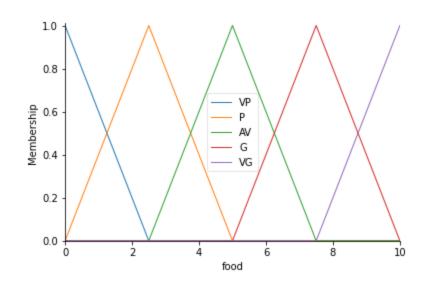
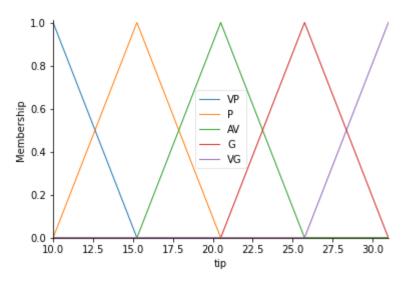
ARTIFICIAL INTELLIGENCE FUZZY LOGIC

Prof. Nguyen Truong Thinh

```
import numpy as np
import skfuzzy.control as ctrl
#Không gian phổ quát giúp tính toán nhanh hơn mà không phải quan tâm đô chính xác.
#Chỉ có những điểm quan trọng được bao gồm ở đây;
#làm cho nó có độ phân giải cao hơn là không cần thiết.
universe = np.linspace(0, 10, 61)
universe1 = np.linspace(10, 31, 61)
# Tạo ra 3 biến mờ: 2 vào 1 ra
food = ctrl.Antecedent(universe, 'food')
service = ctrl.Antecedent(universe, 'service')
tip = ctrl.Consequent(universe1, 'tip')
# Chung ta đặt VP: Very Poor, P: Poor, AV; Average, G: Goog, VG: very Goog
names = ['VP', 'P', 'AV', 'G', 'VG']
food.automf(names=names)
service.automf(names=names)
tip.automf(names=names)
```





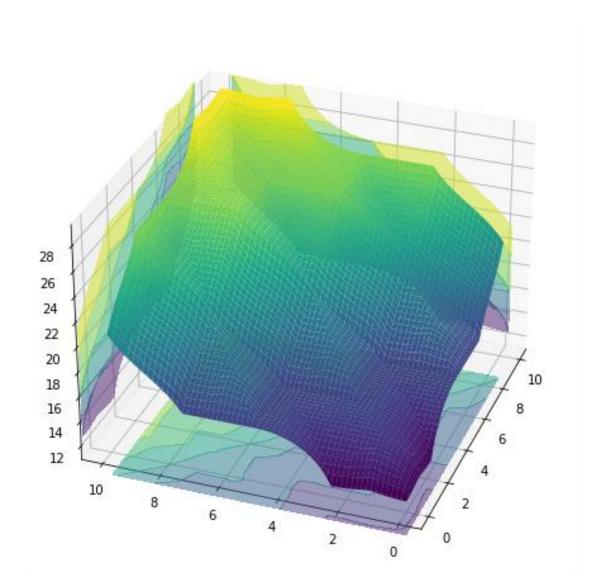
```
rule0 = ctrl.Rule(antecedent= ((food['VP'] & service['VP']) |
                              (food['P'] & service['VP']) |
                              (food['VP'] & service['P'])),
                  consequent=tip['VP'], label='rule VP')
rule1 = ctrl.Rule(antecedent=((food['VP'] & service['AV']) |
                              (food['VP'] & service['G']) |
                              (food['P'] & service['P']) |
                              (food['P'] & service['AV']) |
                              (food['AV'] & service['P']) |
                              (food['AV'] & service['VP']) |
                              (food['G'] & service['VP'])),
                  consequent=tip['P'], label='rule P')
```

```
rule2 = ctrl.Rule(antecedent=((food['VP'] & service['VG']) |
                              (food['P'] & service['G']) |
                              (food['AV'] & service['AV']) |
                              (food['G'] & service['P']) |
                              (food['VG'] & service['VP'])),
                  consequent=tip['AV'], label='rule AV')
rule3 = ctrl.Rule(antecedent=((food['P'] & service['VG']) |
                              (food['AV'] & service['VG']) |
                              (food['AV'] & service['G']) |
                              (food['G'] & service['G']) |
                              (food['G'] & service['AV']) |
                              (food['VG'] & service['AV']) |
                              (food['VG'] & service['P'])),
                  consequent=tip['G'], label='rule G')
rule4 = ctrl.Rule(antecedent=((food['G'] & service['VG']) |
                              (food['VG'] & service['VG']) |
                              (food['VG'] & service['G'])),
                  consequent=tip['VG'], label='rule VG')
```

```
system = ctrl.ControlSystem(rules=[rule0, rule1, rule2, rule3, rule4])
```

