

Assignment #2 (100 points): Due 11:59 pm, Sunday, March 27, 2022

1. In the false data injection (FDI) attack, assume that the normal measurement is $\mathbf{z} = \mathbf{H}\mathbf{x} + \mathbf{n}$ and the attacked measurement is $\mathbf{z}_a = \mathbf{z} + \mathbf{a}$. Assume that the weighted least square (WLS) solution is given by $\hat{\mathbf{x}} = (\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{z}$ for the normal measurement \mathbf{z} , and let $\hat{\mathbf{x}}_a$ be the WLS solution obtained from the attacked measurement \mathbf{z}_a . Use your linear algebra knowledge to prove that the pre-attack residual $\mathbf{r} = \mathbf{z} - \mathbf{H}\hat{\mathbf{x}}$ is the same as the post-attack residual $\mathbf{r}_a = \mathbf{z}_a - \mathbf{H}\hat{\mathbf{x}}_a$ if $\mathbf{a} = \mathbf{H}\mathbf{c}$. (20 points)

Solution:

Given,

Pre-attack residual $\mathbf{r} = \mathbf{z} - \mathbf{H}\hat{\mathbf{x}}$

Attacked measurement is $\mathbf{z}_a = \mathbf{z} + \mathbf{a}$

Normal measurement is $\mathbf{z} = \mathbf{H}\mathbf{x} + \mathbf{n}$

and the Post-attack residual $\mathbf{r}_a = \mathbf{z}_a - \mathbf{H}\hat{\mathbf{x}}_a$.

$$\begin{aligned}
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} [(\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{z}_a] \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} [(\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} (\mathbf{z} + \mathbf{a})] \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} [(\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{z} + (\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{a}] \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} [\hat{\mathbf{x}} + (\mathbf{H}^T \mathbf{R}^{-1} \mathbf{H})^{-1} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{a}] \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} (\hat{\mathbf{x}} + \mathbf{H}^{-1} \mathbf{a}) \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} \hat{\mathbf{x}} - \mathbf{H} \mathbf{H}^{-1} \mathbf{a} \\
 &= \mathbf{z} + \mathbf{a} - \mathbf{H} \hat{\mathbf{x}} - \mathbf{a} \\
 &= \mathbf{z} - \mathbf{H} \hat{\mathbf{x}}
 \end{aligned}$$

$$\text{Since } \mathbf{a} = \mathbf{H}\mathbf{c} \Rightarrow \mathbf{z} - \mathbf{H}\hat{\mathbf{x}} + \mathbf{a} - \mathbf{a}$$

$$\Rightarrow \mathbf{z} - \mathbf{H}\hat{\mathbf{x}}$$

Which implies $\mathbf{r} = \mathbf{r}_a$

Hence it is proved that the pre-attack residual $\mathbf{r} = \mathbf{z} - \mathbf{H}\hat{\mathbf{x}}$ is the same as the post-attack residual $\mathbf{r}_a = \mathbf{z}_a - \mathbf{H}\hat{\mathbf{x}}_a$, if $\mathbf{a} = \mathbf{H}\mathbf{c}$

2. A one-line diagram of a 3-bus power system is shown below. Assume that we have one meter installed on each line to measure the power flows. For bus i and j connected by a line, the power flow P_{ij} , the reactance X_{ij} , and voltage angles θ_i, θ_j are considered to approximately follow the relation below when no noise is considered:

$$P_{ij} = (\theta_i - \theta_j)/X_{ij}$$

The line reactance has been measured in advance as:

$$X_{12} = 0.50 \text{ p.u.}$$

$$X_{13} = 0.10 \text{ p.u.}$$

$$X_{32} = 0.25 \text{ p.u.}$$

The meter readings have been reported as:

