_OLS2).^2);
CI_OLS2 = [alpha_OLS2-
2.16*sigma_e_OLS2, alpha_OLS2+2.16*sigma_e_OLS2];
% Plot
subplot(2,2,1);
plot(x1,y1,'x');
title('Defect-1'); xlabel('x-coordinate'); ylabel('y-
coordinate');
legend('Data');
%% Defect-2
% TLS
[alpha_TLS_def2, beta_TLS_def2, uhat_TLS_def2,
yhat_TLS_def2, s_TLS_def2] = TLS(x2,y2);
sigma_e_TLS_def2 = 1/(N-2)*sum((y2-alpha_TLS_def2*x2-
beta_TLS_def2).^2);
CI_TLS_def2 = [alpha_TLS_def2-
2.16*sigma_e_TLS_def2, alpha_TLS_def2+2.16*sigma_e_TLS_def
2];
% TLS - inverted
[alpha_TLS2_def2, beta_TLS2_def2, uhat_TLS2_def2,
yhat_TLS2_def2, s_TLS2_def2] = TLS(y2,x2);
sigma_e_TLS2_def2 = 1/(N-2)*sum((y2-alpha_TLS2_def2*x2-
beta_TLS2_def2).^2);
CI_TLS2_def2 = [alpha_TLS2_def2-
2.16*sigma_e_TLS2_def2, alpha_TLS2_def2+2.16*sigma_e_TLS2_
def2];
% OLS
[alpha_OLS_def2, beta_OLS_def2, uhat_OLS_def2,
yhat_OLS_def2, s_OLS_def2] = OLS(x2,y2);
sigma_e_OLS_def2 = 1/(N-2)*sum((y2-alpha_OLS_def2*x2-
beta_OLS_def2).^2);
CI_OLS_def2 = [alpha_OLS_def2-
2.16*sigma_e_OLS_def2, alpha_OLS_def2+2.16*sigma_e_OLS_def
2];
% OLS - inverted
[alpha_OLS2_def2, beta_OLS2_def2, uhat_OLS2_def2,
yhat_OLS2_def2, s_OLS2_def2] = OLS(y2,x2);
sigma_e_OLS2_def2 = 1/(N-2)*sum((y2-alpha_OLS2_def2*x2-
beta_OLS2_def2).^2);
```

```
CI_OLS2_def2 = [alpha_OLS2_def2-
2.16*sigma_e_OLS2_def2,alpha_OLS2_def2+2.16*sigma_e_OLS2_
def2];
% Plot
subplot(2,2,2);
plot(x2,y2,'x');
title('Defect-2'); xlabel('x-coordinate'); ylabel('y-
coordinate');
legend('Data');
%% Defect-3
% TLS
[alpha_TLS_def3, beta_TLS_def3, uhat_TLS_def3,
yhat_TLS_def3,s_TLS_def3] = TLS(x3,y3);
sigma_e_TLS_def3 = 1/(N-2)*sum((y3-alpha_TLS_def3*x3-
beta_TLS_def3).^2);
CI_TLS_def3 = [alpha_TLS_def3-
2.16*sigma_e_TLS_def3,alpha_TLS_def3+2.16*sigma_e_TLS_def
3];
% TLS - inverted
[alpha_TLS2_def3, beta_TLS2_def3, uhat_TLS2_def3,
yhat_TLS2_def3,s_TLS2_def3] = TLS(y3,x3);
sigma_e_TLS2_def3 = 1/(N-2)*sum((y3-alpha_TLS2_def3*x3-
beta_TLS2_def3).^2);
CI_TLS2_def3 = [alpha_TLS2_def3-
2.16*sigma_e_TLS2_def3,alpha_TLS2_def3+2.16*sigma_e_TLS2_
def3];
% OLS
[alpha_OLS_def3, beta_OLS_def3, uhat_OLS_def3,
yhat_OLS_def3,s_OLS_def3] = OLS(x3,y3);
sigma_e_OLS_def3 = 1/(N-2)*sum((y3-alpha_OLS_def3*x3-
beta_OLS_def3).^2);
CI_OLS_def3 = [alpha_OLS_def3-
2.16*sigma_e_OLS_def3,alpha_OLS_def3+2.16*sigma_e_OLS_def
3];
% OLS - inverted
[alpha_OLS2_def3, beta_OLS2_def3, uhat_OLS2_def3,
yhat_OLS2_def3,s_OLS2_def3] = OLS(y3,x3);
sigma_e_OLS2_def3 = 1/(N-2)*sum((y3-alpha_OLS2_def3*x2-
beta_OLS2_def3).^2);
CI_OLS2_def3 = [alpha_OLS2_def3-
2.16*sigma_e_OLS2_def3,alpha_OLS2_def3+2.16*sigma_e_OLS2_
def3];
% Plot
subplot(2,2,3);
plot(x3,y3,'x');
title('Defect-3'); xlabel('x-coordinate'); ylabel('y-
coordinate');
```

```
legend('Data');  
%% function to remove NaN values  
function vnew = rem_NaN(v)  
    vnew = v(~isnan(v));  
end
```