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UNIVERSITÄT  
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FAKULTÄT FÜR  
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# **Fuzzy Systems**

## **Fuzzy Control**

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**SS 2020**



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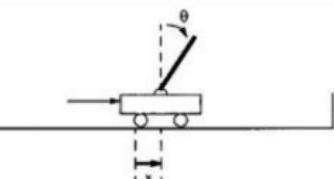
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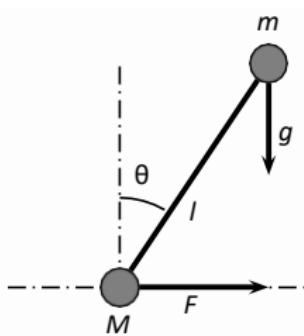
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# Fuzzy Controller

## Example: Cartpole Problem



Balance an upright standing pole



Lower end of pole can be moved unrestrained along horizontal axis.

Mass  $m$  at foot and mass  $M$  at head.

Influence of mass of shaft itself is negligible.

Determine force  $F$  (control variable) that is necessary to balance pole standing upright.

That is measurement of following output variables:

- angle  $\theta$  of pole in relation to vertical axis,
- change of angle, i.e. triangular velocity  $\dot{\theta} = \frac{d\theta}{dt}$

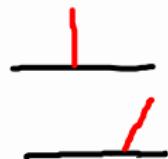
Both should converge to zero.

# Qualitative Description of a controller - as Rule System

		$\dot{\vartheta}$						
		nb	nm	ns	az	ps	pm	pb
$\vartheta$	nb			ps	pb			
	nm				pm			
	ns	nm		ns	ps			
	az	nb	nm	ns	az	ps	pm	pb
	ps				ns	ps		pm
	pm				nm			
	pb				nb	ns		

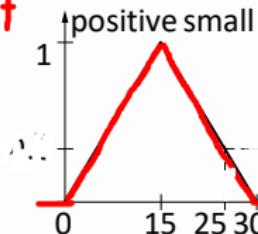
19 rules for cartpole problem, e.g.

If  $\vartheta$  is approximately zero and  $\dot{\vartheta}$  is negative medium  
then  $F$  is positive medium.

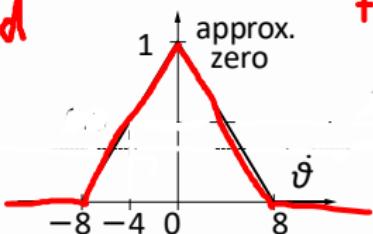


# One Rule: Evaluation

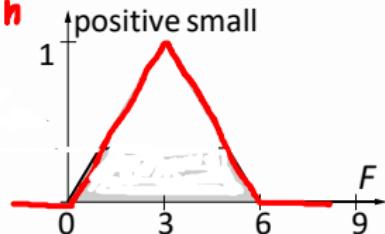
if



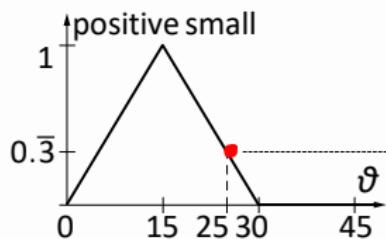
and



then

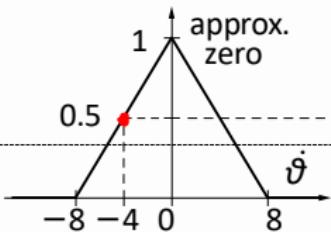


# One Rule: Evaluation

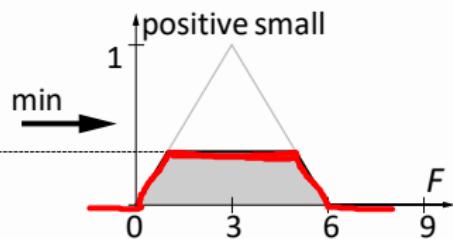


Input

$$\vartheta = 25$$



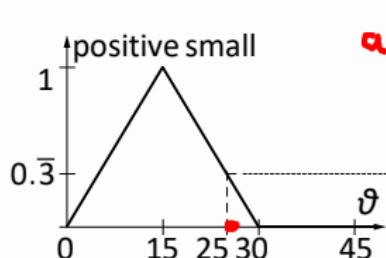
$$\vartheta = -4$$



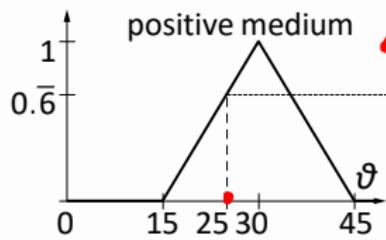
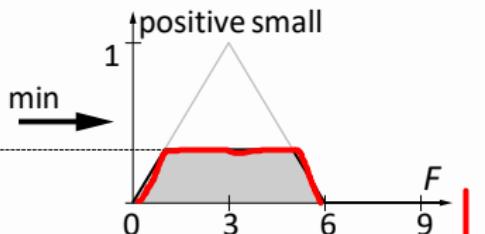
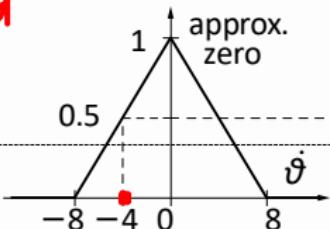
$$F = ?$$

Output

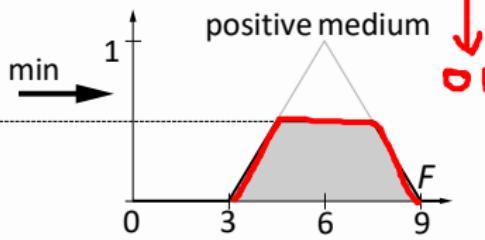
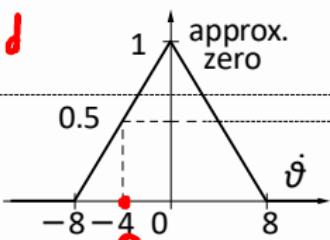
# Several Rules: Evaluation



and



and

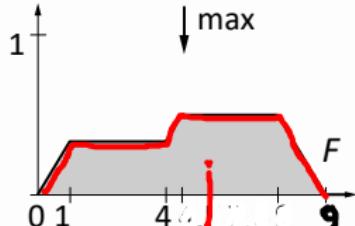


Rule evaluation for Mamdani-Assilian controller.

Input tuple  $(25, -4)$  leads to fuzzy output.

Crisp output is determined by defuzzification.

$(25, -4) \rightarrow ?$





# Definition of Table-based Control Function I

Given is the measurement  $(x_1, \dots, x_n) \in X_1 \times \dots \times X_n$

Consider a rule R

if  $\mu^{(1)}$  and ... and  $\mu^{(n)}$  then  $\eta$ .