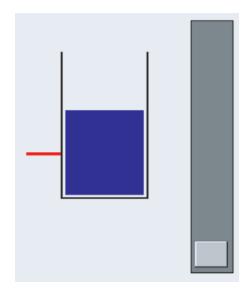
```
sim = ctrl.ControlSystemSimulation(system, flush_after_run=61 * 61 + 1)
```

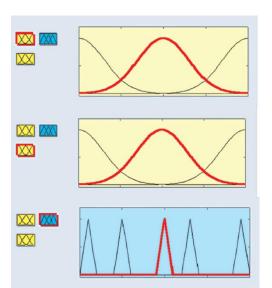
```
upsampled = np.linspace(0, 10, 61)
x, y = np.meshgrid(upsampled, upsampled)
z = np.zeros like(x)
for i in range(61):
   for i in range(61):
        sim.input['food'] = x[i, j]
        sim.input['service'] = y[i, j]
        sim.compute()
        z[i, j] = sim.output['tip']
import matplotlib.pyplot as plt
from mpl toolkits.mplot3d import Axes3D
fig = plt.figure(figsize=(8, 8))
ax = fig.add subplot(111, projection='3d')
surf = ax.plot surface(x, y, z, rstride=1, cstride=1, cmap='viridis',
                      linewidth=0.4, antialiased=True)
cset = ax.contourf(x, y, z, zdir='z', offset=9, cmap='viridis', alpha=0.5)
cset = ax.contourf(x, y, z, zdir='x', offset=11, cmap='viridis', alpha=0.5)
cset = ax.contourf(x, y, z, zdir='y', offset=11, cmap='viridis', alpha=0.5)
ax.view init(30, 200)
```



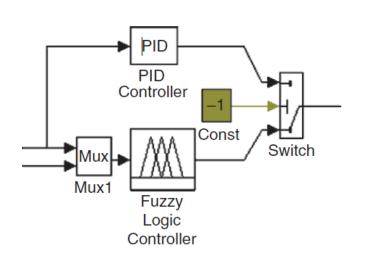
Ex 1: water level control

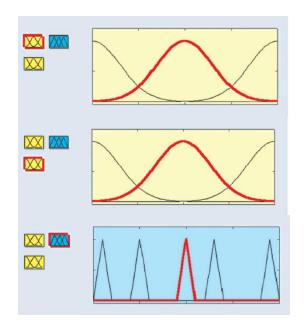
- Rules should be as follows:
 - 1. IF (level is okay) THEN (valve is no change) (1)
 - 2. IF (level is low) THEN (valve is open fast) (1)
 - 3. IF (level is high)THEN (valve is close fast) (1)
- 4. IF (level is okay) and (rate is positive) THEN (valve is close slow) (1)
- 5. IF (level is okay) and (rate is negative) THEN (valve is open slow) (1)

