# **Arguing from Experience To Classifying Noisy Data**

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**Introduction: Arguing from Experience** 

**Classifying Through Argumentation** 

**Example 1** 

Example 2

**More Experiments** 

# **Arguing from Experience**

### **Experience May Differ**

Geographically

**Exception may be only rarely encountered** 

Samples may be abnormal



### **Advantages**

No previous analysis

No knowledge engineering bottleneck

**Existing Databases can be used** 

Can deal with gaps and conflicts

**Previous Experience** 

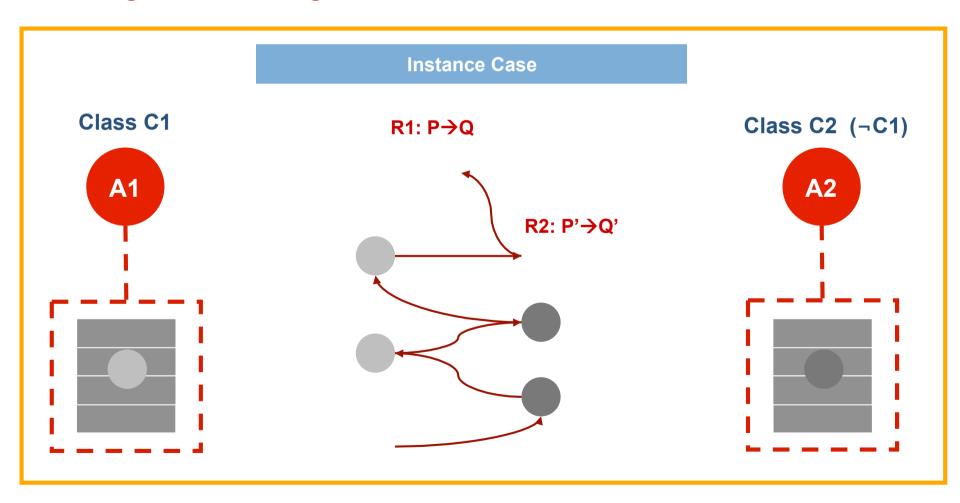
**Common law** 

**Every day argument** 

### What is PADUA?

**Protocol for Argumentation Dialogue Using Association rule mining** 

A Dialogue Game to Argue about CLASSIFICATION



### What is PADUA?

#### **PADUA Speech acts**

#### PATDUM PHOTOCOL

2: Distinguish

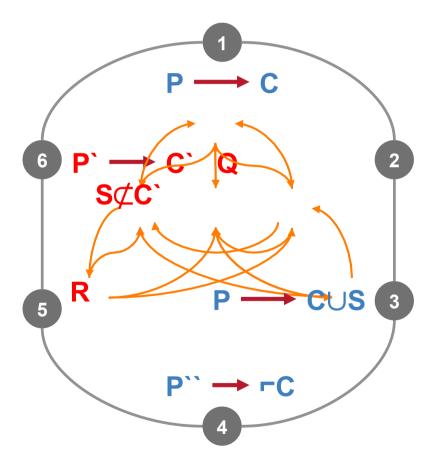
3: unwanted consequences

4: Counter Rule

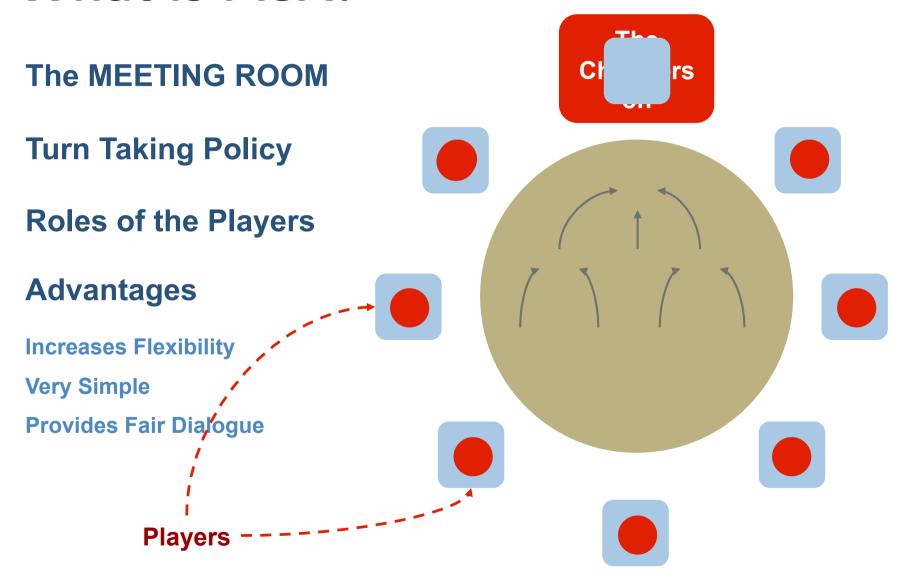
**5: Increase Confidence** 

**6: Withdraw unwanted consequences** 

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### What is PISA?



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### Classifying through Argumentation

#### **Arguing From Experience for Classification**

Provides means to debate the possible classifications of a given case between a number of participants, each defending one possibility.