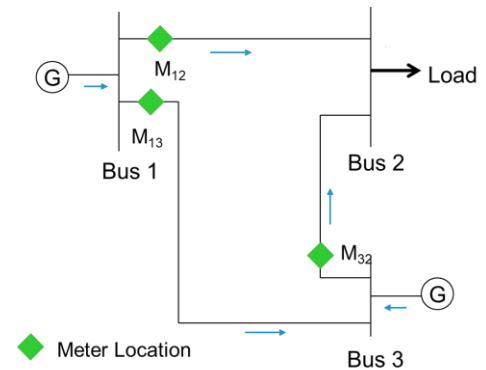
$$M_{12} = 0.60 \text{ p.u.}$$

$$M_{13} = 0.05 \text{ p.u.}$$

$$M_{32} = 0.40 \text{ p.u.}$$

Solve the questions below. (80 points)

- a) Assume that $\theta_1 = 0$ is the reference angle. Using different pairs of meter readings, find the values below:
- θ_2, θ_3 and P_{12} , using only the values of M_{13}, M_{32} and the reactance X .
 - θ_2, θ_3 and P_{13} , using only the values of M_{12}, M_{32} and the reactance X .
 - θ_2, θ_3 and P_{32} , using only the values of M_{12}, M_{13} and the reactance X .



Solution:

i.

Given,

Line reactance has been measured in advance are:

$$X_{12} = 0.50 \text{ p.u.}$$

$$X_{13} = 0.10 \text{ p.u.}$$

$$X_{32} = 0.25 \text{ p.u.}$$

Meter readings are:

$$M_{12} = 0.60 \text{ p.u.}$$

$$M_{13} = 0.05 \text{ p.u.}$$

$$M_{32} = 0.40 \text{ p.u.}$$

$$P_{ij} = (\theta_i - \theta_j)/X_{ij}$$

$$M_{12} = (\theta_1 - \theta_2)/X_{12}$$

$$0.60 = (0 - \theta_2)/0.50$$

$$\Rightarrow 0.300 = 0 - \theta_2$$

$$\Rightarrow \theta_2 = -0.3$$

$$\begin{aligned}
 M_{13} &= (\theta_1 - \theta_3) / X_{13} \\
 0.05 &= (0 - \theta_3) / 0.10 \\
 &\Rightarrow 0.005 = 0 - \theta_3 \\
 &\Rightarrow \theta_3 = -0.005
 \end{aligned}$$

Similarly,

$$\begin{aligned}
 P_{32} &= (\theta_3 - \theta_2) / X_{32} \\
 &\Rightarrow (-0.005 + 0.300) / 0.25 \\
 \mathbf{P_{32}} &\Rightarrow \mathbf{1.18 \text{ p.u.}}
 \end{aligned}$$

ii.

Given,

$$\begin{aligned}
 M_{12} &= (\theta_1 - \theta_2) / X_{12} \\
 0.60 &= (0 - \theta_2) / 0.50 \\
 &\Rightarrow 0.3 = 0 - \theta_2 \\
 &\Rightarrow \theta_2 = -0.3
 \end{aligned}$$

$$\begin{aligned}
 M_{32} &= (\theta_3 - \theta_2) / X_{32} \\
 0.40 &= (\theta_3 + 0.300) / 0.25 \\
 &\Rightarrow 0.10 = \theta_3 + 0.3 \\
 &\Rightarrow \theta_3 = -0.2
 \end{aligned}$$

Similarly,

$$\begin{aligned}
 P_{ij} &= (\theta_i - \theta_j) / X_{ij} \\
 P_{13} &= (\theta_1 - \theta_3) / X_{13} \\
 &\Rightarrow (0 + 0.2) / 0.1 \\
 \mathbf{P_{13}} &\Rightarrow \mathbf{2 \text{ p.u.}}
 \end{aligned}$$

iii.