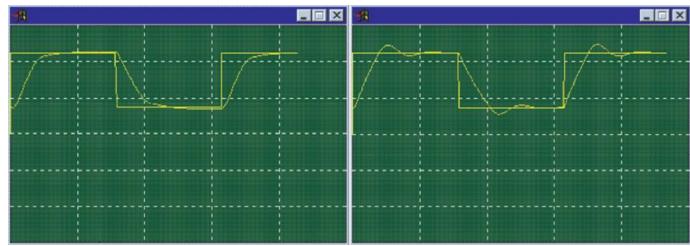


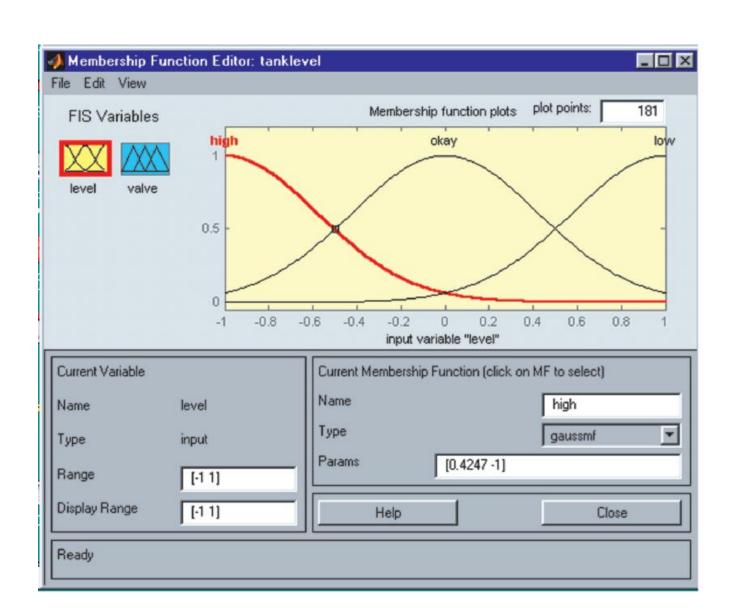


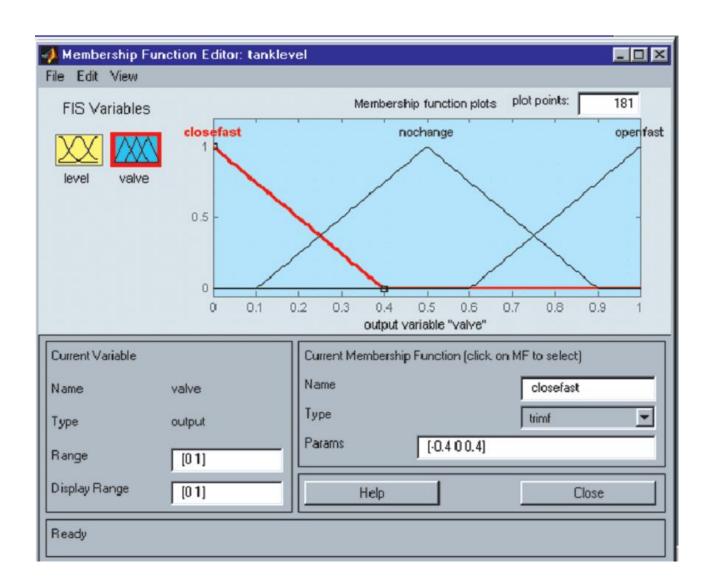
Fuzzy Controllers: water level control

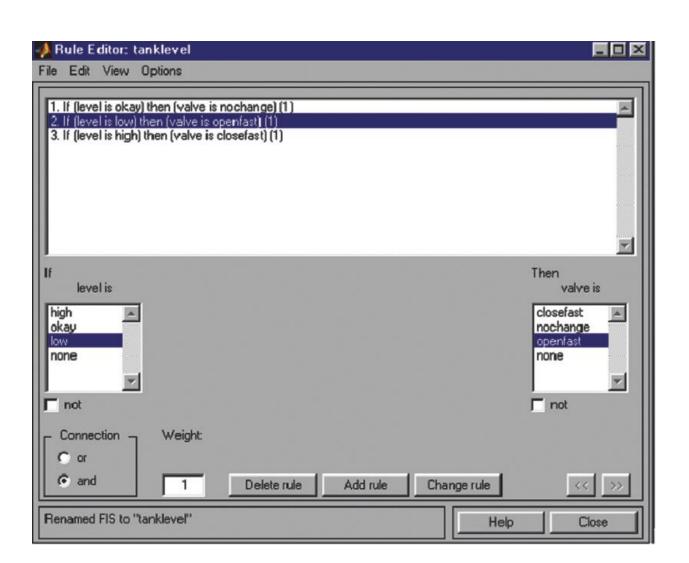


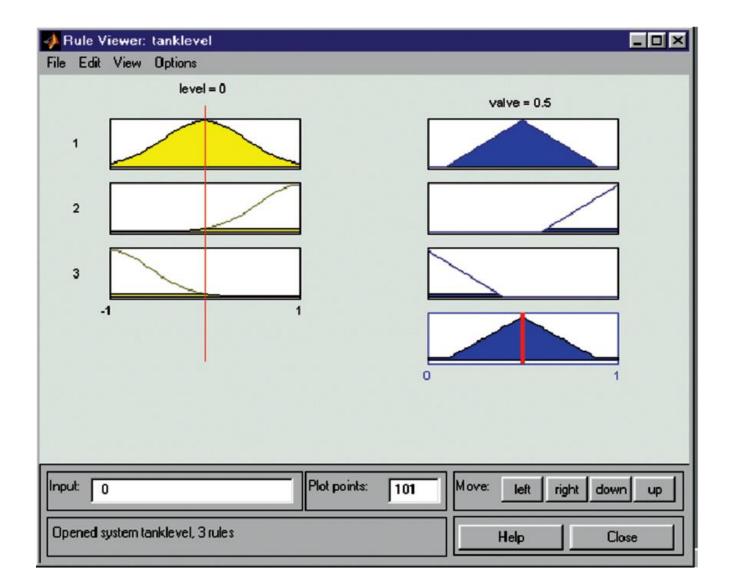


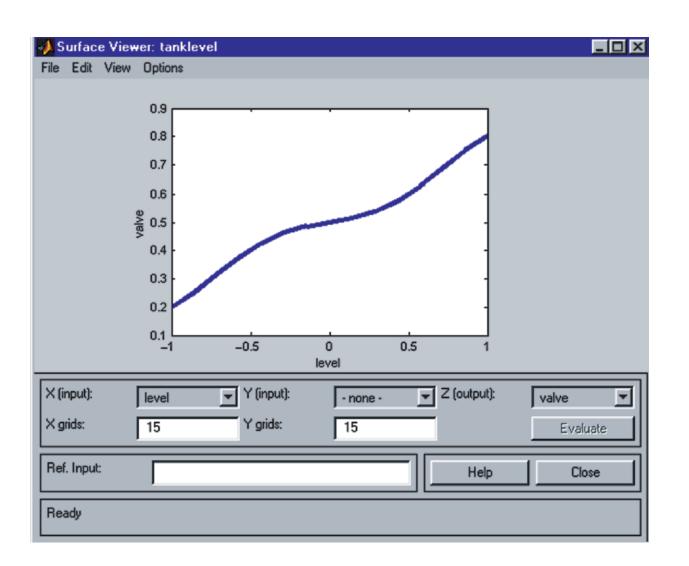




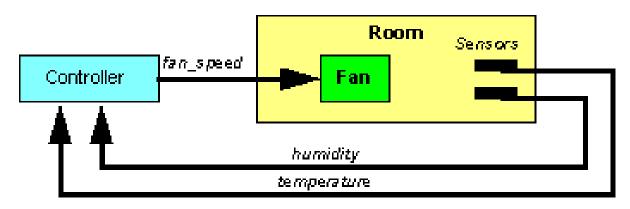








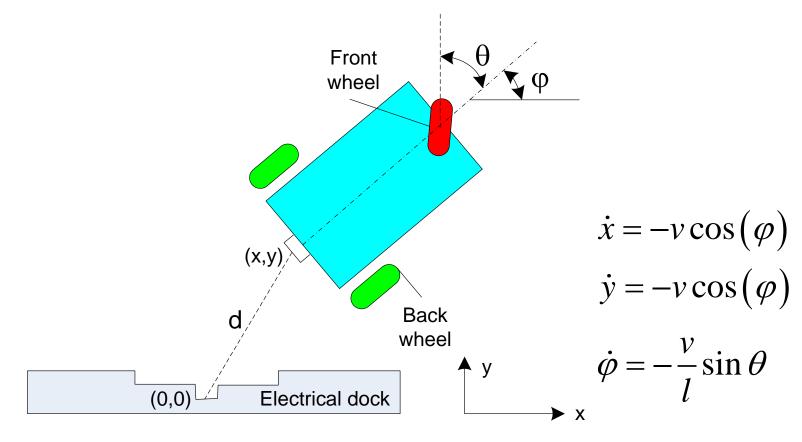
Ex 2: Temperature Controller



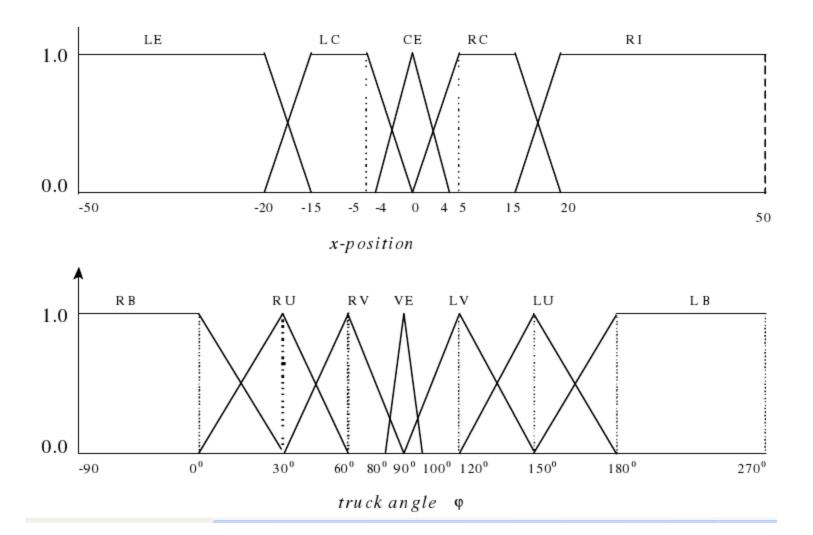
- IF temperature IS cold AND humidity IS high THEN fan_spd IS high IF temperature IS cool AND humidity IS high THEN fan_spd IS medium IF temperature IS warm AND humidity IS high THEN fan_spd IS low IF temperature IS hot AND humidity IS high THEN fan_spd IS zero
