



# Modul: Issues in Decision Tree Learning (DTL)

### **Continuous-valued Attribute**

Pembelajaran Mesin (Machine Learning)

#### **Nur ULFA Maulidevi**

KK IF - Teknik Informatika- STEI ITB



#### **Issues in DTL**

Overfitting training data

Continuous
-valued
attribute

Handling attributes with differing costs

Handling missing attribute value

Alternative measures for selecting attributes

#### Discretization

Continuous valued attributes → new discrete valued (boolean) attribute A



True: A < c



False:  $A < c \text{ (or } A \ge c)$ 

Temperature:	40	48	60	72	80	90
Play Tennis:	No	No	Yes	Yes	Yes	No

Potential optimal breakpoints

What is Best Value

for threshold c?

C = (48+60)/2 = 54C = (80+90)/2 = 85

**Use Information Gain for** each potential breakproint



#### Illustration

1. Sort The Continuous-valued attribute

Day	Outlook	Temperature	Humidity	Wind	Play Tennis
D1		72			Yes
D <sub>2</sub>		40			No
D <sub>3</sub>		90			No
D <sub>4</sub>		60			Yes
D <sub>5</sub>		48			No
D6		80			Yes

-	Temperature	40	48	60	72	80	90
	Play Tennis	No	No	Yes	Yes	Yes	No

2. Identify Adjacent examples that differ in their target class

3. Candidates: midway between corresponding values  $\rightarrow$  C: 54 or C: 85

4. Find the greatest Gain from the candidates, and other discrete-valued attributes

For C: 54

Temperature < 54: 2 examples  $\rightarrow$  yes/o, no/2 Temperature  $\geq$  54: 4 examples  $\rightarrow$  yes/3, no/1

 $Gain(S,T_{54}) = Entropi(S) - [(2/6*Entropi(0,2))+(4/6*Entropi(3,1)]$ 

For C: 85

Temperature < 85: 5 examples  $\rightarrow$  yes/3, no/2 Temperature  $\geq$  85: 1 examples  $\rightarrow$  yes/0, no/1 Gain(S,T<sub>85</sub>) = Entropi(S) – [(5/6\*Entropi(3,2))+(1/6\*Entropi(0,1)]

## **THANK YOU**