The fuzzyfication unit computes for ~~the~~ input  $(x_1, \dots, x_n)$  a „degree of fulfillment“ of the premise of the rule:

For  $1 \leq v \leq n$ , the membership degree  $\mu^{(v)}(x_v)$  is calculated. The n degrees are combined conjunctively with the min-operator and give the fulfillment degree  $\alpha$

For each rule  $R_r$  with  $1 \leq r \leq k$ , compute the fulfillment degree  $\alpha_r$



## Definition of Table-based Control Function II

For the input  $(x_1, \dots, x_n)$  and a rule  $R$  the decision unit calculates the output

$$\begin{aligned}\mu_{x_1, \dots, x_n}^{\text{output}(R)} : Y &\rightarrow [0, 1], \\ y &\mapsto \min (\mu^{(1)}(x_1), \dots, \mu^{(n)}(x_n), \eta(y)).\end{aligned}$$

## Definition of Table-based Control Function III

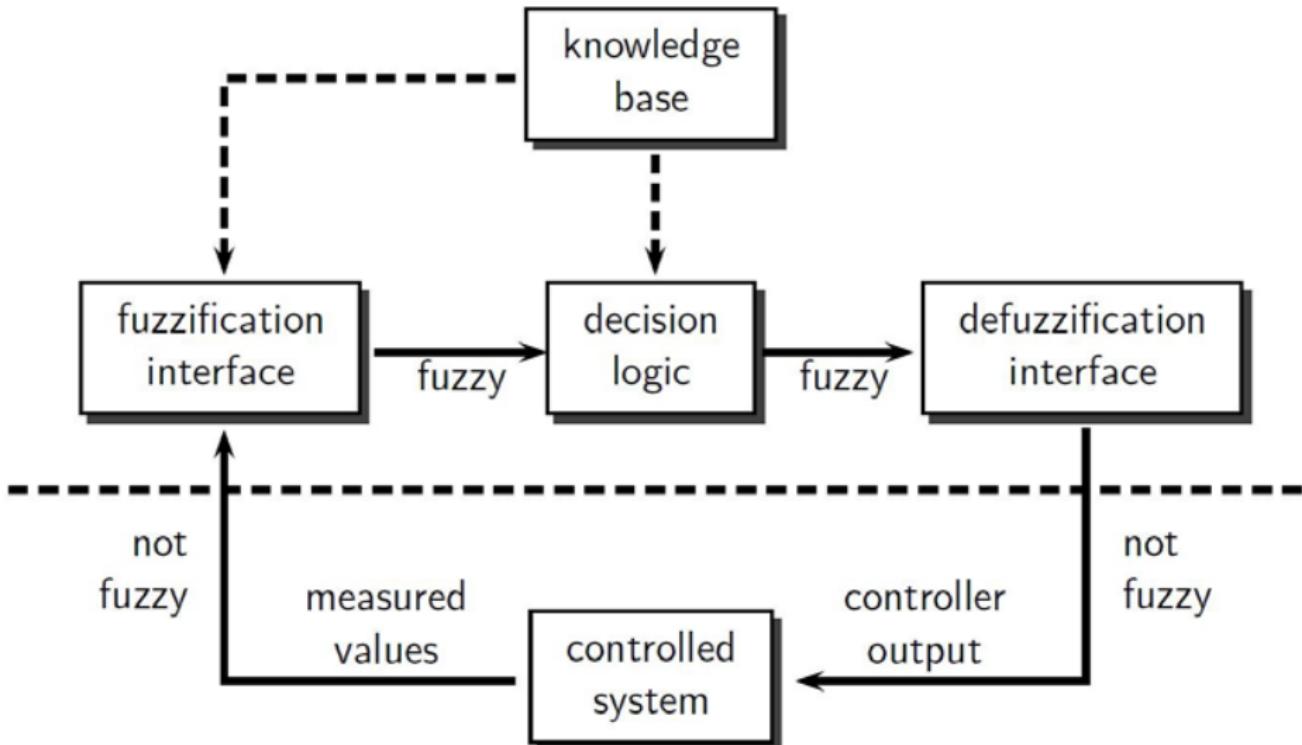
The decision logic combines the output fuzzy sets from all rules  $R_1, \dots, R_k$  by using the or-operator **maximum**. This results in the output fuzzy set

$$\mu_{x_1, \dots, x_n}^{\text{output}} : Y \rightarrow [0, 1]$$



Then  $\mu_{x_1, \dots, x_n}^{\text{output}}$  is passed to defuzzification interface.

# Architecture of a Fuzzy Controller

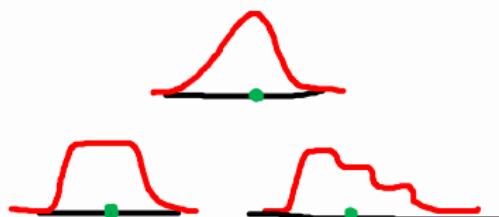


## Definition of Table-based Control Function IV

Defuzzification interface derives crisp value from  $\mu_{x_1, \dots, x_n}^{\text{output}}$ .

Most common **defuzzification** methods:

- max criterion,
- mean of maxima,
- center of gravity.



See Google Patents : More than 1000 methods in real applications



# Center of Gravity (COG) Method

$\eta$  = center of gravity/area of  $\mu_{x_1, \dots, x_n}^{\text{output}}$

If  $Y$  is finite, then

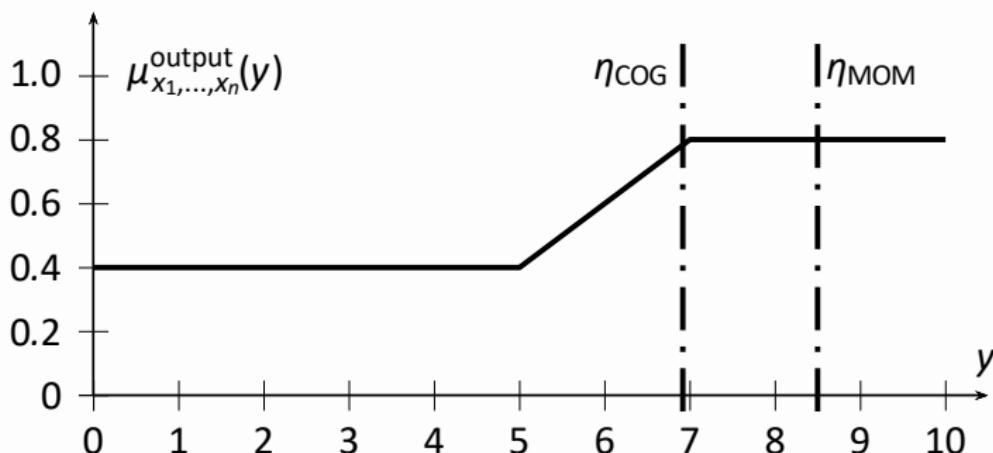
$$\eta = \frac{\sum_{y_i \in Y} y_i \cdot \mu_{x_1, \dots, x_n}^{\text{output}}(y_i)}{\sum_{y_i \in Y} \mu_{x_1, \dots, x_n}^{\text{output}}(y_i)}.$$

If  $Y$  is infinite, then

$$\eta = \frac{\int_{y \in Y} y \cdot \mu_{x_1, \dots, x_n}^{\text{output}}(y) dy}{\int_{y \in Y} \mu_{x_1, \dots, x_n}^{\text{output}}(y) dy}.$$

## Example

Task: compute  $\eta_{COG}$  and  $\eta_{MOM}$  of fuzzy set shown below.

