# SISO

- 1. If temperature is cold then speed is minimal
- 2. If temperature is cool then speed is slow
- 3. If temperature is pleasant then speed is medium
- 4. If temperature is warm then speed is fast
- 5. If temperature is hot then speed is blast

Consider that the temperature is at 16 and we want our knowledge base to compute the speed.

## **Fuzzification**

Crisp value \$\rightarrow\$ Linguistic variables. 16C \$\rightarrow\$ Cool/Pleasant

#### Temp Cold Cool Pleasant Warm Hot

| 0    | Υ* | Ν  | N  | N  | Ν  |
|------|----|----|----|----|----|
| 5    | Υ  | Υ  | Ν  | Ν  | Ν  |
| 10   | Ν  | Υ  | Ν  | Ν  | Ν  |
| 12.5 | Ν  | γ* | Ν  | Ν  | Ν  |
| 15   | Ν  | Υ  | Ν  | Ν  | Ν  |
| 17.5 | Ν  | Ν  | Υ* | Ν  | Ν  |
| 20   | Ν  | Ν  | Ν  | Υ  | Ν  |
| 22.5 | Ν  | Ν  | Ν  | γ* | Ν  |
| 25   | Ν  | Ν  | Ν  | Υ  | Ν  |
| 27.5 | Ν  | Ν  | Ν  | Ν  | Υ  |
| 30   | Ν  | Ν  | Ν  | Ν  | Υ* |

The fuzzification of the crisp temperature gives the following membership for the Temperature fuzzy set:

```
f(x; a, b, c) = \max{(\min{(\frac{x-a}{b-a}, \frac{c-x}{c-b}))}, 0)}
```

 $\sum_{\text{text{pleasant}}(T) = \max{(\min{(\frac{T-15}{17.5-15}, \frac{20-17.5})}, 0)} $$ \sum_{\text{text{pleasant}}(16) = \max{(\min{(\frac{16-15}{17.5-15}, \frac{20-16}{20-17.5})}, 0)} = 0.4$$ 

#### $\sum_{cold}$ $\sum_{cool}$ $\sum_{cool}$ $\sum_{cool}$

| Temp=16 0             | 0.3      | 0.4      | 0       | 0       |
|-----------------------|----------|----------|---------|---------|
| Fire Rule (#) no (#1) | yes (#2) | yes (#3) | no (#4) | no (#5) |

## Inference

Rules containing the linguistic variables

#### Rules 2 \& 3

Rule #2 and #3 are firing and are essentially the fuzzy patches made out of the cross products of \$\$\text{cool} \times \text{slow}\$\$ \$\$\text{pleasant} \times \text{medium}\$\$\$

#### **RPM Slow Medium**

| 10 | Ν  | Ν  |
|----|----|----|
| 20 | Υ  | Ν  |
| 30 | γ* | Ν  |
| 40 | Υ  | Ν  |
| 50 | Ν  | γ* |
| 60 | Ν  | Ν  |

The fuzzification of the crisp volatile gives the following membership for the Voltage fuzzy set:

 $f(x; a, b, c) = \max{(\min{(\frac{x-a}{b-a}, \frac{c-x}{c-b}))}, 0)}$ 

 $\sum_{v\in \mathbb{N}}(v) = \max_{v\in \mathbb{N}}(v) = \min_{v\in \mathbb{N}}(v) = \min_{v$ 

 $\sum_{v\in \mathbb{N}} (\frac{x-40}{50-40}, \frac{60-50}{0.50}), 0)$ 

Using the min inference method, the output slow is clipped off at 0.3 and medium at 0.4.

| Speed | Slow  | Medium |
|-------|-------|--------|
| 10    | 0     | 0      |
| 12.5  | 0.125 | 0      |
| 15    | 0.25  | 0      |
| 17.5  | 0.3   | 0      |
| 20    | 0.3   | 0      |
| 22.5  | 0.3   | 0      |
| 25    | 0.3   | 0      |
| 27.5  | 0.3   | 0      |
| 30    | 0.3   | 0      |
| 32.5  | 0.3   | 0      |
| 35    | 0.3   | 0      |
| 37.5  | 0.3   | 0      |
| 40    | 0.3   | 0      |
| 42.5  | 0.3   | 0.25   |
| 45    | 0.25  | 0.4    |
| 47.5  | 0.125 | 0.4    |
| 50    | 0     | 0.4    |
| 52.5  | 0     | 0.4    |
| 55    | 0     | 0.4    |
| 57.5  | 0     | 0.25   |
| 60    | 0     | 0      |