Given,

$$\begin{aligned}
 M_{13} &= (\theta_1 - \theta_3) / X_{13} \\
 0.05 &= (0 - \theta_3) / 0.10 \\
 &\Rightarrow 0.005 = 0 - \theta_3 \\
 &\Rightarrow \theta_3 = -0.005
 \end{aligned}$$

$$\begin{aligned}
 M_{32} &= (\theta_3 - \theta_2) / X_{32} \\
 0.40 &= (-0.005 - \theta_2) / 0.25 \\
 &\Rightarrow 0.10 = -0.005 - \theta_2
 \end{aligned}$$

$$\Rightarrow \theta_2 = -0.105$$

Similarly,

$$\begin{aligned} P_{ij} &= (\theta_i - \theta_j)/X_{ij}, \\ P_{12} &= (\theta_1 - \theta_2)/X_{12} \\ &\Rightarrow (0 + 0.105)/0.50 \\ \mathbf{P}_{12} &\Rightarrow \mathbf{0.21 p.u.} \end{aligned}$$

- b) Let $\mathbf{z} = [M_{12}, M_{13}, M_{32}]^T$ be the measurement vector and $\mathbf{x} = [\theta_1, \theta_2, \theta_3]^T$ be the state variables, where T indicates the vector/matrix transpose. Then from the ideal case $\mathbf{z} = \mathbf{H}\mathbf{x}$ when there is no noise, we have:

$$\begin{bmatrix} M_{12} \\ M_{13} \\ M_{32} \end{bmatrix} = \begin{bmatrix} P_{12} \\ P_{13} \\ P_{32} \end{bmatrix} = \mathbf{H} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix}, \text{ where } \mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$

Based on the relation $P_{ij} = (\theta_i - \theta_j)/X_{ij}$, find the entries h_{ij} in the 3-by-3 matrix \mathbf{H} .

Solution:

Given,

line reactance has been measured are:

$$X_{12} = 0.50 \text{ p.u.}$$

$$X_{13} = 0.10 \text{ p.u.}$$

$$X_{32} = 0.25 \text{ p.u.}$$

$$\begin{bmatrix} P_{12} \\ P_{13} \\ P_{32} \end{bmatrix} = \mathbf{H} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix},$$

$$\text{And we have, } \mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$

$$\text{This implies, } \begin{bmatrix} P_{12} \\ P_{13} \\ P_{32} \end{bmatrix} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \theta_3 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} P_{12} \\ P_{13} \\ P_{32} \end{bmatrix} = \begin{bmatrix} h_{11}\theta_1 & h_{12}\theta_2 & h_{13}\theta_3 \\ h_{21}\theta_1 & h_{22}\theta_2 & h_{23}\theta_3 \\ h_{31}\theta_1 & h_{32}\theta_2 & h_{33}\theta_3 \end{bmatrix}$$

$$\Rightarrow P_{12} = h_{11}\theta_1 + h_{12}\theta_2 + h_{13}\theta_3$$

$$\Rightarrow P_{13} = h_{21}\theta_1 + h_{22}\theta_2 + h_{23}\theta_3$$

$$\Rightarrow P_{32} = h_{31}\theta_1 + h_{32}\theta_2 + h_{33}\theta_3$$

We know that,

$$P_{ij} = (\theta_i - \theta_j) / X_{ij}$$

$$P_{12} = (\theta_1 - \theta_2) / X_{12}$$

$$= (\theta_1 - \theta_2) / 0.50$$

$$= 2\theta_1 - 2\theta_2$$

$$P_{13} = (\theta_1 - \theta_3) / X_{13}$$

$$= (\theta_1 - \theta_3) / 0.10$$

$$= 10\theta_1 - 10\theta_3$$

$$P_{32} = (\theta_3 - \theta_2) / X_{32}$$

$$= (\theta_3 - \theta_2) / 0.25$$

$$= 4\theta_3 - 4\theta_2$$

$$\begin{matrix} P_{12} & 2 & -2 & 0 & \theta_1 \\ [P_{13}] & 10 & 0 & -10 & [\theta_2] \\ P_{32} & 0 & -4 & 4 & \theta_3 \end{matrix} \Rightarrow \mathbf{H} = \begin{bmatrix} 2 & -2 & 0 \\ 10 & 0 & -10 \\ 0 & -4 & 4 \end{bmatrix}$$