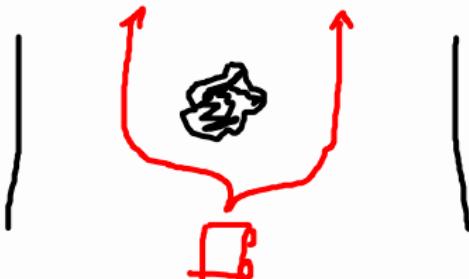
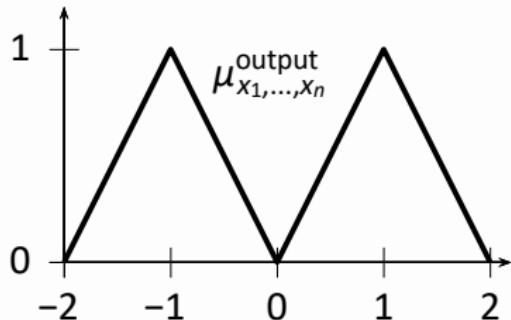




## Example for COG

$$\begin{aligned}\eta_{COG} &= \frac{\int_0^{10} y \cdot \mu_{x_1, \dots, x_n}^{\text{output}}(y) dy}{\int_0^{10} \mu_{x_1, \dots, x_n}^{\text{output}}(y) dy} \\ &= \frac{\int_0^5 0.4y dy + \int_5^7 (0.2y - 0.6)y dy + \int_7^{10} 0.8y dy}{5 \cdot 0.4 + 2 \cdot \frac{0.8+0.4}{2} + 3 \cdot 0.8} \\ &\approx \frac{38.7333}{5.6} \approx 6.917\end{aligned}$$

# Problem Cases for MOM and COG





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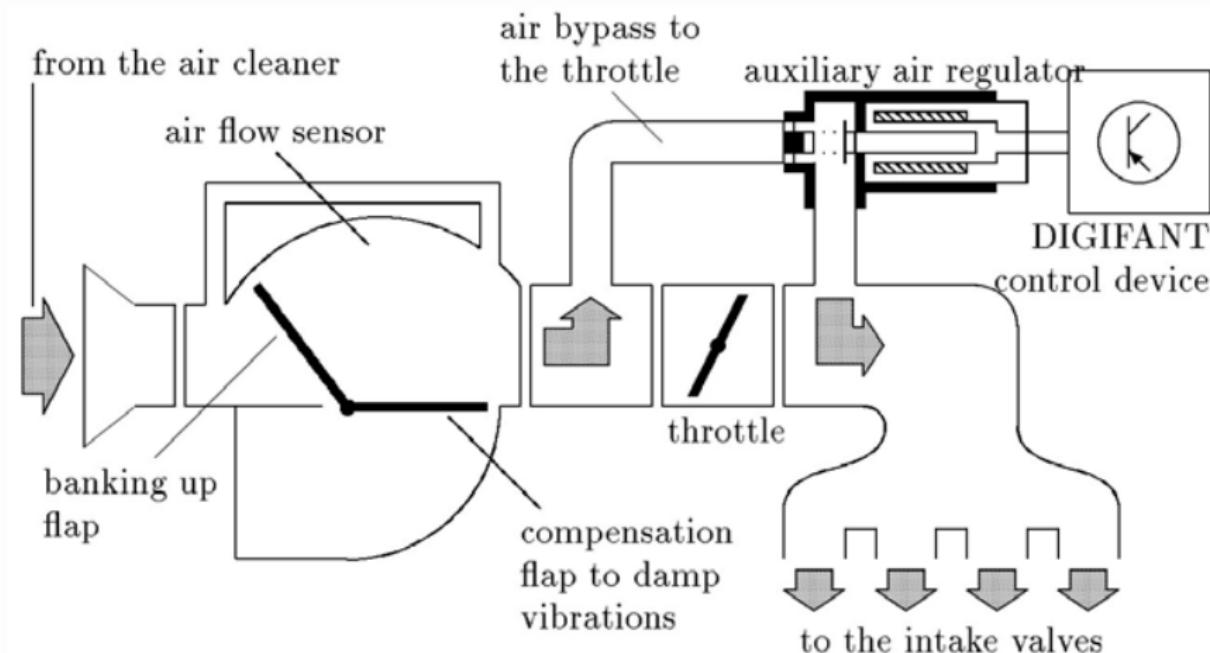
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# Mamdani Controller

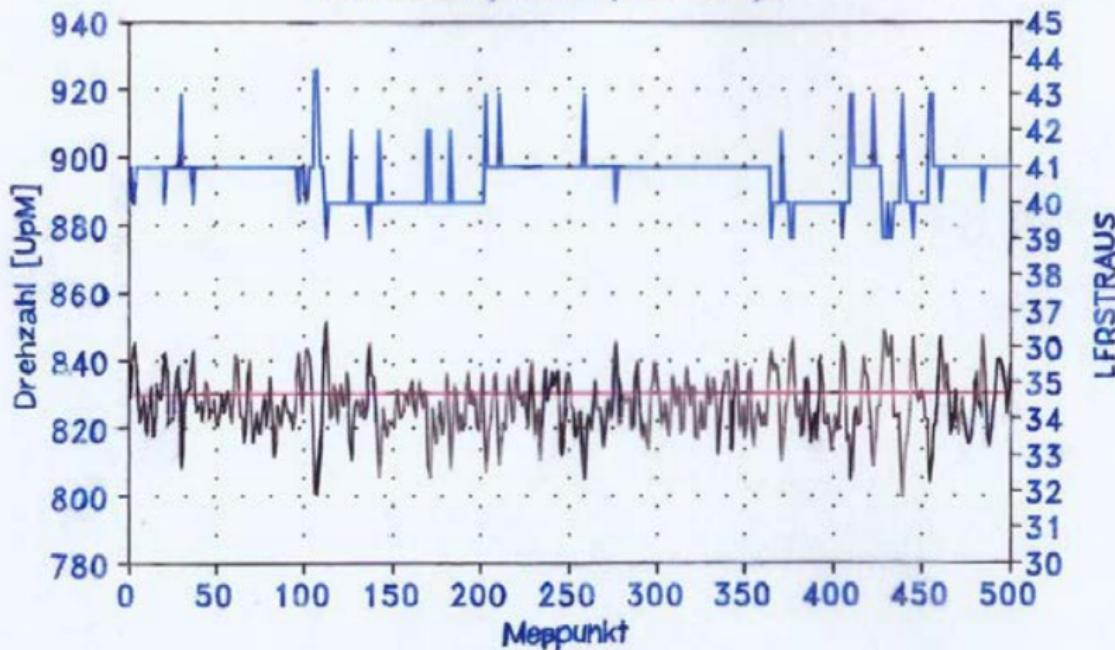
# Example: Engine Idle Speed Control

VW 2000cc 116hp Motor (Golf GTI)