Einchasententine at the bring intermedent of a three points of the correct one Distinguish a given rule by adding additional attributes

Eacheragentaforvewlates by gramonts of catton background dataset of past examples which is mined as required.

#### The process of classification – 3 ARM queries

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**Example 1** 

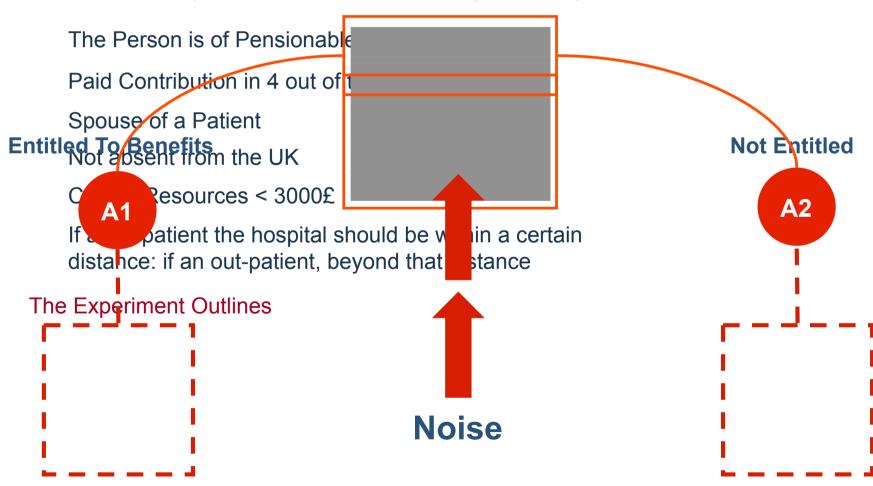
Example 2

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# Example 1 – Welfare Benefits

#### The Welfare Benefits

The benefit is payable if 6 conditions, covering wide range of possibilities.



# Example 1 – Welfare Benefits

Noise	PADUA	RDT	IGDT	TFPC	СВА	CMAR	FOIL	CPAR	PRM	CN2	ABCN2
0	99.9	100	92.5	98.5	99.2	96.8	99.7	67.1	66.7	99.5	99.8
2	99.9	98.6	88.2	98.3	100	98.8	100	65.4	65.4	97.8	98.4
5	99.3	99.6	93.3	99.86	98.8	98.1	94.2	65.4	65.4	96.4	96.9
10	98.5	98.3	92.8	97.08	91.9	97.2	93.2	64.4	64.4	93.5	94.7
20	97.8	97.3	90.9	98.8	86.9	97.3	88.9	61.7	63.6	88.7	92.0
40	97.1	96.4	90.4	96.3	94.0	96.8	89.4	58.1	57.9	83.3	85.0

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# **Example 2 – Housing Benefits**

#### The Housing Benefits

Housing benefit is payable if 6 conditions, covering wide range of possibilities.

