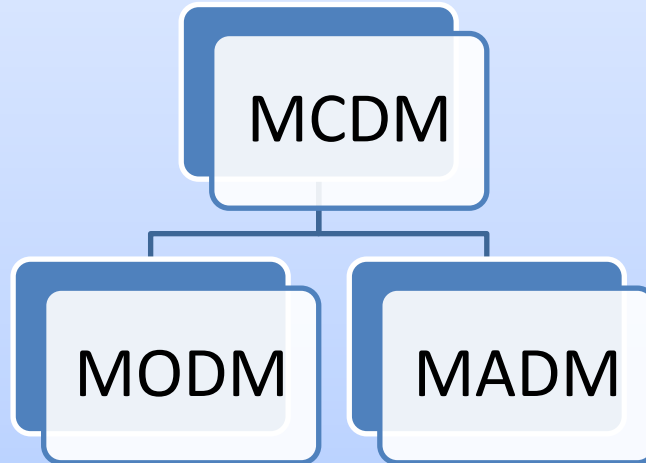


# Multi Criteria Decision Making (MCDM)

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# MCDM

- MCDM deals with decision problems under the presence of a number of **decision criteria**



- MODM: Multi-Objective Decision Making
- MADM: Multi-Attribute Decision Making

# Alternatives

- Represent the different **choices of action** available to the decision maker.
- Usually, the set of alternatives is assumed to be **finite**
- They are supposed to be **screened, prioritized** and eventually **ranked**.



# Multiple attributes

- Each MADM problem is associated with multiple attributes.
- Attributes are also referred to as *"goals"* or *"decision criteria"*.
- Attributes represent the different **dimensions** from which the alternatives can be viewed

# Examples

## Small Group:

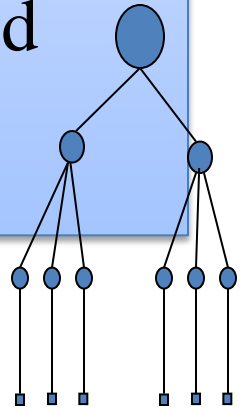
- Take the next 5 minutes to discuss in your group about attributes of a car

## Class

- Discuss findings

# Multiple attributes

- In cases in which the **number of attributes** is **large** (e.g., more than a few dozens), attributes may be arranged in a **hierarchical manner**.
  - Some attributes may be **major** attributes.
  - Each major attribute may be associated with several **sub-attributes**.
  - Similarly, each sub-attribute may be associated with several **sub-sub-attributes** and so on.



# Examples

## Small Group:

- Now, take the next 5 minutes to arrange all findings from previous activity in a hierarchy manner

## Class

- Discuss new findings

# Conflict among attributes

- Since different attributes represent different dimensions of the alternatives, they may **conflict** with each other.

*Eg. Cost vs. benefit*



# Problems in doing trade-off

- Criteria have **different units**
- Some are **quantitative**, some are **qualitative**
- For some, **smaller is better** (resource use), for others, **bigger is better** (quality issues)

How to picture diverse criteria together so we can decide between alternatives?

# Examples

- Car buying
- Job selection
- Human resource planning
- Computer buying

# Classification of MCDM Methods

- Based on data:
  - Deterministic
  - Stochastic
  - Fuzzy
- Based on number of decision makers:
  - Single
  - Group

single decision maker deterministic  
MADM methods

# Single decision maker deterministic MADM methods

- WSM: Weighted Sum Model
- AHP: Analytical Hierarchy Process

# Weighted Sum Model (WSM)

- **Step 1:** Define all factors or attributes
- **Step 2:** Assign weight to each factor
- **Step 3:** List all non-dominated alternatives
- **Step 4:** Evaluate each alternative based on each factor using scale of (0, 1); (1, 10); or (1, 100)
- **Step 5:** Calculate weighted sum for each alternative
- **Step 6:** Compare and select the best alternative

# Example

A just graduated student consider for a job

- **Step 1: Three important factors:**
  - Salary
  - Promotion
  - Company location

# Ex.

- Step 2: Assign weight
  - Salary 0.5
  - Promotion 0.3
  - Company location 0.2

# Ex.

- Step 3: Alternatives
  - Company A, B and C
- Step 4: Evaluate

Factor	A	B	C
Salary	0.8	0.4	0.7
Promotion	0.3	0.9	0.4
Location	0.6	0.6	0.2



# Ex.

- Step 5: Calculate weight sum for each alternative

Factor	Weight	A	B	C
Salary	0.5	0.8	0.4	0.7
Promotion	0.3	0.3	0.9	0.4
Location	0.2	0.6	0.6	0.2
Weighted sum		<b>0.61</b>	<b>0.59</b>	<b>0.56</b>

# EX.

- Step 6: Decision making  
Select company A

# Activity

Small  
group

- Take the next 5 minutes to discuss all advantages and disadvantages of this method

Class

- Class discussion

# The Analytic Hierarchy Process (AHP)

- Founded by **Saaty** in 1980.
- It is a popular and widely used method for MCDM.
- Allows the use of **qualitative**, as well as **quantitative** criteria in evaluation.
- Wide range of **applications** exists:
  - Selecting a car for purchasing
  - Deciding upon a place to visit for vacation
  - Deciding upon an MBA program after graduation.

