

# Roughness by Residuals

M. E. Müller

Univ. Augsburg

Wednesday 30<sup>th</sup> September, 2015

# Contents

Introduction

Approximations

Rough sets (without elements)

- ... but with characteristic functions

- ... but with subidentities

Finding core relations and reducts

Conclusion

# Introduction

At-least and at-most approximations

$\square$  and  $\diamond$

Morphology :  $\epsilon(\cdot)$  and  $\delta(\cdot)$  w.r.t binary relations

KAT :  $[\cdot|\cdot]$  and  $\langle\cdot|\cdot\rangle$  w.r.t tests

Residuals :  $\text{wp}$  and  $\text{sp}$  w.r.t p.o., preorder, ...

Approximations:  $\llbracket\cdot\rrbracket$  and  $\langle\langle\cdot\rangle\rangle$  w.r.t equivalences

# Information systems and kernel relations

## Features F








<i>U</i>	<b>F</b>			
	<i>col</i>	<i>shp</i>	<i>edg</i>	<i>siz</i>
□	w	square	4	S
■	b	square	4	B
■	b	square	4	S
●	g	circle	1	S
△	w	triangle	3	B
◆	b	diamond	4	S
○	w	circle	1	S

## Attributes A








<b>A</b>													
<i>col</i>			<i>shp</i>				<i>edg</i>			<i>siz</i>			
w	g	b	c	t	d	s	1	3	4	S	B		
1	0	0	0	0	0	1	0	0	1	1	0		
0	0	1	0	0	0	1	0	0	1	0	1		
0	0	1	0	0	0	1	0	0	1	1	0		
0	1	0	1	0	0	0	1	0	0	1	0		
1	0	0	0	1	0	0	0	1	0	0	1		
0	0	1	0	0	1	0	0	0	1	1	0		
1	0	0	1	0	0	0	1	0	0	1	0		

# Information systems and kernel relations

## Features $F$

$U$	$F$			
	$col$	$shp$	$edg$	$siz$
	w	square	4	S
	b	square	4	B
	b	square	4	S
	g	circle	1	S
	w	triangle	3	B
	b	diamond	4	S
	w	circle	1	S








## Kernel relations

							$\tilde{col}$
							
<b>1</b>	0	0	0	<b>1</b>	0	<b>1</b>	
0	<b>1</b>	<b>1</b>	0	0	<b>1</b>	0	
0	<b>1</b>	<b>1</b>	0	0	<b>1</b>	0	
0	0	0	<b>1</b>	0	0	0	
<b>1</b>	0	0	0	<b>1</b>	0	<b>1</b>	
0	<b>1</b>	<b>1</b>	0	0	<b>1</b>	0	
<b>1</b>	0	0	0	<b>1</b>	0	<b>1</b>	

**Def.** Kernel relation:  $x \tilde{f} y : \iff f(x) = f(y)$








# Information systems and kernel relations

## Features $F$

$U$	$F$			
	$col$	$shp$	$edg$	$siz$
	w	square	4	S
	b	square	4	B
	b	square	4	S
	g	circle	1	S
	w	triangle	3	B
	b	diamond	4	S
	w	circle	1	S

## Indiscernability

$$\tilde{\mathbf{R}} = \bigcap \{ \tilde{col}, \tilde{shp} \}$$

						
<b>1</b>	0	0	0	0	0	0
0	<b>1</b>	<b>1</b>	0	0	0	0
0	<b>1</b>	<b>1</b>	0	0	0	0
0	0	0	<b>1</b>	0	0	0
0	0	0	0	<b>1</b>	0	0
0	0	0	0	0	<b>1</b>	0
0	0	0	0	0	0	<b>1</b>

