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Kelas : Praktikum Pemrograman Berorientasi Objek – IF E

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## Tugas 4

### Jaringan Syaraf Tiruan

- Source Code

```
>> net = newp([0 1; 0 1], 1);
>> net.IW{1,1} = [-1 1];
>> net.b{1} = [1];
>> p = [[1;1] [1;0] [0;1] [0;0]];
>> t = [1 1 1 0];
>> y = sim(net,p)

y =

     1     1     1     1

>> net = train(net,p,t)

net =

    Neural Network

        name: 'Custom Neural Network'
    userdata: (your custom info)

    dimensions:

        numInputs: 1
        numLayers: 1
        numOutputs: 1
        numInputDelays: 0
        numLayerDelays: 0
        numFeedbackDelays: 0
        numWeightElements: 3
        sampleTime: 1

    connections:

        biasConnect: true
        inputConnect: true
        layerConnect: false
        outputConnect: true

    subobjects:

        input: Equivalent to inputs{1}
        output: Equivalent to outputs{1}

        inputs: {1x1 cell array of 1 input}
        layers: {1x1 cell array of 1 layer}
        outputs: {1x1 cell array of 1 output}
```

```

        biases: {1x1 cell array of 1 bias}
        inputWeights: {1x1 cell array of 1 weight}
        layerWeights: {1x1 cell array of 0 weights}

functions:

    adaptFcn: 'adaptwb'
    adaptParam: (none)
    derivFcn: 'defaultderiv'
    divideFcn: (none)
    divideParam: (none)
    divideMode: 'sample'
    initFcn: 'initlay'
    performFcn: 'mae'
    performParam: .regularization, .normalization
    plotFcns: {'plotperform', plottrainstate}
    plotParams: {1x2 cell array of 2 params}
    trainFcn: 'trainc'
    trainParam: .showWindow, .showCommandLine, .show, .epochs,
                .time, .goal, .max_fail

weight and bias values:

    IW: {1x1 cell} containing 1 input weight matrix
    LW: {1x1 cell} containing 0 layer weight matrices
    b: {1x1 cell} containing 1 bias vector

methods:

    adapt: Learn while in continuous use
    configure: Configure inputs & outputs
    gensim: Generate Simulink model
    init: Initialize weights & biases
    perform: Calculate performance
    sim: Evaluate network outputs given inputs
    train: Train network with examples
    view: View diagram
    unconfigure: Unconfigure inputs & outputs

evaluate:      outputs = net(inputs)

>> net.IW{1,1}

ans =

     1     1

>> net.b{1}

ans =

    -1

>> y = sim(net,p)

y =

     1     1     1     0

```

```
>> e = t-y
```

```
e =
```

```
0    0    0    0
```

### •Penjelasan

```
>> net = newp([0 1; 0 1], 1);
```

Membuat *perceptron* yang dapat mengenali pola fungsi logika “*or*” dengan dua (2) variabel  $x_1$  dan  $x_2$

```
>> net.IW{1,1} = [-1 1];
```

Pendefinisian bobot awal  $w = [-1, 1]$  pada variabel  $x_1$  dan  $x_2$

```
>> net.b{1} = [1];
```

Pendefinisian bias awal  $b=[1]$

```
>> p = [[1;1] [1;0] [0;1] [0;0]];
```

Pendefinisian input

Input	Target
$\begin{pmatrix} 1 \\ 1 \end{pmatrix}$	1
$\begin{pmatrix} 1 \\ 0 \end{pmatrix}$	1
$\begin{pmatrix} 0 \\ 1 \end{pmatrix}$	1
$\begin{pmatrix} 0 \\ 0 \end{pmatrix}$	0

```
>> t = [1 1 1 0];
```

Pendefinisian output

```
>> y = sim(net,p)
```

*Statement* sederhana untuk menghitung hasil keluaran *perceptron*

```
>> y
```

Output *perceptron* tanpa memperdulikan target

```
>> net = train(net,p,t)
```

Perintah untuk menjalankan pelatihan *perceptron*

```
>> net.IW{1,1}
```

Menampilkan nilai bobot optimal (setelah menjalankan pelatihan yang telah dilakukan)

```
>> net.b{1}
```

Menampilkan nilai bias optimal (setelah menjalankan pelatihan yang telah dilakukan)

```
>> y = sim(net,p)
```

*Statement* sederhana untuk menghitung hasil keluaran *perceptron* (setelah menjalankan pelatihan yang telah dilakukan)

```
>> e = t-y
```

Menampilkan tingkat error

- Hasil pelatihan *perceptron* menggunakan *statement train* yang menunjukkan 4 kali epochs untuk mencapai best training performance nya