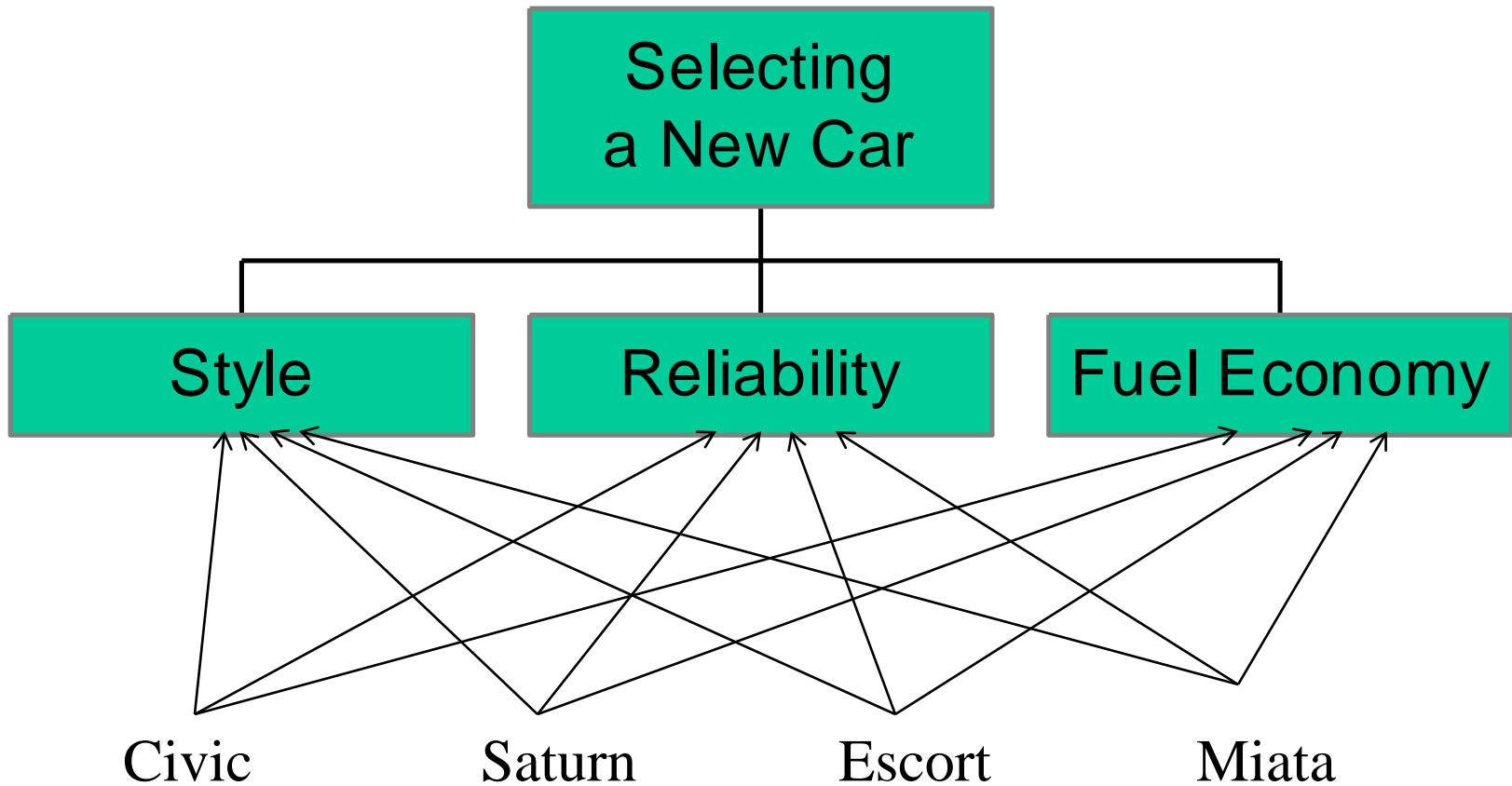
# AHP-General Idea

- Develop an **hierarchy of decision criteria** and define the **alternative courses of actions**.
- AHP algorithm is basically composed of two steps:
  1. Determine the relative **weights** of the decision criteria
  2. Determine the relative **rankings** (priorities) of alternatives

# Example: Car Selection

- Objective
  - Selecting a car
- Criteria
  - Style, Reliability, Fuel-economy
- Alternatives
  - Civic Coupe, Saturn Coupe, Ford Escort, Mazda Miata

# Hierarchy tree



Alternative courses of action

# Ranking of Criteria and Alternatives

- **Pairwise comparisons** are made with the grades ranging from 1-9.
- A basic, but very reasonable assumption for comparing alternatives:  
*If attribute A is absolutely more important than attribute B and is rated at 9, then B must be absolutely less important than A and is graded as 1/9.*
- These pairwise comparisons are carried out for all factors to be considered, usually not more than 7, and the matrix is completed.