# Composition

Create new membership function of the alpha levelled functions for cool and pleasant

The cool and pleasant sets have an output of \$0.3\$ and \$0.4\$ respectively. Using the max composition method, the fuzzy sets for slow and medium have to be given an \$\alpha\$-level cut for these output values respectively.

| Speed | Slow  | Medium | Output of 2 Rules |
|-------|-------|--------|-------------------|
| 10    | 0     | 0      | 0                 |
| 12.5  | 0.125 | 0      | 0.125             |
| 15    | 0.25  | 0      | 0.25              |
| 17.5  | 0.3   | 0      | 0.3               |
| 20    | 0.3   | 0      | 0.3               |
| 22.5  | 0.3   | 0      | 0.3               |
| 25    | 0.3   | 0      | 0.3               |
| 27.5  | 0.3   | 0      | 0.3               |
| 30    | 0.3   | 0      | 0.3               |
| 32.5  | 0.3   | 0      | 0.3               |
| 35    | 0.3   | 0      | 0.3               |
| 37.5  | 0.3   | 0      | 0.3               |
| 40    | 0.3   | 0      | 0.3               |
| 42.5  | 0.3   | 0.25   | 0.3               |
| 45    | 0.25  | 0.4    | 0.4               |
| 47.5  | 0.125 | 0.4    | 0.4               |
| 50    | 0     | 0.4    | 0.4               |
| 52.5  | 0     | 0.4    | 0.4               |
| 55    | 0     | 0.4    | 0.4               |
| 57.5  | 0     | 0.25   | 0.25              |
| 60    | 0     | 0      | 0                 |

# Defuzzification

Examine the fuzzy sets of slow and medium and contain a speed value in rpm.

Now we have to find a way to obtain one single number from the curve. One number corresponding to the speed of the air conditioner's motor.

#### Centroid

Weighted Speed = \$\text{output of 2 rules} \times \text{speed}\$

| Speed Slow Medium ( | Output of 2 Rules | Weighted Speed |
|---------------------|-------------------|----------------|
|---------------------|-------------------|----------------|

| 10    | 0   | 0 | 0     | 0 |
|-------|-----|---|-------|---|
| 12 0. | 125 | 0 | 0.125 | 0 |

| Speed | Slow  | Medium | Output of 2 Rules | <b>Weighted Speed</b> |
|-------|-------|--------|-------------------|-----------------------|
| 12.5  | 0.25  | 0      | 0.25              | 1.5625                |
| 15    | 0.3   | 0      | 0.3               | 3.75                  |
| 17.5  | 0.3   | 0      | 0.3               | 5.25                  |
| 20    | 0.3   | 0      | 0.3               | 6                     |
| 22.5  | 0.3   | 0      | 0.3               | 6.75                  |
| 25    | 0.3   | 0      | 0.3               | 7.5                   |
| 27.5  | 0.3   | 0      | 0.3               | 8.25                  |
| 30    | 0.3   | 0      | 0.3               | 9                     |
| 32.5  | 0.3   | 0      | 0.3               | 9.75                  |
| 35    | 0.3   | 0      | 0.3               | 10.5                  |
| 37.5  | 0.3   | 0      | 0.3               | 11.25                 |
| 40    | 0.3   | 0      | 0.3               | 12                    |
| 42.5  | 0.3   | 0.25   | 0.3               | 12.75                 |
| 45    | 0.25  | 0.4    | 0.4               | 18                    |
| 47.5  | 0.125 | 0.4    | 0.4               | 19                    |
| 50    | 0     | 0.4    | 0.4               | 20                    |
| 52.5  | 0     | 0.4    | 0.4               | 21                    |
| 55    | 0     | 0.4    | 0.4               | 22                    |
| 57.5  | 0     | 0.25   | 0.25              | 14.375                |
| 60    | 0     | 0      | 0                 | 0                     |

Sum of output is 5.925

Sum of weighted speed is 218.687

The computation leads to a single value for the speed - an average computed with respect to the centre of gravity of the output fuzzy set.