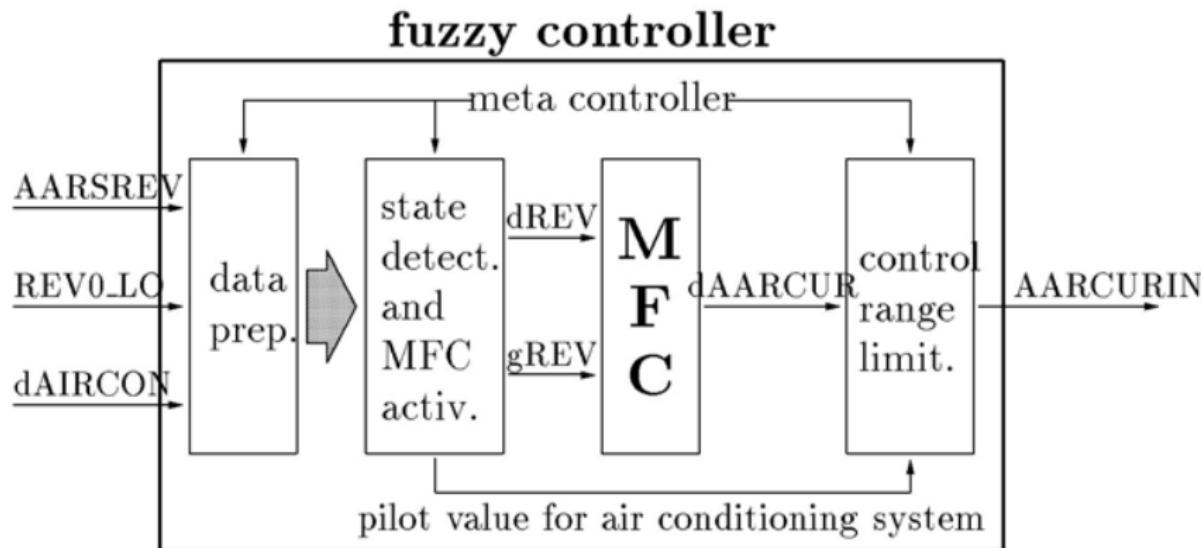




M163B1 Fz SR stat.Zust.  
20.8.92 MW(4500MP):827.13UpM

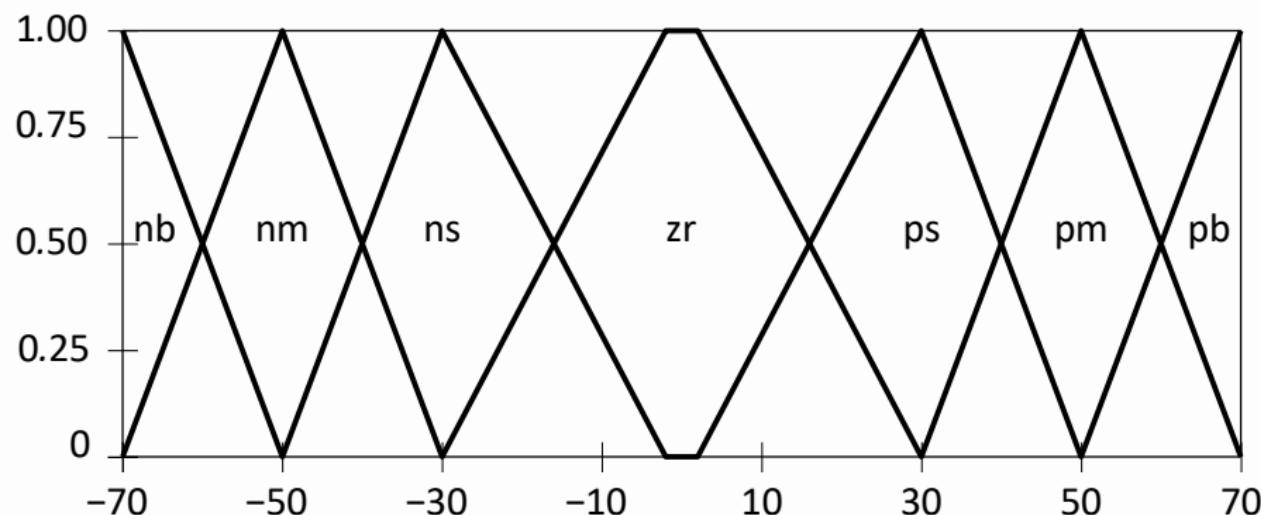


# Structure of the Fuzzy Controller



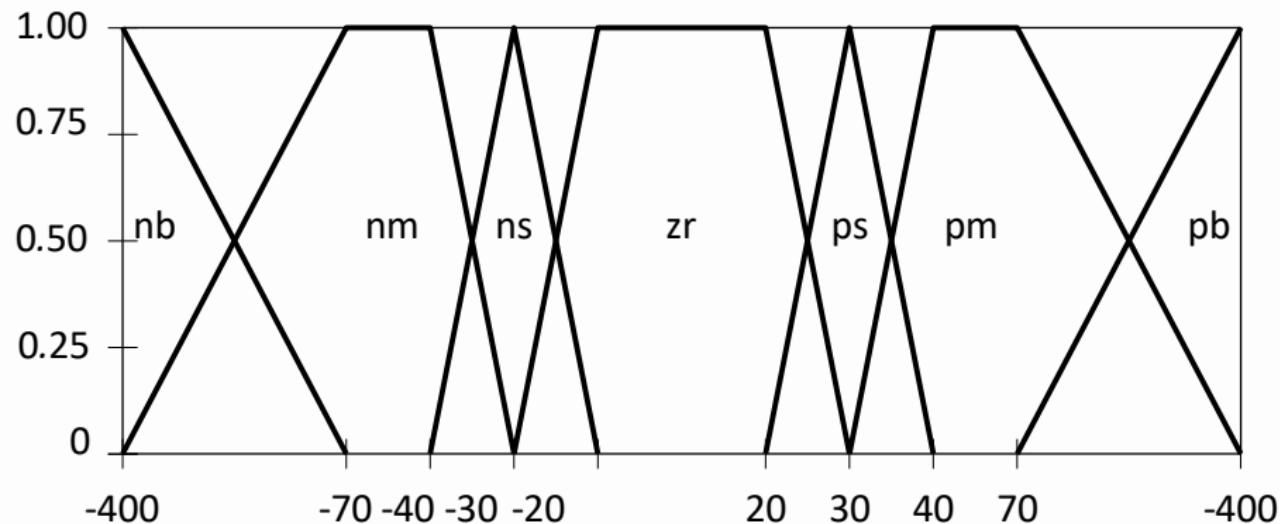
# Deviation of the Number of Revolutions

dREV



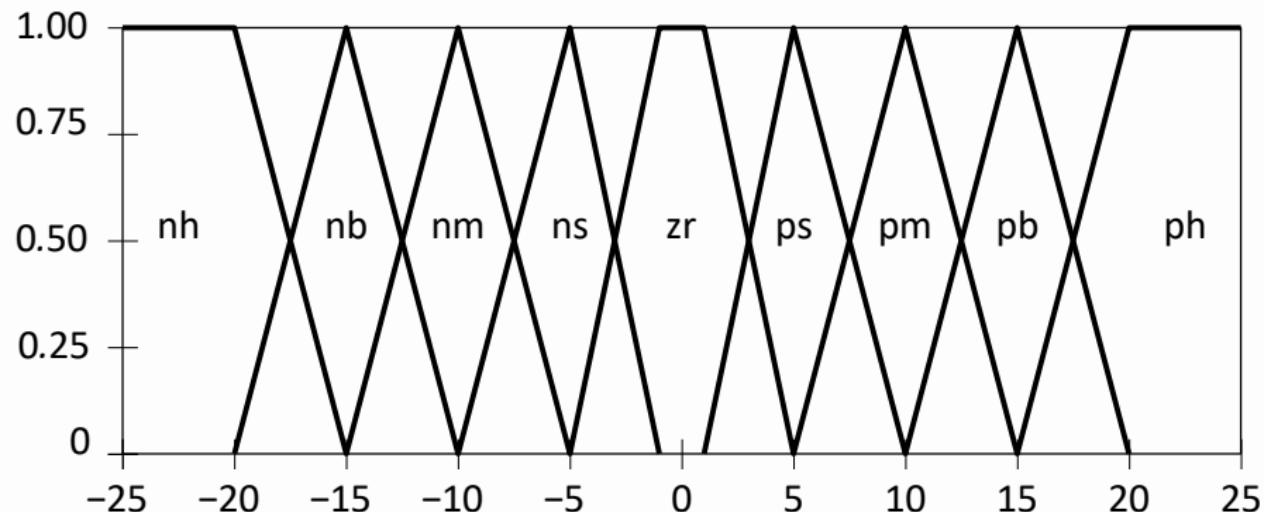
# Gradient of the Number of Revolutions

gREV



# Change of Current for Auxiliary Air Regulator

dAARCUR



## Rule Base

If the deviation from the desired number of revolutions is negative small **and** the gradient is negative medium,  
**then** the change of the current for the auxiliary air regulation should be positive medium.