# 1 -9 Scale

## Intensity of Importance

## Definition

1

**Equal Importance**

3

**Moderate Importance**

5

**Strong Importance**

7

**Very Strong Importance**

9

**Extreme Importance**

**2, 4, 6, 8**

**For compromises between the above**

**Reciprocals of above**

**In comparing elements i and j**

**- if i is 3 compared to j**

**- then j is 1/3 compared to i**

**Rationals**

**Force consistency**

**Measured values available**

# Ranking of criteria

	Style	Reliability	Fuel Economy
Style	1	1/2	3
Reliability	2	1	4
Fuel Economy	1/3	1/4	1

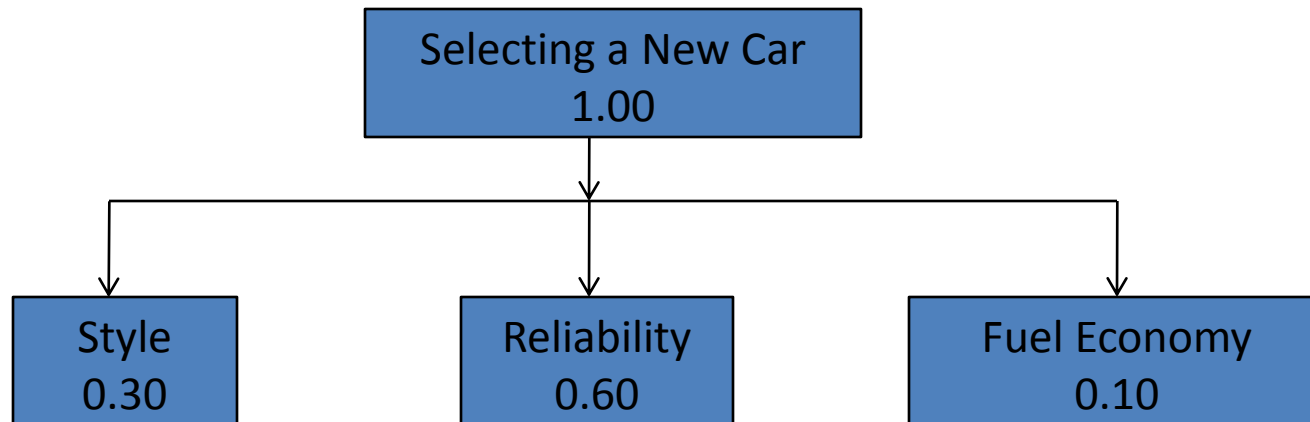
# Ranking of priorities

- Consider  $[Ax = \lambda_{\max} X]$  where
  - $A$  is the comparison matrix of size  $n \times n$ , for  $n$  criteria, also called the priority matrix.
  - $x$  is the Eigenvector of size  $n \times 1$ , also called the **priority vector**.
  - $\lambda_{\max}$  is the **Eigenvalue**,  $\lambda_{\max} \in \mathbb{R} > n$ .
- To find the ranking of priorities, namely the Eigen Vector  $X$ :
  - 1) Normalize the column entries by **dividing each entry by the sum** of the column.
  - 2) Take the overall row **averages**.

$$\begin{array}{c}
 A = \begin{bmatrix} 1 & 0.5 & 3 \\ 2 & 1 & 4 \\ 0.33 & 0.25 & 1.0 \end{bmatrix} \xrightarrow{\text{Normalized Column Sums}} \begin{bmatrix} 0.30 & 0.29 & 0.38 \\ 0.60 & 0.57 & 0.50 \\ 0.10 & 0.14 & 0.13 \end{bmatrix} \xrightarrow{\text{Row averages}} X = \begin{bmatrix} 0.30 \\ 0.60 \\ 0.10 \end{bmatrix} \\
 \text{Column sums } 3.33 \quad 1.75 \quad 8.00 \qquad \qquad \qquad 1.00 \quad 1.00 \quad 1.00 \qquad \qquad \qquad \text{Priority vector}
 \end{array}$$

## Criteria weights

- Style .30
- Reliability .60
- Fuel Economy .10



# Checking for Consistency

- The next stage is to calculate a **Consistency Ratio (CR)** to measure how consistent the judgments have been relative to large samples of purely random judgments.
- AHP evaluations are based on the assumption that the decision maker is rational, i.e., if A is preferred to B and B is preferred to C, then A is preferred to C.
- If the CR is greater than 0.1 the judgments are untrustworthy because they are too close for comfort to randomness and the exercise is valueless or must be repeated.