Computed speed is \$\frac{218.687}{5.925}=36.91\$RPM

## Mean of maxima

| Speed | <b>Weighted</b> | Output of 2 Rules | Medium | Slow  | Speed |
|-------|-----------------|-------------------|--------|-------|-------|
| 18    | ļ.              | 0.4               | 0.4    | 0.25  | 45    |
| 19    | ļ               | 0.4               | 0.4    | 0.125 | 47.5  |
| 20    | ļ.              | 0.4               | 0.4    | 0     | 50    |
| 21    | ļ.              | 0.4               | 0.4    | 0     | 52.5  |
| 22    | ļ               | 0.4               | 0.4    | 0     | 55    |

Sum of output is 2

Sum of weighted speed is 100

Computed speed it \$\frac{100}{2}=50\$RPM

# **MISO**

We have a pole of length \$1\$, mass \$m\$ as its head and mass \$M\$ at its base, both connected by a weightless shaft. As application of a force \$F\$ is required to control the pole at the base, which can be moved on a horizontal axis.

It is possible to approach the control of the cart pole without the use of differential equations by rules like:

```
If \hat s \approx \hat z_{approximately_zero}(a_{z}) & $d \theta / dt$ is \hat z_{approximately_zero}(a_{z})
Then $F$ is \hat z_{approximately_zero}(a_{z})
```

#### **Fuzzification**

Consider the fuzzy partition of the linguistic variable *angle* (\$\theta\$) expressed through the linguistic terms *negative* (\$n\_{\theta}\$), *approximately zero* (\$az\_{\theta}\$) and *positive* (\$p\_{\theta}\$).

The same can be said about the angular velocity (\$\Theta\$) and for the applied force (\$F\$)

- negative (\$n\_{\Theta}\$), approximately zero (\$az\_{\Theta}\$) and positive (\$p\_{\Theta}\$)
- negative (\$n\_{F}\$), approximately zero (\$az\_{F}\$) and positive (\$p\_{F}\$).

There are nine possible rules for the partitioned fuzzy sets of angle, angular velocity and force.

## $n_{\Omega}$ $az_{\Omega}$ \$p\_{\Theta}\$

| \$n_{\theta}\$  | \$n_{f}\$  | \$n_{f}\$  | \$az_{f}\$ |
|-----------------|------------|------------|------------|
| \$az_{\theta}\$ | \$n_{f}\$  | \$az_{f}\$ | \$p_{f}\$  |
| \$p_{\theta}\$  | \$az_{f}\$ | \$p_{f}\$  | \$p_{f}\$  |

The rule base must be refined further to fine tune the controle. More terms are added, comprising:

- negative big
- negative medium
- negative small
- approximately zero
- positive small
- positive medium
- positive big
  - 1. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$a\_{z}\$
  - 2. If \$\theta\$ is \$p\_{s}\$ and \$\Theta\$ is \$p\_{s}\$ then \$F\$ is \$p\_{s}\$
  - 3. If  $\hat s_{s}^{s}$  and  $\hat s_{z}^{s}$  then  $F^{s}$  is  $p_{s}^{s}$
  - 4. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$n\_{s}\$ then \$F\$ is \$p\_{s}\$
  - 5. If \$\theta\$ is \$n\_{s}\$ and \$\Theta\$ is \$n\_{b}\$ then \$F\$ is \$p\_{s}\$

```
6. If $\theta$ is $p_{b}$ and $\Theta$ is $p_{s}$ then $F$ is $p_{m}$
```