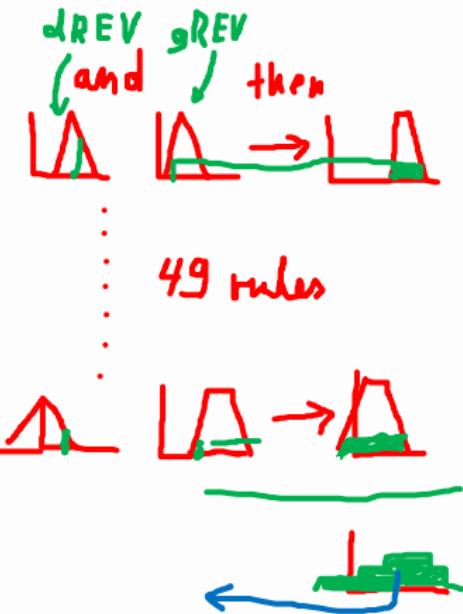
		gREV						
		nb	nm	ns	az	ps	pm	pb
dREV	nb	ph	pb	pb	pm	pm	ps	ps
	nm	ph	pb	pm	pm	ps	ps	az
	ns	pb	pm	ps	ps	az	az	az
	az	ps	ps	az	az	az	nm	ns
	ps	az	az	az	ns	ns	nm	nb
	pm	az	ns	ns	ns	nb	nb	nh
	pb	ns	ns	nm	nb	nb	nb	nh

## Rule Base

If the deviation from the desired number of revolutions is negative small **and** the gradient is negative medium,  
**then** the change of the current for the auxiliary air regulation should be positive medium.

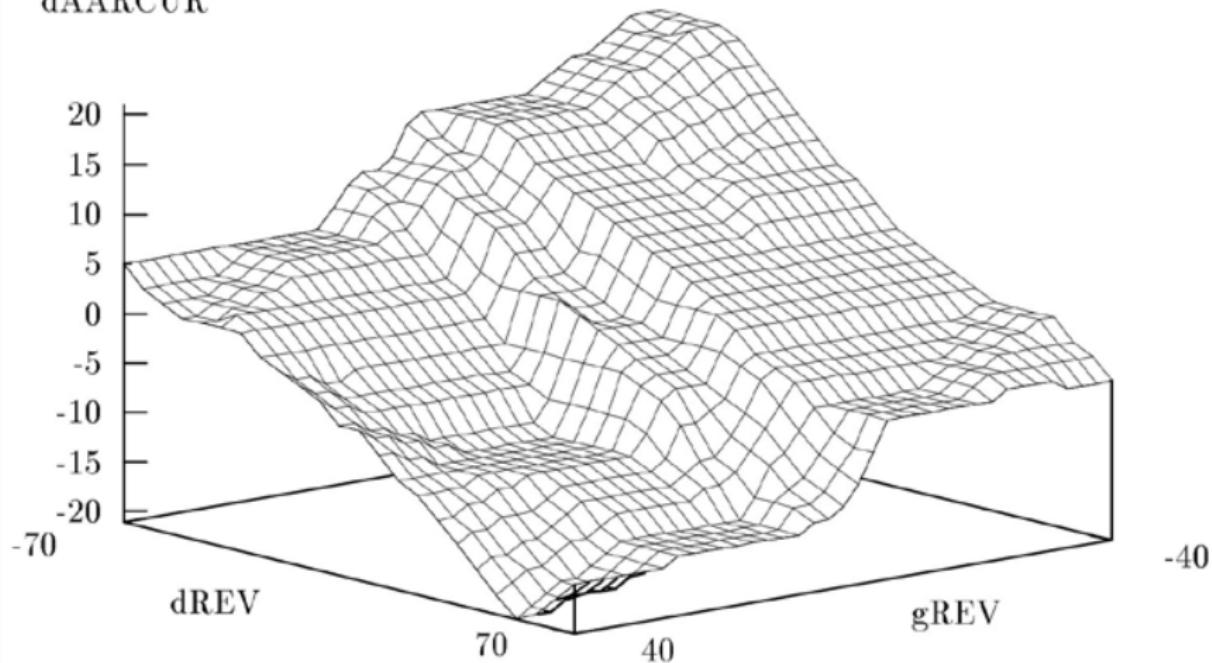
gREV

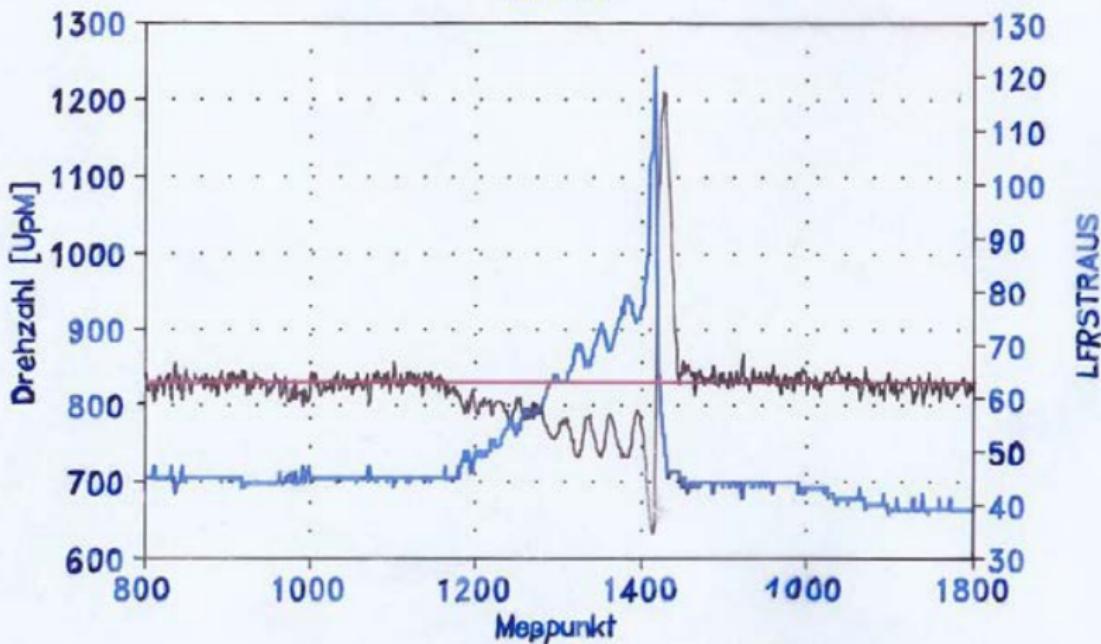
		gREV						
		nb	nm	ns	az	ps	pm	pb
dREV	nb	ph	pb	pb	pm	pm	ps	ps
	nm	ph	pb	pm	pm	ps	ps	az
	ns	pb	pm	ps	ps	az	az	az
	az	ps	ps	az	az	az	nm	ns
	ps	az	az	az	ns	ns	nm	nb
	pm	az	ns	ns	ns	nb	nb	nh
	pb	ns	ns	nm	nb	nb	nb	nh