# Calculation of Consistency Ratio

- The next stage is to calculate  $\lambda_{\max}$  so as to lead to the **Consistency Index** and the **Consistency Ratio**.
- Consider  $[Ax = \lambda_{\max} x]$  where  $x$  is the Eigenvector.

$$\begin{array}{c} A \\ \left[ \begin{array}{ccc} 1 & 0.5 & 3 \\ 2 & 1 & 4 \\ 0.333 & 0.25 & 1.0 \end{array} \right] \end{array} \begin{array}{c} x \\ \left[ \begin{array}{c} 0.30 \\ 0.60 \\ 0.10 \end{array} \right] \end{array} = \begin{array}{c} Ax \\ \left[ \begin{array}{c} 0.90 \\ 1.60 \\ 0.35 \end{array} \right] \end{array} = \begin{array}{c} x \\ \left[ \begin{array}{c} 0.30 \\ 0.60 \\ 0.10 \end{array} \right] \end{array} = \lambda_{\max} \begin{array}{c} x \\ \left[ \begin{array}{c} 0.30 \\ 0.60 \\ 0.10 \end{array} \right] \end{array}$$

$$\lambda_{\max} = \text{average}\{0.90/0.30, 1.60/0.6, 0.35/0.10\} = 3.06$$

■ **Consistency index , CI** is found by

$$CI = (\lambda_{\max} - n) / (n - 1) = (3.06 - 3) / (3 - 1) = 0.03$$

# Consistency Ratio

- The final step is to **calculate the Consistency Ratio, CR** by using the table below, derived from Saaty's book.
- The upper row is the **order of the random matrix**, and the lower row is the corresponding index of consistency for random judgments (**Random index**).

1	2	3	4	5	6	7
0.00	0.00	0.58	0.90	1.12	1.24	1.32

8	9	10	11	12	13	14	15
1.41	1.45	1.49	1.51	1.48	1.56	1.57	1.59

# Consistency Ratio

An inconsistency of **10%** or less implies that the adjustment is small as compared to the actual values of the eigenvector entries.

A CR as high as, say, 90% would mean that the pairwise judgments are just about random and are completely **untrustworthy**! In this case, comparisons should be repeated.

In the above example:  $CR = CI / 0.58 = 0.03 / 0.58 = 0.05$   
 **$0.05 < 0.1$ , so the evaluations are consistent!**



# Ranking alternatives

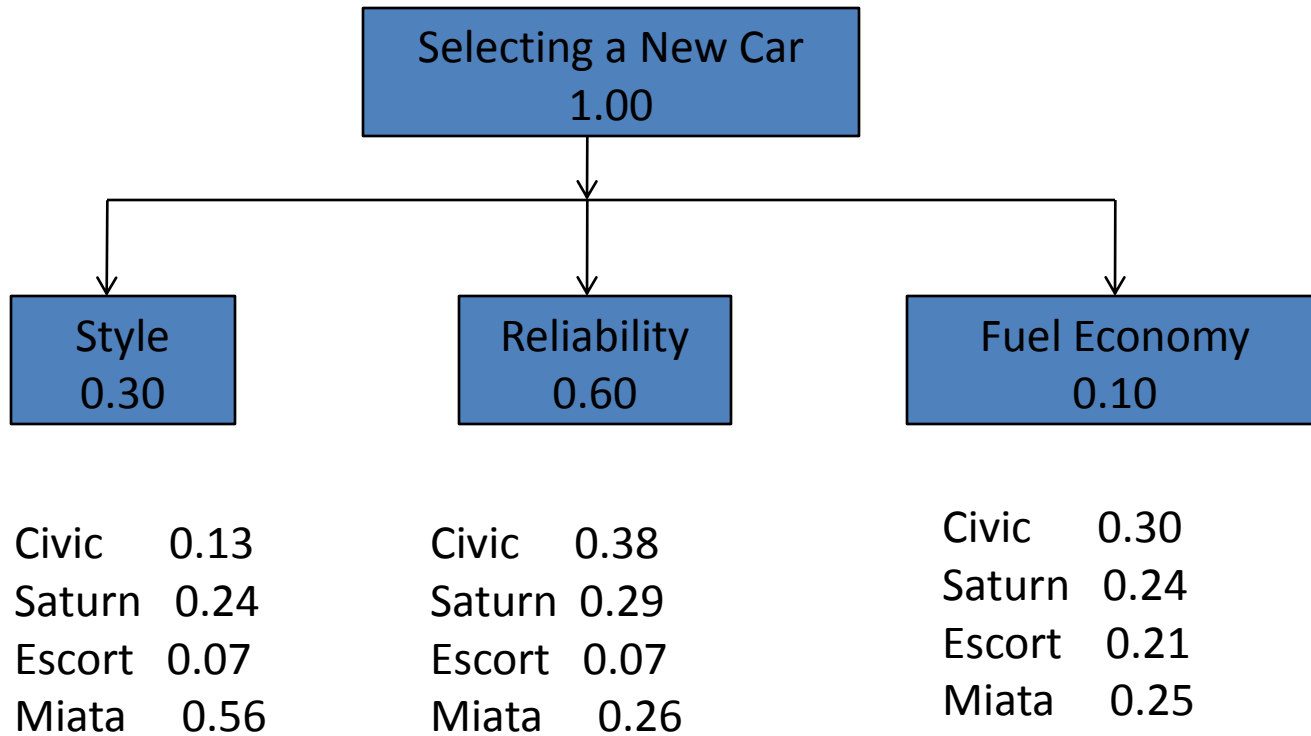
<u>Style</u>	Civic	Saturn	Escort	Miata	<u>Priority vector</u>
Civic	1	1/4	4	1/6	$\begin{bmatrix} 0.13 \\ 0.24 \\ 0.07 \\ 0.56 \end{bmatrix}$
Saturn	4	1	4	1/4	
Escort	1/4	1/4	1	1/5	
Miata	6	4	5	1	