- 7. If \$\theta\$ is \$p\_{m}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$p\_{m}\$
- 8. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$n\_{m}\$ then \$F\$ is \$p\_{m}\$
- 9. If \$\theta\$ is \$p\_{b}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$p\_{b}\$
- 10. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$n\_{b}\$ then \$F\$ is \$p\_{b}\$
- 11. If \$\theta\$ is \$p\_{s}\$ and \$\Theta\$ is \$p\_{b}\$ then \$F\$ is \$n\_{s}\$
- 12. If  $\hat s= x_{z}\$  and  $\hat s= x_{s}\$  then \$F\$ is  $x_{s}\$
- 13. If \$\theta\$ is \$n\_{s}\$ and \$\Theta\$ is \$n\_{s}\$ then \$F\$ is \$n\_{s}\$
- 14. If \$\theta\$ is \$n\_{s}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$n\_{s}\$
- 15. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$p\_{m}\$ then \$F\$ is \$n\_{m}\$
- 16. If \$\theta\$ is \$n\_{m}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$n\_{m}\$
- 17. If \$\theta\$ is \$n {b}\$ and \$\Theta\$ is \$n {s}\$ then \$F\$ is \$n {m}\$
- 18. If \$\theta\$ is \$a\_{z}\$ and \$\Theta\$ is \$p\_{b}\$ then \$F\$ is \$n\_{b}\$
- 19. If \$\theta\$ is \$n\_{b}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$n\_{b}\$

|                      | \$n_{b}\$ | \$n_{m}\$ | \$n_{s}\$ | \$a_{z}\$ | \$p_{s}\$ | \$p_{m}\$ | \$p_{b}\$ |
|----------------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|
| \$\theta_{max}\$     | -67.5     | -45       | -22.5     | 0         | 22.5      | 45.0      | 67.5      |
| \$\theta_{min}^{1}\$ | -45       | -22.5     | 0         | 22.5      | 45        | 67.5      | 90        |
| \$\theta_{min}^{2}\$ | -90       | -67.5     | -45       | -22.5     | 0         | 22.5      | 45        |
| \$\Theta_{max}\$     | -33.75    | -22.5     | -11.25    | 0         | 11.25     | 22.5      | 33.75     |
| \$\Theta_{min}^{1}\$ | -22.5     | -11.25    | 0         | 11.25     | 22.5      | 33.75     | 45        |
| \$\Theta_{min}^{2}\$ | -45       | -33.75    | -22.5     | -11.25    | 0         | 11.25     | 22.5      |
| \$F_{max}\$          | -7.5      | -5        | -2.5      | 0         | 2.5       | 5         | 7.5       |
| \$F_{min}^{1}\$      | -5        | -2.5      | 0         | 2.5       | 5         | 7.5       | 10        |
| \$F_{min}^{2}\$      | -10       | -7.5      | -5        | -2.5      | 0         | 2.5       | 5         |

Consider  $\hat s=36$  ( $p_{s}$  and  $p_{s}$ ) and  $\hat s=2.25$  ( $a_{z}$ ).

 $\mbox{\mu_{ps}(\theta) = \max{(\min{(\frac{a}-a){b-a}, \frac{c-b})}, 0)}$$ $\mu_{ps}(36) = \max{(\min{(\frac{36-0}{22.5-0}, \frac{45-36}{45-22.5})}, 0)}$$ $\mu_{ps}(36) = \max{(\min{(1.6, 0.4)}, 0)} = 0.4$$$ 

 $\mbox{\mu_{pm}(\theta) = \max{(\min{(\sqrt{c-b})}, 0)} $ $\sum_{g=\max{(\sqrt{(\sqrt{c-1})}, 0)} $ $\sum_{g=\max{(\sqrt{(0.6, 1.4)}, 0)} = 0.6$} $$ 

 $\begin{tabular}{l} $$\mu_{az}(\theta) = \max\{(\min\{(\frac{az}(-2.25), 0)\} $ $\\mu_{az}(-2.25) = \max\{(\min\{(\frac{11.25-(-2.25)}{11.25-0})\}, 0)\} $ $\\mu_{az}(-2.25) = \max\{(\min\{(0.8, 1.2)\}, 0)\} = 0.8$$ 

- 1. If \$\theta\$ is \$p\_{s}\$ and \$\Theta\$ is \$a\_{z}\$ then \$F\$ is \$p\_{s}\$
- 2. If  $\hat s = f^{s}$  and  $f^{s}$  then  $f^{s}$  is  $p_{m}$

## Inference

 $\sum_{ps}(F) = \max{(\min_{(\frac{F-a}{b-a}, \frac{c-F}{c-b})}, 0)}$  \$\$\mu\_{ps}(F) = \max{(\min\_{(\frac{F-0}{2.5-0}, \frac{5-F}{5-2.5})}, 0)}\$\$