The Person is of Pensionable Age

Baid Contain the last 5 years

Simphilans and from the Calmers Armed Forces

Income < 20%

Capital Resources < 3000£

The Experiment Outlines

# **Example 1 – Housing Benefits**

Noise	PADUA	RDT	IGDT	TFPC	СВА	CMAR	FOIL	CPAR	PRM
0%	99.86	99.72	77.00	98.33	97.36	99.31	100.00	64.03	66.81
2%	99.72	97.78	76.25	98.61	99.86	98.01	96.67	63.75	64.72
5%	99.58	98.89	64.31	96.53	97.50	98.61	94.44	65.28	65.14
10%	98.61	98.75	73.61	93.61	91.11	95.69	87.08	63.61	64.92
20%	96.81	98.19	73.06	93.89	96.25	96.50	86.39	62.28	64.58
40%	96.11	92.22	64.44	83.06	92.08	92.92	86.11	60.97	61.25
50%	94.03	88.75	62.22	54.72	84.17	85.31	78.19	59.58	61.81

# Housing Benefits – 4 Classes

#### **Housing Benefits for PISA**

Housing Benefit is either Payable, Payable with Priority, Partial Benefit or Not Payable.

One dataset is generated and noise is added in the same manner.

The training set is split into four equal datasets, each is given to one participant.

PISA is applied to classify the cases in the test set.

# Housing Benefits – 4 Classes

Noise	PISA	RDT	IGDT	TFPC	СВА	CMAR	FOIL	CPAR	PRM
0%	98.47	94.44	68.19	92.56	90.28	86.75	92.25	75.83	75.83
2%	97.64	90.56	67.75	91.81	90.14	86.25	92.22	75.42	68.06
5%	97.36	93.47	62.92	89.72	90.69	85.00	91.39	73.33	73.89
10%	96.53	92.92	60.97	86.81	89.17	84.25	92.36	70.83	72.64
20%	95.69	91.94	60.56	80.83	88.89	83.75	89.31	70.78	70.61
40%	94.44	90.31	56.35	69.86	86.81	81.75	80.56	63.06	63.06
50%	93.75	88.36	61.81	45.83	62.71	80.50	70.42	63.06	65.83

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### Real – World Datasets

#### **Applying PISA and PADUA with Real Datasets**

PISA and PADUA were applied using 7 real world datasets.

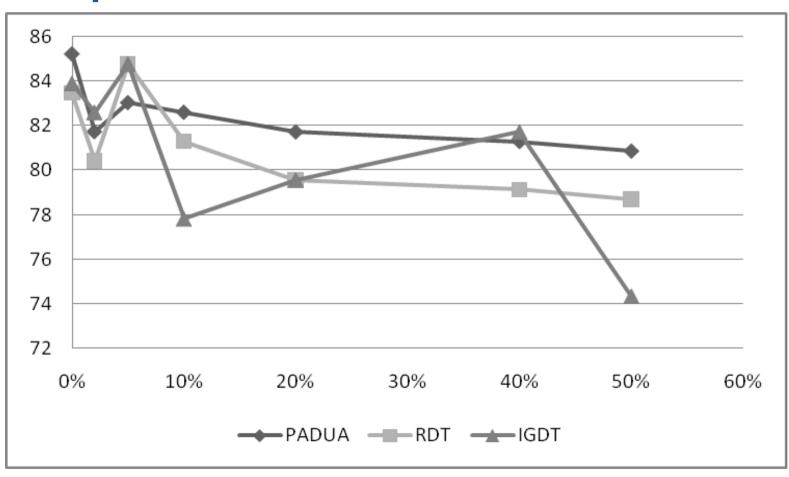
The tests compared PISA and PADUA to the same classifiers as before.

The tests used the same noise model as before.

The obtained results show a similar pattern to the benefits experiments: the accuracy of almost all the classes dropped when the noise percentage was increased.

### Real - World Datasets

### Example MIMSA Datase Dataset



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### **Discussion and Conclusions**

#### **Arguing From Experience**

Provides a means for an <u>unlimited number</u> of agents to engage in discussion about a classification on the basis of raw data, unmediated by knowledge representation effort to present this data in the form of rules.

#### **Arguing from Experience Applications**

In domains in which there are large volumes of data available.

several distributed datasets generated from different samples

With high level of noise in the training dataset that are expensive to clean.

### **Thanks For Listening**