<u>Reliability</u>	Civic	Saturn	Escort	Miata	
Civic	1	2	5	1	$\begin{bmatrix} 0.38 \\ 0.29 \\ 0.07 \\ 0.26 \end{bmatrix}$
Saturn	1/2	1	3	2	
Escort	1/5	1/3	1	1/4	
Miata	1	1/2	4	1	

# Ranking alternatives

		<u>Miles/gallon</u>	<u>Normalized</u>
<u>Fuel Economy</u>	Civic	34	.30
	Saturn	27	.24
	Escort	24	.21
	Miata	<u>28</u>	<u>.25</u>
		113	1.0

! Since fuel economy is a **quantitative** measure, fuel consumption ratios can be used to determine the relative ranking of alternatives; however this is not obligatory. Pairwise comparisons may still be used in some cases.



# Ranking of alternatives

	Style	Reliability	Fuel Economy			
<b>Civic</b>	$\begin{bmatrix} .13 & .38 & .30 \\ .24 & .29 & .24 \\ .07 & .07 & .21 \\ .56 & .26 & .25 \end{bmatrix}$	$\begin{bmatrix} .30 \\ .60 \\ .10 \end{bmatrix}$	$\begin{bmatrix} .30 \\ .27 \\ .08 \\ .35 \end{bmatrix}$	$\times$	$=$	
<b>Saturn</b>						
<b>Escort</b>						
<b>Miata</b>						

↓

Priority matrix

↓

Criteria Weights

**Winner**

**The Miata is the highest ranked car**

In summary AHP provides a logical framework to determine benefits of each alternatives

1. Miata	0.35
2. Civic	0.30
3. Saturn	0.27
4. Escort	0.08

WHAT ABOUT COSTS?



**ALTHOUGH COSTS COULD HAVE BEEN INCLUDED, IN MANY COMPLEX DECISIONS, **COSTS** SHOULD BE SET ASIDE UNTIL THE BENEFITS OF THE ALTERNATIVES ARE EVALUATED**

**OTHERWISE THIS COULD HAPPEN...**

**YOUR PROGRAM COST TOO MUCH I  
DON'T CARE ABOUT ITS BENEFITS**

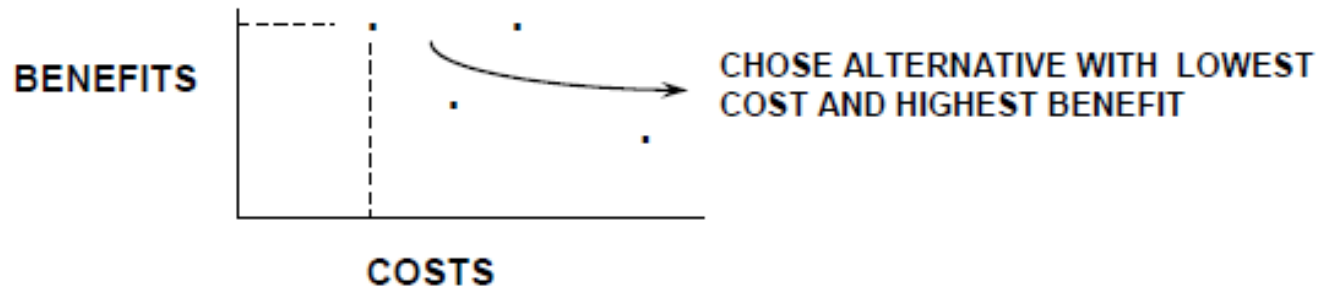
**DISCUSSING COSTS  
TOGETHER WITH BENEFITS  
CAN SOMETIMES BRING FORTH  
MANY POLITICAL AND  
EMOTIONAL RESPONSES**