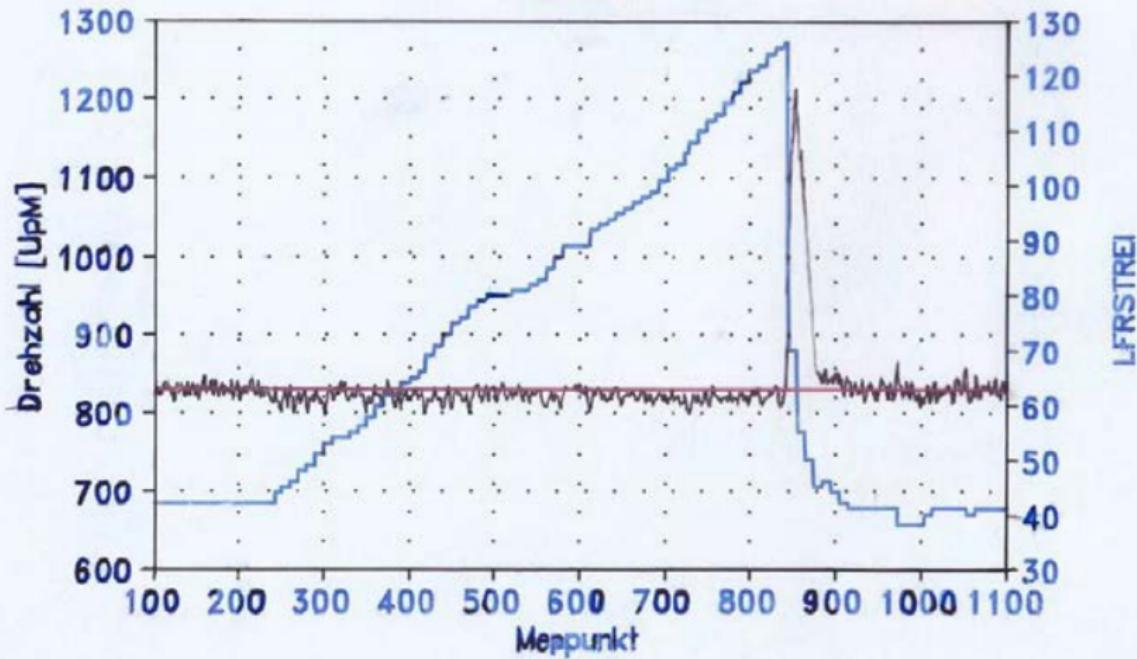


# Performance Characteristics

dAARCUR



M151B1 Fz SR Kupplung  
18.8.92

M155B2 Fz FC Kupplung  
18.8.92

— DRZO\_LO — LFRSTREI — LFRSDRZ



## Example: Automatic Gear Box I

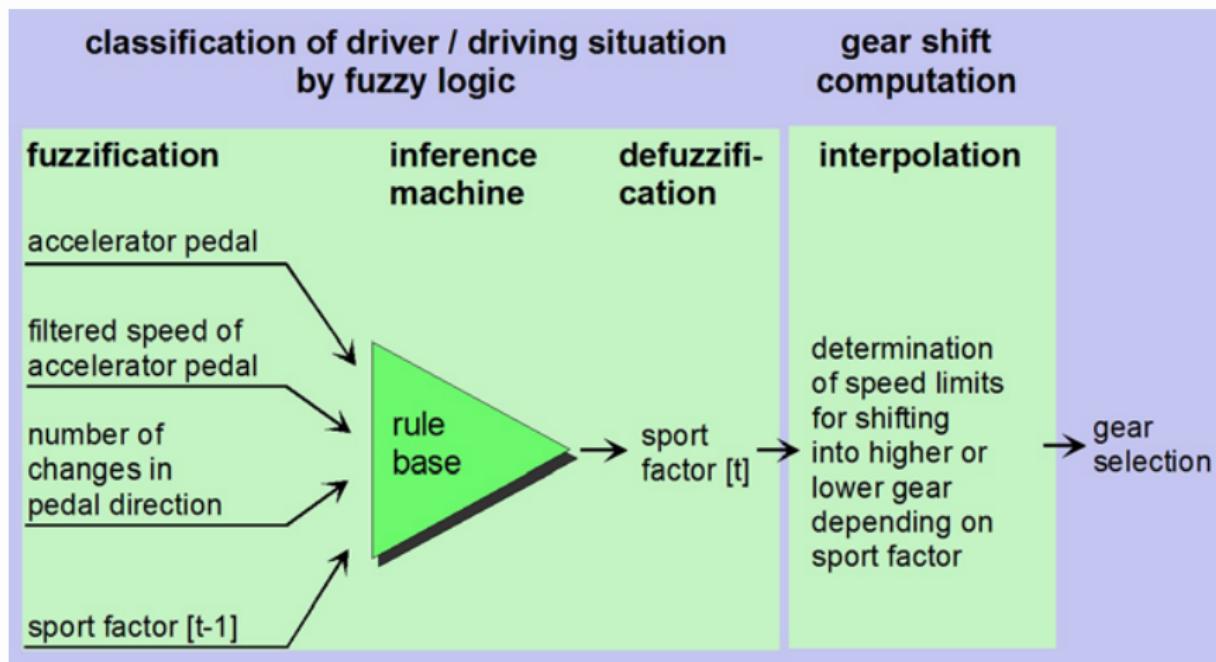
Idea: car “watches” driver and classifies him/her into calm, normal, sportive (assign sport factor [0, 1]), or nervous (calm down driver).

Test car: different drivers, classification by expert (passenger).

Simultaneous measurement of 14 attributes, *e.g.* , speed, position of accelerator pedal, speed of accelerator pedal, kick down, steering wheel angle.

# Example: Automatic Gear Box II

Continuously Adapting Gear Shift Schedule in VW New Beetle



# Example: Automatic Gear Box III

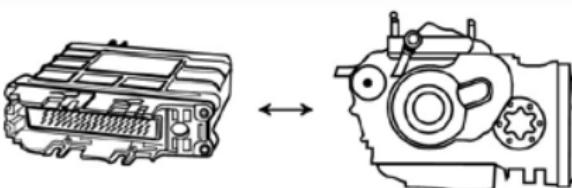
## Technical Details

Optimized program on Digimat:

24 byte RAM

702 byte ROM

uses 7 Mamdani fuzzy rules



Runtime: 80ms

12 times per second new sport factor is assigned.



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# Takagi Sugeno Control



# Takagi-Sugeno Controller

Proposed by Tomohiro Takagi and Michio Sugeno.

Modification/extension of Mamdani controller.

Both in common: fuzzy partitions of input domain  $X_1, \dots, X_n$ .

Difference to Mamdani controller:

- no fuzzy partition of output domain  $Y$ ,
- controller rules  $R_1, \dots, R_k$  are given by

$$R_r : \text{if } \xi_1 \text{ is } A_{i_{1,r}}^{(1)} \text{ and } \dots \text{ and } \xi_n \text{ is } A_{i_{n,r}}^{(n)} \\ \text{then } \eta_r = f_r(\xi_1, \dots, \xi_n),$$

$$f_r : X_1 \times \dots \times X_n \rightarrow Y.$$

- Generally,  $f_r$  is linear, i.e.  $f_r(x_1, \dots, x_n) = a_0^{(r)} + \sum_{i=1}^n a_i^{(r)} x_i$ .



# Takagi-Sugeno Controller

For given input  $(x_1, \dots, x_n)$  and for each  $R_r$ , decision logic computes truth value  $\alpha_r$  of each premise, and then  $f_r(x_1, \dots, x_n)$ .

Analogously to Mamdani controller:

$$\alpha_r = \min \left\{ \mu_{i_{1,r}}^{(1)}(x_1), \dots, \mu_{i_{n,r}}^{(n)}(x_n) \right\}.$$

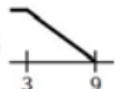
Output equals crisp control value

$$\eta = \frac{\sum_{r=1}^k \alpha_r \cdot f_r(x_1, \dots, x_n)}{\sum_{r=1}^k \alpha_r}.$$

Thus no defuzzification method necessary.

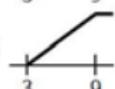
## Example

$R_1$  : if  $\xi_1$  is



$$\text{then } \eta_1 = 1 \cdot \xi_1 + 0.5 \cdot \xi_2 + 1$$

$R_2$  : if  $\xi_1$  is

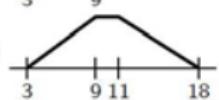


and  $\xi_2$  is



$$\text{then } \eta_2 = -0.1 \cdot \xi_1 + 4 \cdot \xi_2 + 1.2$$

$R_3$  : if  $\xi_1$  is

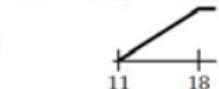


and  $\xi_2$  is



$$\text{then } \eta_3 = 0.9 \cdot \xi_1 + 0.7 \cdot \xi_2 + 9$$

$R_4$  : if  $\xi_1$  is



and  $\xi_2$  is



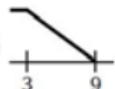
$$\text{then } \eta_4 = 0.2 \cdot \xi_1 + 0.1 \cdot \xi_2 + 0.2$$

If a certain clause " $x_j$  is  $A_{i_{j,r}}^{(j)}$ " in rule  $R_r$  is missing,  
then  $\mu_{i_{j,r}}(x_j) \equiv 1$  for all linguistic values  $i_{j,r}$ .