- c) As we use $\theta_1 = 0$ as the reference angle, we do not actually need to estimate θ_1 , and the state variables to be estimated are only $\mathbf{x}' = [\theta_2, \theta_3]^T$. For $\mathbf{z} = \mathbf{H}\mathbf{x}$, this suggests the column in \mathbf{H} corresponding to the coefficients of θ_1 should be removed. Find the new 3-by-2 matrix \mathbf{H}' with the column in \mathbf{H} corresponding to θ_1 is removed.

Solution:

Given,

$$\mathbf{z} = \mathbf{H}\mathbf{x}$$

When we use,

$$\theta_1 = 0 \text{ as the reference angle, the state variables to be estimated are only } \mathbf{x}' = [\theta_2, \theta_3]^T$$

So, by removing the column in \mathbf{H} corresponding to the coefficients of θ_1

$$\mathbf{z}' = \mathbf{H}' \mathbf{x}'$$

$$\mathbf{H} = \begin{bmatrix} 2 & -2 & 0 \\ 10 & 0 & -10 \\ 0 & -4 & 4 \end{bmatrix}$$

$$\mathbf{z}' = \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix} \begin{bmatrix} \theta_2 \\ \theta_3 \end{bmatrix}$$

$$\mathbf{H}' = \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix}$$

- d) Let $\mathbf{R} = [\sigma_{ij}^2]$ be the 3-by-3 covariance matrix of the additive white Gaussian noise \mathbf{n} , where σ_{ii}^2 is the variance of noise of meter i and σ_{ij}^2 ($i \neq j$) is the covariance between noises on two different meters i and j . Normally we assume the noises on different meters are independent, so the covariance $\sigma_{ij}^2 = 0$ for $i \neq j$; we also know the variances of noise on meters is: $\sigma_{11}^2 = 0.05$ (for M_{12}), $\sigma_{22}^2 = 0.045$ (for M_{13}) and $\sigma_{33}^2 = 0.05$ (for M_{32}) Find the covariance matrix \mathbf{R} by specifying all its entries σ_{ij}^2 in the matrix form.

Solution:

Given,

$$\sigma_{11}^2 = 0.05$$

$$\sigma_{22}^2 = 0.045$$

$$\sigma_{33}^2 = 0.05$$

According to the above, covariance $\sigma^2 = 0$ for $i \neq j$, So

$$\sigma_{12}^2 = \sigma_{13}^2 = \sigma_{21}^2 = \sigma_{23}^2 = \sigma_{31}^2 = \sigma_{32}^2 = 0$$

Covariance matrix,

$$\mathbf{R} = \begin{bmatrix} \sigma_{11}^2 & \sigma_{12}^2 & \sigma_{13}^2 \\ \sigma_{21}^2 & \sigma_{22}^2 & \sigma_{23}^2 \\ \sigma_{31}^2 & \sigma_{32}^2 & \sigma_{33}^2 \end{bmatrix}$$

So,

$$\mathbf{R} = \begin{bmatrix} 0.05 & 0 & 0 \\ 0 & 0.045 & 0 \\ 0 & 0 & 0.05 \end{bmatrix}$$

- e) With the answers to \mathbf{H}' , \mathbf{R} , and \mathbf{z} from Questions 2-4, the weighted least square (WLS) solution is given by $\hat{\mathbf{x}} = (\mathbf{H}'^T \mathbf{R}^{-1} \mathbf{H}')^{-1} \mathbf{H}'^T \mathbf{R}^{-1} \mathbf{z}$. Use MATLAB, Excel, or any other program that can solve the matrix inverse to find \mathbf{R}^{-1} , $(\mathbf{H}'^T \mathbf{R}^{-1} \mathbf{H}')^{-1}$, and finally, the state variable estimate $\hat{\mathbf{x}}$.