**WAYS TO HANDLE BENEFITS  
AND COSTS INCLUDE THE  
FOLLOWING:**



**1. GRAPHING BENEFITS AND COSTS OF EACH ALTERNATIVE**



**2. BENEFIT TO COST RATIOS**

**3. LINEAR PROGRAMMING**

**4. SEPARATE BENEFIT AND COST HIERARCHICAL TREES  
AND THEN COMBINE THE RESULTS**

**IN OUR EXAMPLE...**

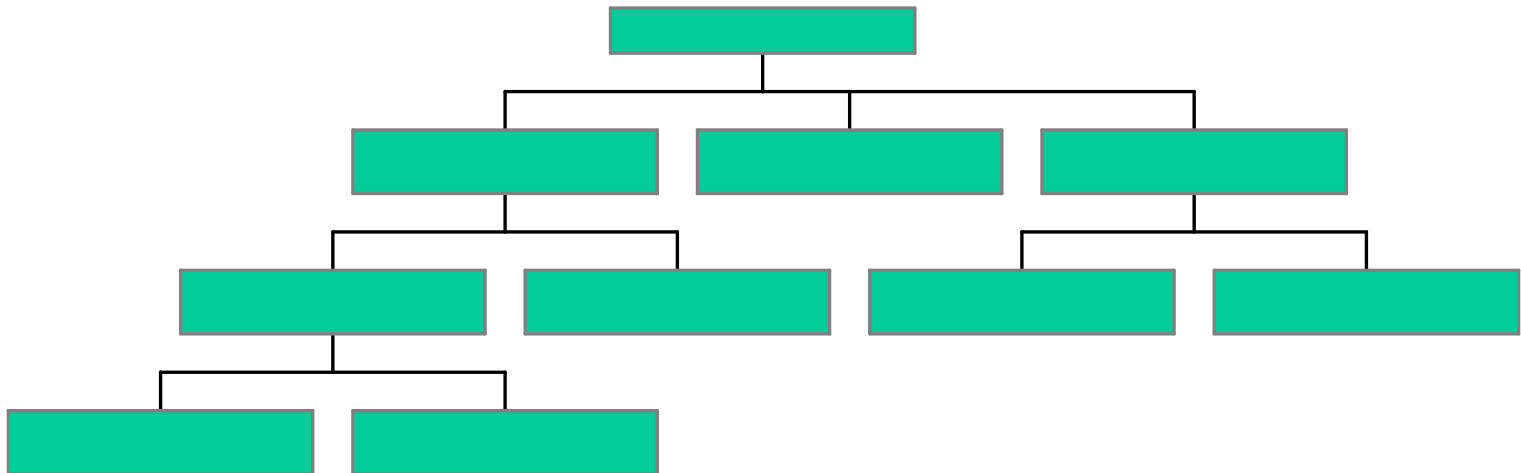
No.	Brand	Price (cost) \$	Normalized cost	Benefit	B/C
1	Miata	18,000	0.3333	0.35	1.0501
2	Civic	12,000	0.2222	0.30	1.3501
3	Saturn	15,000	0.2778	0.27	0.9719
4	Escort	9,000	0.1667	0.08	0.4799
Total		54,000	1		