- IF temperature IS cold AND humidity IS med THEN fan_spd IS medium IF temperature IS cool AND humidity IS med THEN fan_spd IS low IF temperature IS war m AND humidity IS med THEN fan_spd IS zero IF temperature IS hot AND humidity IS med THEN fan_spd IS zero
- IF temperature IS cold AND humidity IS low THEN fan_spd IS medium
 IF temperature IS cool AND humidity IS low THEN fan_spd IS low
 IF temperature IS warm AND humidity IS low THEN fan_spd IS zero
 IF temperature IS hot AND humidity IS low THEN fan_spd IS zero

Ex 3: Mobile Robot

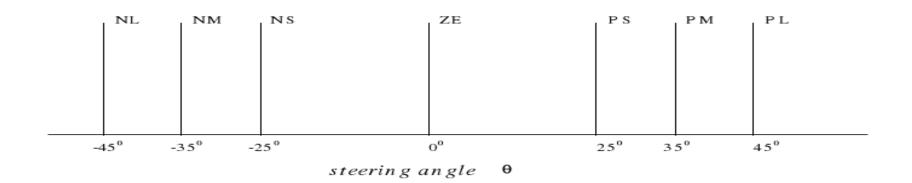
• Design a Fuzzy Logic Controller (FLC) able to back up a mobile robot into a docking stration from any initial position that has enough clearance from the docking station.