For instance, here  $x_2$  in  $R_1$ , so  $\mu_{i_{2,1}}(x_2) \equiv 1$  for all  $i_{2,1}$ .

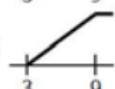
## Example

$R_1$  : if  $\xi_1$  is



$$\text{then } \eta_1 = 1 \cdot \xi_1 + 0.5 \cdot \xi_2 + 1$$

$R_2$  : if  $\xi_1$  is

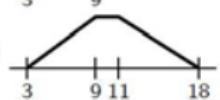


and  $\xi_2$  is



$$\text{then } \eta_2 = -0.1 \cdot \xi_1 + 4 \cdot \xi_2 + 1.2$$

$R_3$  : if  $\xi_1$  is

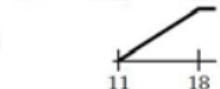


and  $\xi_2$  is

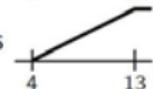


$$\text{then } \eta_3 = 0.9 \cdot \xi_1 + 0.7 \cdot \xi_2 + 9$$

$R_4$  : if  $\xi_1$  is



and  $\xi_2$  is



$$\text{then } \eta_4 = 0.2 \cdot \xi_1 + 0.1 \cdot \xi_2 + 0.2$$

If a certain clause " $x_j$  is  $A_{i_{j,r}}^{(j)}$ " in rule  $R_r$  is missing,  
then  $\mu_{i_{j,r}}(x_j) \equiv 1$  for all linguistic values  $i_{j,r}$ .

For instance, here  $x_2$  in  $R_1$ , so  $\mu_{i_{2,1}}(x_2) \equiv 1$  for all  $i_{2,1}$ .



## Example: Output Computation

input:  $(\xi_1, \xi_2) = (6, 7)$

$$\alpha_1 = 1/2 \wedge 1 = 1/2$$

$$\eta_1 = 6 + 7/2 + 1 = 10.5$$

$$\alpha_2 = 1/2 \wedge 2/3 = 1/2$$

$$\eta_2 = -0.6 + 28 + 1.2 = 28.6$$

$$\alpha_3 = 1/2 \wedge 1/3 = 1/3$$

$$\eta_3 = 0.9 \cdot 6 + 0.7 \cdot 7 + 9 = 19.3$$

$$\alpha_4 = 0 \wedge 1/3 = 0$$

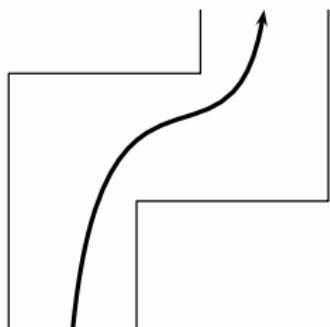
$$\eta_4 = 6 + 7/2 + 1 = 10.5$$

output:  $\eta = f(6, 7) =$

$$\frac{1/2 \cdot 10.5 + 1/2 \cdot 28.6 + 1/3 \cdot 19.3}{1/2 + 1/2 + 1/3} = 19.5$$



## Example: Passing a Bend



Pass a bend with a car at constant speed.

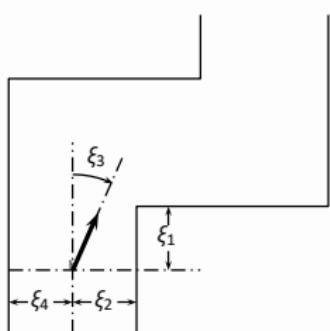
Measured inputs:

$\xi_1$ : distance of car to beginning of bend

$\xi_2$ : distance of car to inner barrier

$\xi_3$ : direction (angle) of car

$\xi_4$ : distance of car to outer barrier

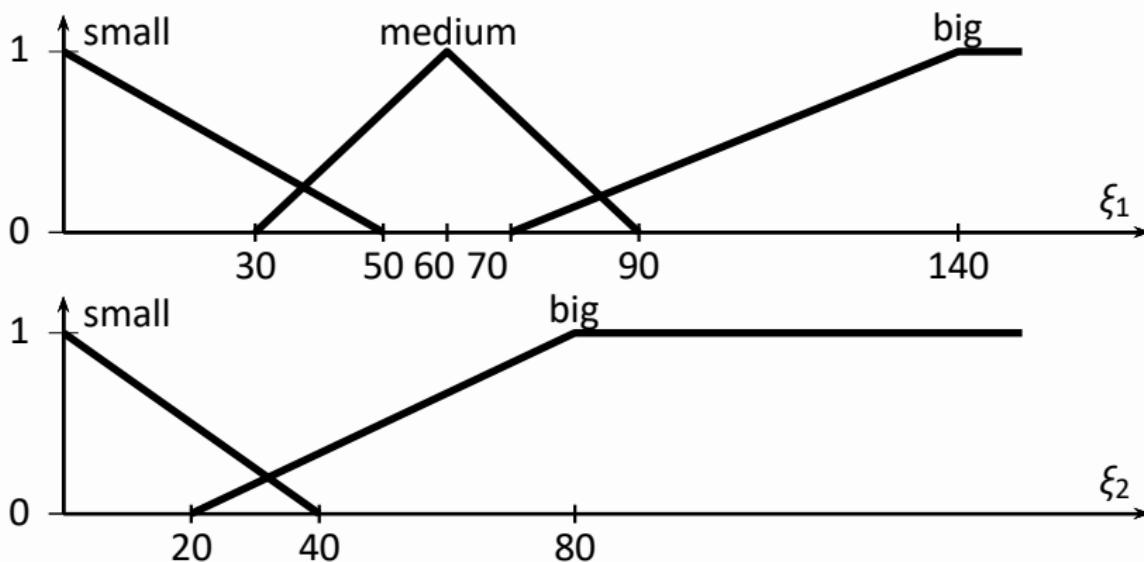


$\eta$  = rotation speed of steering wheel

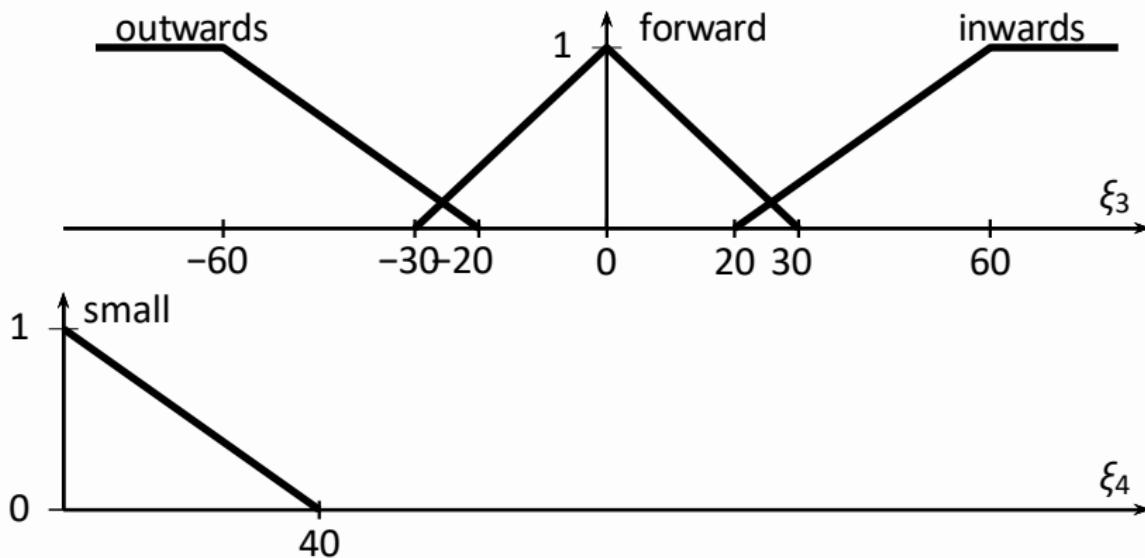
$X_1 = [0\text{cm}, 150\text{cm}], X_2 = [0\text{cm}, 150\text{cm}]$