The winner is Civic

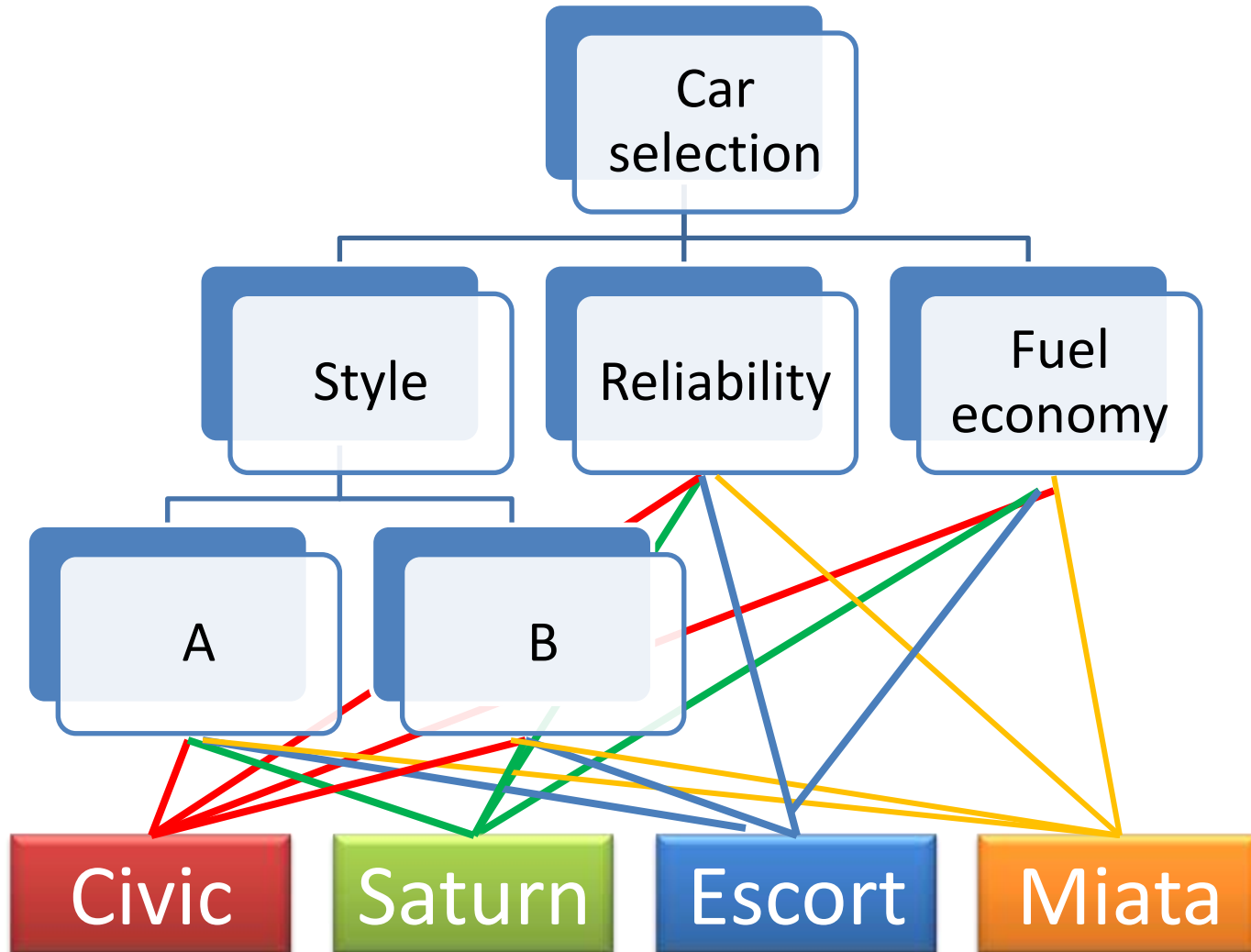


# Complex decisions

- Many levels of criteria and sub-criteria exists for complex problems.



# VÍ DỤ



# Priority vector for A and B

Style	A	B
A	1	5
B		1

PRIORITY VECTOR

	<b>0.83</b>	
	<b>0.17</b>	

# Ranking alternatives

## Factor A

	Civic	Saturn	Escort	Miata
Civic	1	1/5	7	1/6
Saturn	5	1	3	1/4
Escort	1/7	1/3	1	1/5
Miata	6	4	5	1

## Priority vector

$$\begin{bmatrix} .1648 \\ .2337 \\ .0646 \\ .5370 \end{bmatrix}$$

## Factor B

	Civic	Saturn	Escort	Miata
Civic	1	2	5	1
Saturn		1	3	2
Escort			1	1/4
Miata				1

$$\begin{bmatrix} \\ \\ \\ \end{bmatrix}$$

# Ranking alternatives

## Factor A

	Civic	Saturn	Escort	Miata
Civic	1	1/5	1/7	1/6
Saturn	5	1	1/2	1/3
Escort	1/7	2	1	1/4
Miata	6	3	4	1

## Priority vector

.1648
.2337
.0646
.5370

Remember  
CR ?!

$$CR = 0.3626 > 0.1$$

# Ranking alternatives

## Factor A

	Civic	Saturn	Escort	Miata
Civic	1	2	5	1/6
Saturn		1	2	1/4
Escort			1	1/7
Miata				1