Mobile robot



Mobile robot



φx	LE	LC	CE	RC	RI
RL	NL 1	NL 2	NM ³	NM ⁴	NS ⁵
RU	NL 6	NL 7	NM	NS	P S
RV	NL	NM	NS	PS	P M
VE	NM	NM	ZE 18	P M	P M
LV	NM	NS	PS	P M	PL
LU	NS	PS	РМ	PL	30 P L
LL	P S 31	P M 32	P M	94 P L	35 P L

Ex 4: A fuzzy vacuum cleaner

- Let us try to develop the rules table for the fuzzy controller of a vacuum cleaner. This controller should regulate the force of sucking dust from a surface being cleaned. This force can be described as a linguistic variable with values: very strong, strong, ordinary, weak, very weak. The input of this controller should obviously consider an amount of dust on the surface. The surface can be very dirty, dirty, rather dirty, almost clean, clean. The controller can change the force depending on how dirty the surface is. One can propose the following set of rules to describe the controller operation:
 - if surface is very dirty then force is very strong,
 - if surface is dirty then force is strong,
 - if surface is rather dirty then force is ordinary,
 - if surface is almost clean then force is weak,
 - if surface is clean then force is very weak.

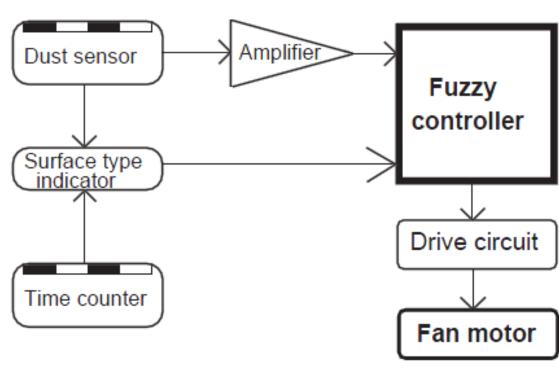
A fuzzy vacuum cleaner

Table 3.3 Rules table for a fuzzy vacuum cleaner						
Surface Force						
very dirty	very strong					
dirty	strong					
rather dirty	ordinary					
almost clean	weak					
clean	very weak					

Table 3.4 Rules table for surface type and dust amount						
	clean	almost clean	rather dirty	dirty	very dirty	
wood	very weak	very weak	weak	ordinary	strong	
tatami	very weak	weak	ordinary	strong	very strong	
carpet	weak	ordinary	ordinary	strong	very strong	

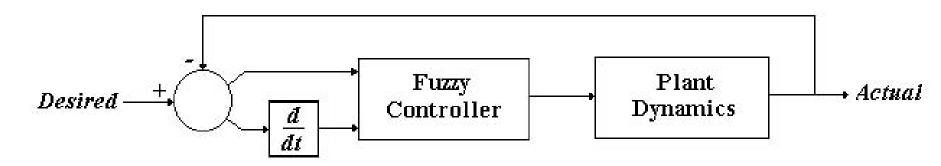
A fuzzy vacuum cleaner





Ex 5: Automobile cruise control

 we discussed the design of an automobile cruise control system using the standard approach. Here we solve the same problem using the Mamdani method, a fuzzy approach. It should become clear that this fuzzy approach, which provides a model-free approach to developing a controller, is simpler.

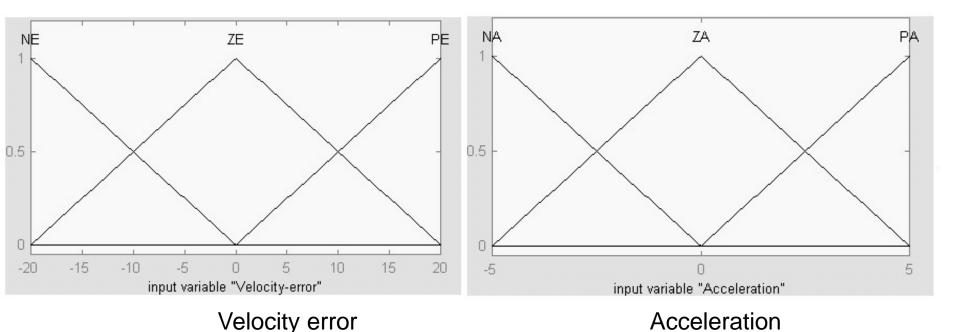


Fuzzy variables: velocity error, acceleration, and engine force.