We know that

$$\mathbf{R} = \begin{bmatrix} 0.05 & 0 & 0 \\ 0 & 0.045 & 0 \\ 0 & 0 & 0.05 \end{bmatrix}$$

$$\text{So, } \mathbf{R}^{-1} = \begin{bmatrix} 0.05 & 0 & 0 \\ 0 & 0.045 & 0 \\ 0 & 0 & 0.05 \end{bmatrix}^{-1} = \begin{bmatrix} 20 & 0 & 0 \\ 0 & 22.22 & 0 \\ 0 & 0 & 20 \end{bmatrix}$$

$$\text{Now, by removing the coefficients of } \theta_1, \mathbf{H} = \begin{bmatrix} 2 & -2 & 0 \\ 10 & 0 & -10 \\ 0 & -4 & 4 \end{bmatrix} \text{ is replaced by } \mathbf{H}' = \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix}$$

Now,

$$\begin{aligned} (\mathbf{H}'^T \mathbf{R}^{-1} \mathbf{H}')^{-1} &= \left(\begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix}^T \begin{bmatrix} 0.05 & 0 & 0 \\ 0 & 0.045 & 0 \\ 0 & 0 & 0.05 \end{bmatrix}^{-1} \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix} \right)^{-1} \\ &=> \left(\begin{bmatrix} -2 & 0 & -4 \\ 0 & -10 & 4 \end{bmatrix} \begin{bmatrix} 20 & 0 & 0 \\ 0 & 22.22 & 0 \\ 0 & 0 & 20 \end{bmatrix} \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix} \right)^{-1} \\ &=> \left(\begin{bmatrix} -40 & 0 & -80 \\ 0 & -222.2 & 80 \end{bmatrix} \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix} \right)^{-1} \\ &=> \begin{bmatrix} 400 & -320 \\ -320 & 2542 \end{bmatrix}^{-1} \\ &=> \begin{bmatrix} 0.00278 & 0.00034 \\ 0.00034 & 0.00043 \end{bmatrix} \end{aligned}$$

The state variable $\hat{\mathbf{x}} = (\mathbf{H}'^T \mathbf{R}^{-1} \mathbf{H}')^{-1} \mathbf{H}'^T \mathbf{R}^{-1} \mathbf{z}$

$$\Rightarrow \begin{bmatrix} 0.00278 & 0.00034 \\ 0.00034 & 0.00043 \end{bmatrix} \begin{bmatrix} -2 & 0 \\ 0 & -10 \\ -4 & 4 \end{bmatrix}^T \begin{bmatrix} 0.05 & 0 & 0 \\ 0 & 0.045 & 0 \\ 0 & 0 & 0.05 \end{bmatrix}^{-1} \begin{bmatrix} 0.21 \\ 2 \\ 1.18 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 0.00278 & 0.00034 \\ 0.00034 & 0.00043 \end{bmatrix} \begin{bmatrix} -2 & 0 & -4 \\ 0 & -10 & 4 \end{bmatrix} \begin{bmatrix} 20 & 0 & 0 \\ 0 & 22.22 & 0 \\ 0 & 0 & 20 \end{bmatrix} \begin{bmatrix} 0.21 \\ 2 \\ 1.18 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -0.00556 & -0.0034 & -0.00976 \\ -0.00068 & -0.0043 & 0.00036 \end{bmatrix} \begin{bmatrix} 20 & 0 & 0 \\ 0 & 22.22 & 0 \\ 0 & 0 & 20 \end{bmatrix} \begin{bmatrix} 0.21 \\ 2 \\ 1.18 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -0.1112 & -0.075548 & -0.1952 \\ -0.0136 & -0.095546 & 0.0072 \end{bmatrix} \begin{bmatrix} 0.21 \\ 2 \\ 1.18 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} -0.404768 \\ -0.18544 \end{bmatrix}$$