CR = 0.083

## Priority vector

$$\begin{bmatrix} .2092 \\ .1230 \\ .0627 \\ .6051 \end{bmatrix}$$

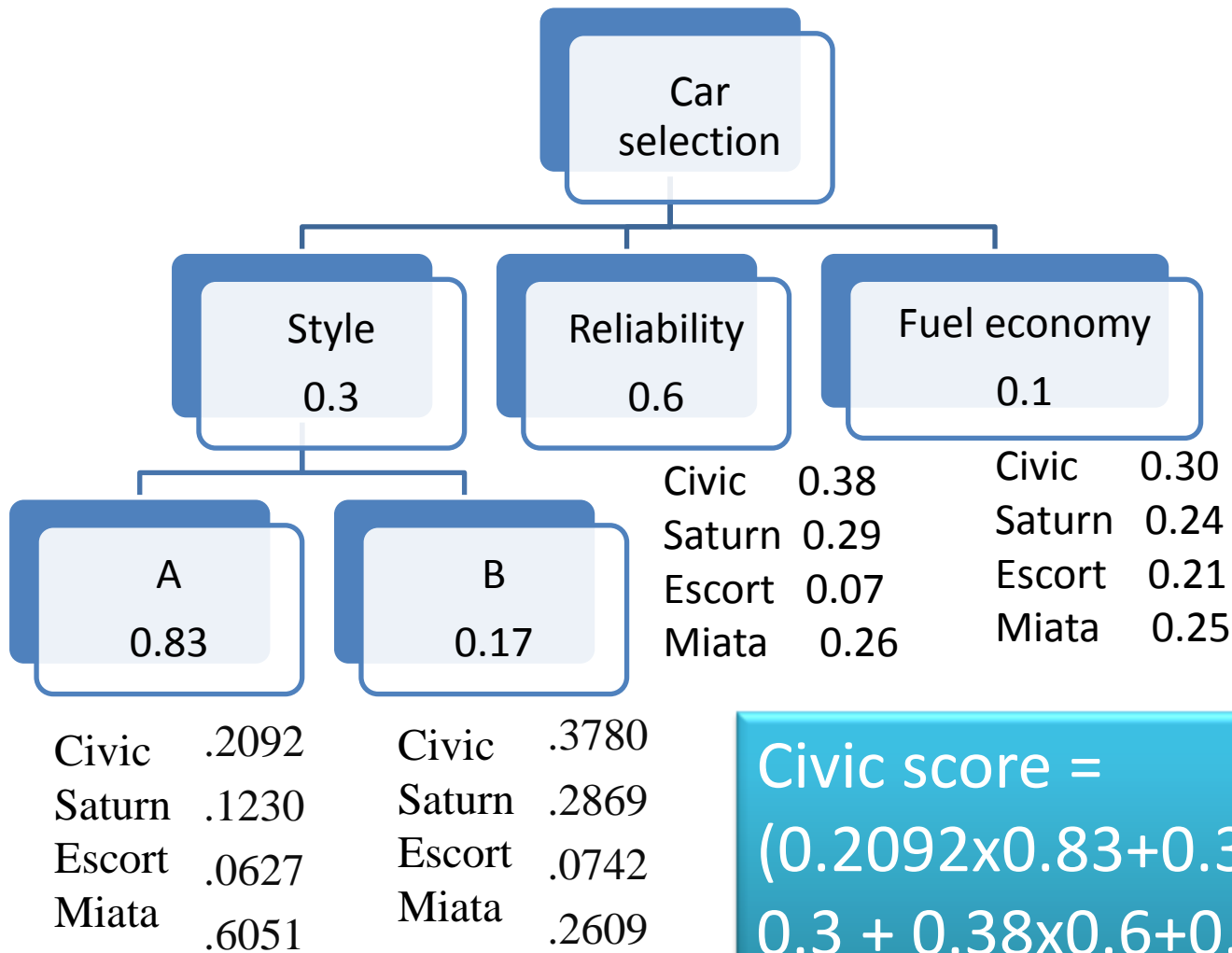
## Factor B

	Civic	Saturn	Escort	Miata
Civic	1	2	5	1
Saturn		1	3	2
Escort			1	1/4
Miata				1

CR = 0.079

$$\begin{bmatrix} .3780 \\ .2869 \\ .0742 \\ .2609 \end{bmatrix}$$

# Calculate score



Civic score =  
 $(0.2092 \times 0.83 + 0.378 \times 0.17) \times$   
 $0.3 + 0.38 \times 0.6 + 0.3 \times 0.1$

# AHP Software:

Professional commercial software **Expert Choice** developed by Expert Choice Inc. is available which simplifies the implementation of the AHP's steps and automates many of its computations

- computations
- sensitivity analysis
- graphs, tables



# Ex 2: Evaluation of Job Offers

Ex: Peter is offered 4 jobs from Acme Manufacturing (A), Bankers Bank (B), Creative Consulting (C), and Dynamic Decision Making (D).  
He bases his evaluation on the criteria such as location, salary, job content, and long-term prospects.

Step 1: Decide upon the relative importance of the selection criteria:

	Location	Salary	Content	Long-term
Location	1	1/5	1/3	1/2
Salary	5	1	2	4
Content	3	1/2	1	3
Long-term	2	1/2	1/3	1

# Priority Vectors:

- 1) Normalize the column entries by dividing each entry by the sum of the column.
- 2) Take the overall row averages

	Location	Salary	Content	Long-term	Average
Location	0.091	0.102	0.091	0.059	0.086
Salary	0.455	0.513	0.545	0.471	0.496
Content	0.273	0.256	0.273	0.353	0.289
Long-term	0.182	0.128	0.091	0.118	0.130
	$\frac{+}{1}$	$\frac{+}{1}$	$\frac{+}{1}$	$\frac{+}{1}$	$\frac{+}{1}$

# Example 2: Evaluation of Job Offers

Step 2: Evaluate alternatives w.r.t. each criteria

Location Scores

	A	B	C	D
A	1	1/2	1/3	5
B	2	1	1/2	7
C	3	2	1	9
D	1/5	1/7	1/9	1

Relative Location Scores

	A	B	C	D	Avg.
A	0.161	0.137	0.171	0.227	0.174
B	0.322	0.275	0.257	0.312	0.293
C	0.484	0.549	0.514	0.409	0.489
D	0.032	0.040	0.057	0.045	0.044

# Example 2: Calculation of Relative Scores

Relative Scores for Each Criteria					Relative weights for each criteria	Relative scores for each alternative
	Location	Salary	Content	Long-Term		
<b>A</b>	0.174	0.050	0.210	0.510	x	0.164
<b>B</b>	0.293	0.444	0.038	0.012		0.256
<b>C</b>	0.489	0.312	0.354	0.290		0.335
<b>D</b>	0.044	0.194	0.398	0.188		0.238