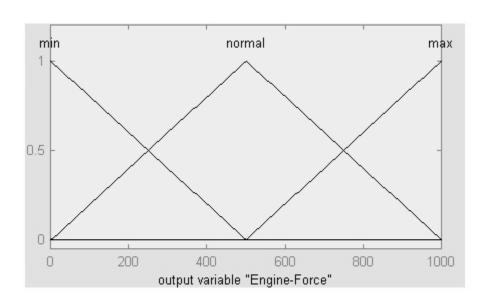
Automobile cruise control

Following are some example rules:

- If velocity error is positive and acceleration is negative then apply maximum force.
- If velocity error is negative and acceleration is positive then apply minimum force.
- If velocity error is zero and acceleration is zero then apply normal force.



Membership functions

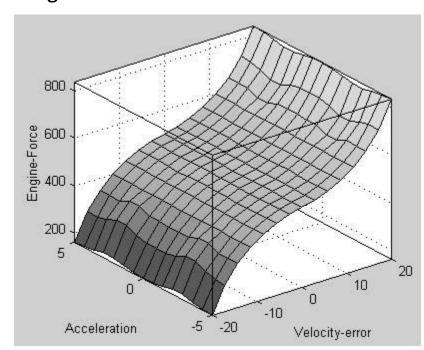


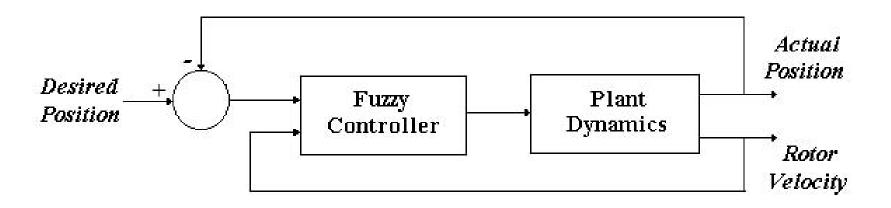
Engine force

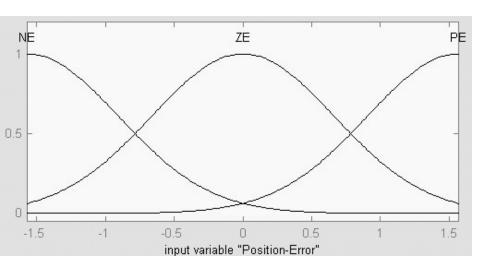
		Acceleration				
	г	NA	ZA	PA		
ror	NE	Min	Min	Min		
velocity-Error	ZE	Normal	Normal	Normal		
let.	PE	Max	Мах	Мах		

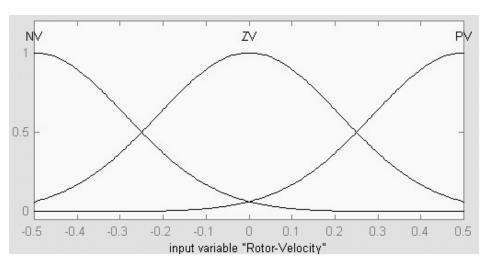
Control Rules

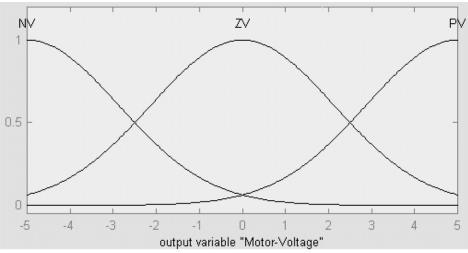
- 1. If velocity error is NE and acceleration is NA, then engine force is Min.
- 2. If velocity error is NE and acceleration is ZE, then engine force is Min.
- 3. If velocity error is NE and acceleration is PA, then engine force is Min.
- 4. If velocity error is ZE and acceleration is NA, then engine force is Normal.
- 5. If velocity error is ZE and acceleration is ZA, then engine force is Normal.
- 6. If velocity error is ZE and acceleration is PA, then engine force is Normal.
- 7. If velocity error is PE and acceleration is NA, then engine force is Max.
- 8. If velocity error is PE and acceleration is ZA, then engine force is Max.
- 9. If velocity error is PE and acceleration is PA, then engine force is Max.