$X_3 = [-90^\circ, 90^\circ], X_4 = [0\text{cm}, 150\text{cm}]$

## Fuzzy Partitions of $X_1$ and $X_2$



## Fuzzy Partitions of $X_3$ and $X_4$





## Rules for Car

$R_r : \text{if } \xi_1 \text{ is } A \text{ and } \xi_2 \text{ is } B \text{ and } \xi_3 \text{ is } C \text{ and } \xi_4 \text{ is } D$

$$\begin{aligned} \text{then } \eta = & p_0^{(A,B,C,D)} + p_1^{(A,B,C,D)} \cdot \xi_1 + p_2^{(A,B,C,D)} \cdot \xi_2 \\ & + p_3^{(A,B,C,D)} \cdot \xi_3 + p_4^{(A,B,C,D)} \cdot \xi_4 \end{aligned}$$

$$A \in \{\text{small, medium, big}\}$$

$$B \in \{\text{small, big}\}$$

$$C \in \{\text{outwards, forward, inwards}\}$$

$$D \in \{\text{small}\}$$

$$p_0^{(A,B,C,D)}, \dots, p_4^{(A,B,C,D)} \in \mathbb{R}$$



# Control Rules for the Car

rule	$\xi_1$	$\xi_2$	$\xi_3$	$\xi_4$	$p_0$	$p_1$	$p_2$	$p_3$	$p_4$
$R_1$	-	-	outwards	small	3.000	0.000	0.000	-0.045	-0.004
$R_2$	-	-	forward	small	3.000	0.000	0.000	-0.030	-0.090
$R_3$	small	small	outwards	-	3.000	-0.041	0.004	0.000	0.000
$R_4$	small	small	forward	-	0.303	-0.026	0.061	-0.050	0.000
$R_5$	small	small	inwards	-	0.000	-0.025	0.070	-0.075	0.000
$R_6$	small	big	outwards	-	3.000	-0.066	0.000	-0.034	0.000
$R_7$	small	big	forward	-	2.990	-0.017	0.000	-0.021	0.000
$R_8$	small	big	inwards	-	1.500	0.025	0.000	-0.050	0.000
$R_9$	medium	small	outwards	-	3.000	-0.017	0.005	-0.036	0.000
$R_{10}$	medium	small	forward	-	0.053	-0.038	0.080	-0.034	0.000
$R_{11}$	medium	small	inwards	-	-1.220	-0.016	0.047	-0.018	0.000
$R_{12}$	medium	big	outwards	-	3.000	-0.027	0.000	-0.044	0.000
$R_{13}$	medium	big	forward	-	7.000	-0.049	0.000	-0.041	0.000
$R_{14}$	medium	big	inwards	-	4.000	-0.025	0.000	-0.100	0.000
$R_{15}$	big	small	outwards	-	0.370	0.000	0.000	-0.007	0.000
$R_{16}$	big	small	forward	-	-0.900	0.000	0.034	-0.030	0.000
$R_{17}$	big	small	inwards	-	-1.500	0.000	0.005	-0.100	0.000
$R_{18}$	big	big	outwards	-	1.000	0.000	0.000	-0.013	0.000
$R_{19}$	big	big	forward	-	0.000	0.000	0.000	-0.006	0.000
$R_{20}$	big	big	inwards	-	0.000	0.000	0.000	-0.010	0.000



# Sample Calculation

Assume that the car is 10 cm away from beginning of bend ( $\xi_1 = 10$ ).

The distance of the car to the inner barrier be 30 cm ( $\xi_2 = 30$ ).

The distance of the car to the outer barrier be 50 cm ( $\xi_4 = 50$ ).

The direction of the car be “forward” ( $\xi_3 = 0$ ).

Then according to all rules  $R_1, \dots, R_{20}$ ,  
only premises of  $R_4$  and  $R_7$  have a value  $\neq 0$ .



## Membership Degrees to Control Car

	small	medium	big
$\xi_1 = 10$	0.8	0	0

	small	big
$\xi_2 = 30$	0.25	0.167

	outwards	forward	inwards
$\xi_3 = 0$	0	1	0

	small
$\xi_4 = 50$	0



## Sample Calculation (cont.)

For the premise of  $R_4$  and  $R_7$ ,  $\alpha_4 = 1/4$  and  $\alpha_7 = 1/6$ ,

$R_4$  yields

$$\begin{aligned}\eta_4 &= 0.303 - 0.026 \cdot 10 + 0.061 \cdot 30 - 0.050 \cdot 0 + 0.000 \cdot 50 \\ &= 1.873.\end{aligned}$$

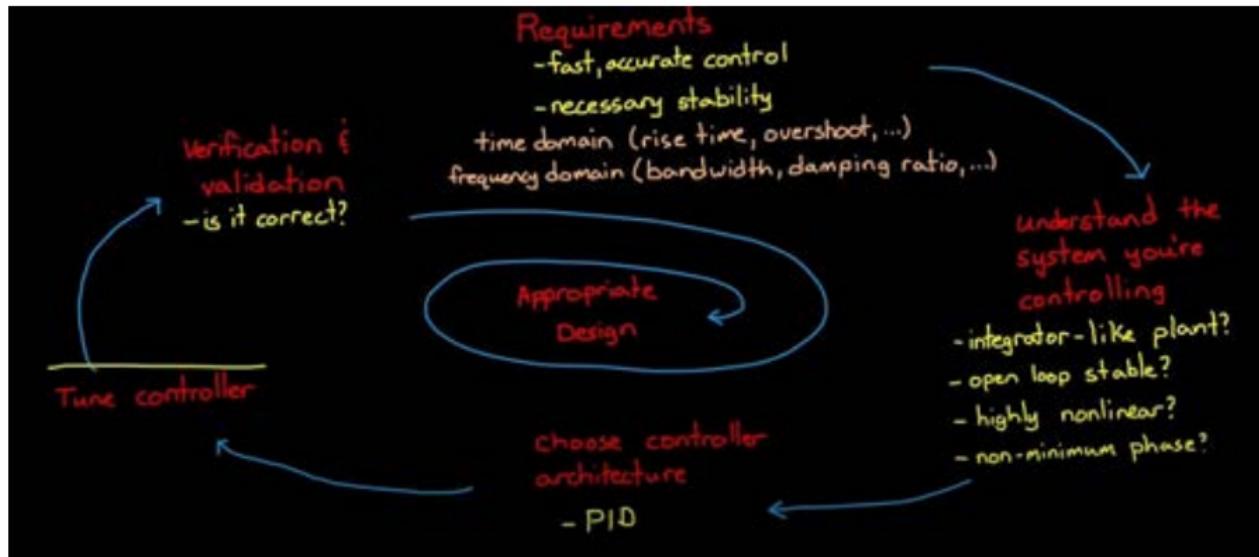
$R_7$  yields

$$\begin{aligned}\eta_7 &= 2.990 - 0.017 \cdot 10 + 0.000 \cdot 30 - 0.021 \cdot 0 + 0.000 \cdot 50 \\ &= 2.820.\end{aligned}$$

The final value for control variable is thus

$$\eta =$$

$$2.2518.$$



## Control System Design



# References