 $\mbox{pm}(F) = \max{(\min{(\frac{F-a}{b-a}, \frac{c-F}{c-b})}, 0)}$  \$\$\mu\_{pm}(F) = \max{(\min{(\frac{F-2.5}{5-2.5}, \frac{7.5-F}{7.5-5})}, 0)}\$\$

For rule number 3, we take the  $\alpha = 1.4\$  For rule number 7, we take the  $\alpha = 1.4\$  For rule number 7, we take the  $\alpha = 1.4\$ 

| F   | \$\mu_{ps}\$ | \$\mu_{pm}\$ |
|-----|--------------|--------------|
| 0   | 0            | 0            |
| 0.5 | 0.2          | 0            |
| 1   | 0.4          | 0            |
| 1.5 | 0.4          | 0            |
| 2   | 0.4          | 0            |
| 2.5 | 0.4          | 0            |
| 3   | 0.4          | 0.2          |
| 3.5 | 0.4          | 0.3          |
| 4   | 0.4          | 0.6          |
| 4.5 | 0.4          | 0.6          |
| 5   | 0.2          | 0.6          |
| 5.5 | 0            | 0.6          |
| 6   | 0            | 0.6          |
| 6.5 | 0            | 0.4          |
| 7   | 0            | 0.2          |
| 7.5 | 0            | 0            |

# Composition

| F   | \$\mu_{ps}\$ | \$\mu_{pm}\$ | Output |
|-----|--------------|--------------|--------|
| 0   | 0            | 0            | 0      |
| 0.5 | 0.2          | 0            | 0.2    |
| 1   | 0.4          | 0            | 0.4    |
| 1.5 | 0.4          | 0            | 0.4    |
| 2   | 0.4          | 0            | 0.4    |
| 2.5 | 0.4          | 0            | 0.4    |
| 3   | 0.4          | 0.2          | 0.4    |
| 3.5 | 0.4          | 0.3          | 0.4    |
| 4   | 0.4          | 0.6          | 0.6    |
| 4.5 | 0.4          | 0.6          | 0.6    |
| 5   | 0.2          | 0.6          | 0.6    |
| 5.5 | 0            | 0.6          | 0.6    |
| 6   | 0            | 0.6          | 0.6    |
| 6.5 | 0            | 0.4          | 0.4    |
| 7   | 0            | 0.2          | 0.2    |
| 7.5 | 0            | 0            | 0      |

# Defuzzification

## **Maximum Criterion Model**

## F Output Weighted Output

| 4   | 0.6 | 2.4 |
|-----|-----|-----|
| 4.5 | 0.6 | 2.7 |
| 5   | 0.6 | 3   |
| 5.5 | 0.6 | 3.3 |
| 6   | 0.6 | 3.6 |

Any value of the force between 4 and 6 will do to steady the cartpole.

## Mean of Maximum

# F Output Weighted Output

| 4   | 0.6 | 2.4 |
|-----|-----|-----|
| 4.5 | 0.6 | 2.7 |
| 5   | 0.6 | 3   |
| 5.5 | 0.6 | 3.3 |
| 6   | 0.6 | 3.6 |

Sum of output is 3
Sum of weighted output is 15
\$\$F=\frac{15}{3} = 5\$\$

#### COG

## F Output Weighted Output

| 0   | 0   | 0   |
|-----|-----|-----|
| 0.1 | 0.2 | 0.5 |
| 0.4 | 0.4 | 1   |
| 0.6 | 0.4 | 1.5 |
| 0.8 | 0.4 | 2   |
| 1   | 0.4 | 2.5 |
| 1.2 | 0.4 | 3   |
| 1.4 | 0.4 | 3.5 |
| 2.4 | 0.6 | 4   |
| 2.7 | 0.6 | 4.5 |
| 3   | 0.6 | 5   |
| 3.3 | 0.6 | 5.5 |
| 3.6 | 0.6 | 6   |
| 3.9 | 0.4 | 6.5 |
| 4.2 | 0.2 | 7   |
| 0   | 0   | 7.5 |
|     |     |     |

Sum of output is 6.2 Sum of weighted output is 24.5  $F=\frac{24.5}{6.2} = 3.95$ \$