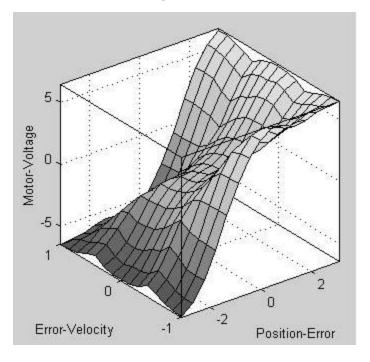


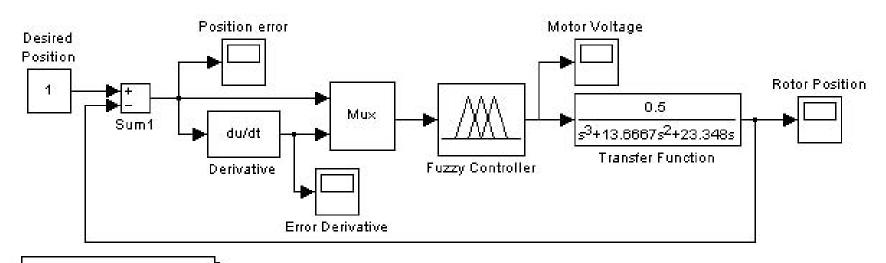


- This knowledge of the system behavior allows us to formulate a set of general rules as, for example,
- 1. If position error is positive and velocity is negative, then apply positive voltage.
- 2. If position error is negative and velocity is positive, then apply negative voltage.
- 3. If position error is zero and velocity is zero, then apply zero voltage.

		Error-Velocity						
		NV .	ZV	PV				
.or	NE	Negative Voltage	Negative Voltage	Negative Voltage				
Position-Error	ZE	Negative Voltage	Zero Voltage	Positive Voltage				
Posi	PE	Positive Voltage	Positive Voltage	Positive Voltage				

- 1. If position error is NE and velocity is NA, then motor voltage is Negative.
- 2. If position error is NE and velocity is ZE, then motor voltage is Negative.
- 3. If position error is NE and velocity is PA, then motor voltage is Negative.
- 4. If position error is ZE and velocity is NA, then motor voltage is Zero.
- 5. If position error is ZE and velocity is ZA, then motor voltage is Zero.
- 6. If position error is ZE and velocity is PA, then motor voltage is Zero.
- 7. If position error is PE and velocity is NA, then motor voltage is Positive.
- 8. If position error is PE and velocity is ZA, then motor voltage is Positive.
- 9. If position error is PE and velocity is PA, then motor voltage is Positive.





Variable Initialization

Ex 6:

Example 4.4. Using your own intuition and definitions of the universe of discourse, plot fuzzy membership functions for "weight of people."

Solution. The universe of discourse is the weight of people. Let the weights be in "kg" – kilogram.

Let the linguistic variables are:

 $\begin{array}{lll} \mbox{Very light} & - & w \leq 30 \\ \mbox{Light} & - & 30 < w \leq 45 \\ \mbox{Average} & - & 45 < w \leq 60 \\ \mbox{Heavy} & - & 60 < w \leq 75 \\ \end{array}$

Very heavy - w > 75

Representing this using triangular membership function, as shown in Fig. 4.12.

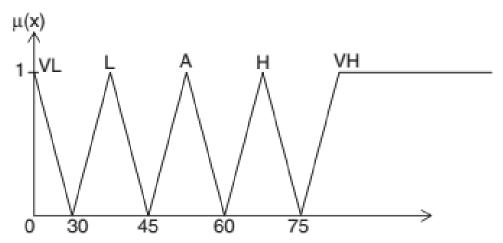


Fig. 4.12. Membership function of weight of people

Ex 7:

Example 4.5. Using your own intuition, plot the fuzzy membership function for the age of people.

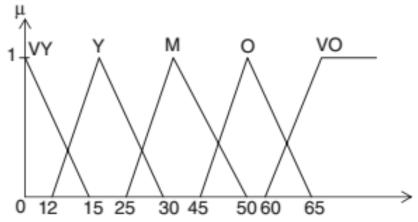


Fig. 4.13. Membership function for age of profile

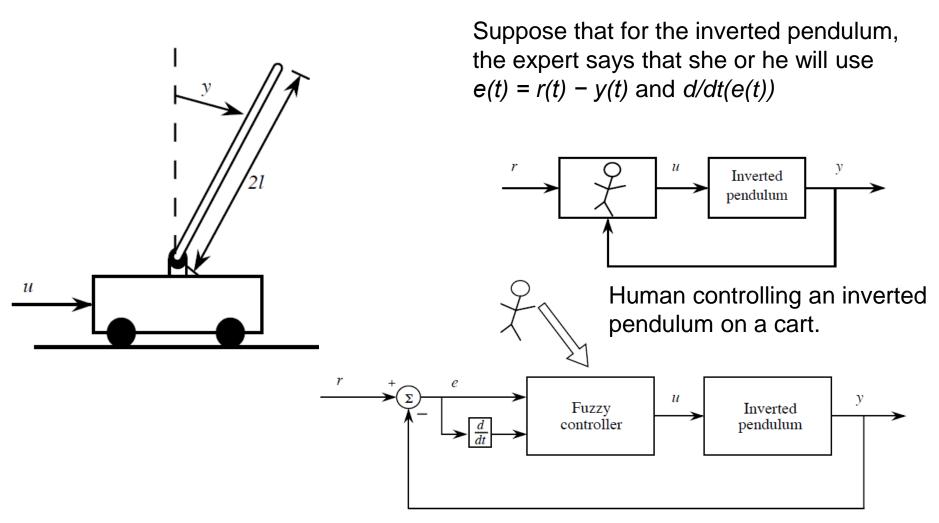
Solution.

The linguistic variables are defined as, let A denotes age in years.

- (1) Very young (vy) -A < 15
- (2) Young (y) $-12 \le A < 30$
- (3) Middle aged (m) $-25 \le A < 50$
- (4) Old (o) $-45 \le A < 65$
- (5) Very old (vo) 60 < A</p>

This is represented using triangular membership, as shown in Fig. 4.13.

Ex 8: Inverted pendulum on a cart



Fuzzy controller for an inverted pendulum on a cart

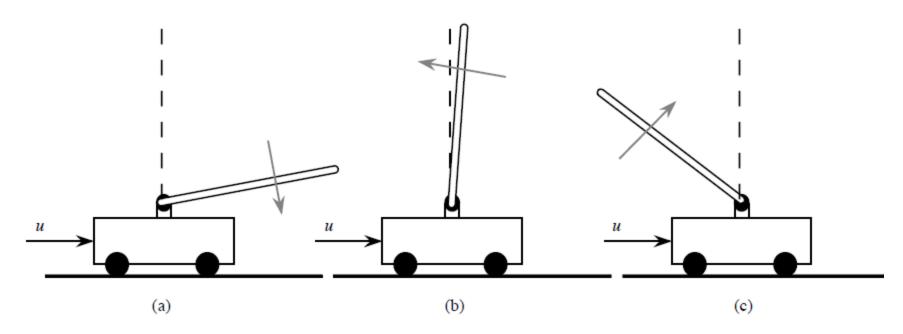
Linguistic Descriptions

- There will be "linguistic variables" that describe each of the time varying fuzzy controller inputs and outputs.
- For the inverted pendulum,
 - * "error" describes e(t)
 - * "change-in-error" describes d/dt(e(t))
 - * "force" describes *u(t)*

Linguistic Descriptions

- Suppose for the pendulum example that "error," "change-in-error," and "force" take on the following values:
 - * "neglarge"
 - * "negsmall"
 - * "zero"
 - * "possmall"
 - * "poslarge"

Rules



Inverted pendulum in various positions.

If error is neglarge and change-in-error is neglarge Then force is poslarge

If error is zero and change-in-error is possmall Then force is negsmall

If error is poslarge and change-in-error is negsmall Then force is negsmall

Linguistic Descriptions

For an even shorter description we could use integers:

```
"-2" to represent "neglarge"
```

```
"-1" to represent "negsmall"
```

```
"0" to represent "zero"
```

- "1" to represent "possmall"
- "2" to represent "poslarge"

Rule-Bases

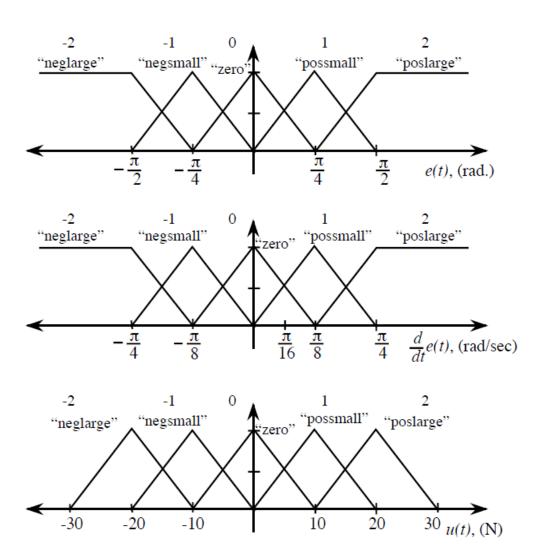
• For the pendulum problem, with two inputs and five linguistic values for each of these, there are at most $5^2 = 25$ possible rules

"force"		"change-in-error" \dot{e}				
u		-2	-1	0	1	2
	-2	2	2	2	1	0
"error"	-1	2	2	1	0	-1
e	0	2	1	0	-1	-2
	1	1	0	-1	-2	-2
	2	0	-1	-2	-2	-2

Rule Table for the Inverted Pendulum

if (input 1 is membership function 1) and/or (input 2 is membership function 2) and/or... then (outputn is output membership functionn).

Membership Functions



Mamdani's Fuzzy Inference Method