Def. Kernel relation:  $x \tilde{f} y : \iff f(x) = f(y)$

Def. Indiscernability:  $x \tilde{\mathbf{R}} y : \iff x(\bigcap \mathbf{R})y$

# Information systems and kernel relations

Features **F**

$\mathcal{U}$	<b>F</b>			
	<i>col</i>	<i>shp</i>	<i>edg</i>	<i>siz</i>
□	w	square	4	S
■	b	square	4	B
■	b	square	4	S
●	g	circle	1	S
△	w	triangle	3	B
◆	b	diamond	4	S
○	w	circle	1	S

$$\tilde{\mathbf{R}} = \bigcap \{ \tilde{col}, \tilde{shp} \}$$

□	■	■	●	△	◆	○
<b>1</b>	0	0	0	0	0	0
0	<b>1</b>	<b>1</b>	0	0	0	0
0	<b>1</b>	<b>1</b>	0	0	0	0
0	0	0	<b>1</b>	0	0	0
0	0	0	0	<b>1</b>	0	0
0	0	0	0	0	<b>1</b>	0
0	0	0	0	0	0	<b>1</b>

**Goal:** Find a smallest  $\mathbf{R} \subseteq \text{EquR}(\mathcal{U})$  that creates a finest  $\tilde{\mathbf{R}}$ !  
 ( $\tilde{\mathbf{R}} = 1$  or  $\tilde{\mathbf{R}} \approx T$ )

# Rough sets, “pointwise”

**Def.** Upper and Lower Approximations

N.B.  $R := \tilde{R}$

$$[[R]]s := \{x \in \mathcal{U} : [x]_R \subseteq s\} \quad (1)$$

$$\langle\langle R \rangle\rangle s := \{x \in \mathcal{U} : [x]_R \cap s \neq \emptyset\}. \quad (2)$$

Iso-/Antitony of  $[[ \ ]]$  and  $\langle\langle \ \rangle\rangle$  w.r.t. set and relation arguments

$$s \subseteq t \implies \begin{array}{l} \langle\langle R \rangle\rangle s \subseteq \langle\langle R \rangle\rangle t \\ [[R]]s \subseteq [[R]]t \end{array} \quad \text{but } P \subseteq R \implies \begin{array}{l} \langle\langle P \rangle\rangle s \subseteq \langle\langle R \rangle\rangle s \\ [[P]]s \supseteq [[R]]s \end{array}$$

Duality (as desired)

$$[[R]]\bar{s} = \overline{\langle\langle R \rangle\rangle s}. \quad (3)$$



# Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff [x]_R \subseteq s$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \setminus \dot{s} \mid \{\mathbf{1}\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \setminus \bar{\dot{s}} \mid \{\mathbf{0}\} \rangle \quad (5)$$

where

$$\text{RR: } P \setminus Q := \overline{P^\vee \setminus \bar{Q}} \quad \text{or} \quad R \subseteq P \setminus Q \iff P \setminus R \subseteq Q$$

$$\text{LR: } Q // P := \overline{\bar{Q} \setminus P^\vee} \quad \text{or} \quad R \subseteq Q // P \iff R \setminus P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

# Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff xRy \longrightarrow y \in s$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \ll \dot{s} \mid \{1\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \ll \bar{\dot{s}} \mid \{0\} \rangle \quad (5)$$

where

$$\text{RR: } P \ll Q := \overline{P^\cup \bar{Q}} \quad \text{or} \quad R \subseteq P \ll Q \iff P \dot{;} R \subseteq Q$$

$$\text{LR: } Q // P := \overline{\bar{Q} \dot{;} P^\cup} \quad \text{or} \quad R \subseteq Q // P \iff R \dot{;} P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

# Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff x \bar{R} y \vee y \dot{s} \mathbf{1}$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \ll \dot{s} \mid \{\mathbf{1}\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \ll \bar{\dot{s}} \mid \{\mathbf{0}\} \rangle \quad (5)$$

where

$$\text{RR: } P \ll Q := \overline{P^\vee \dot{;} \bar{Q}} \quad \text{or} \quad R \subseteq P \ll Q \iff P \dot{;} R \subseteq Q$$

$$\text{LR: } Q // P := \overline{\bar{Q} \dot{;} P^\vee} \quad \text{or} \quad R \subseteq Q // P \iff R \dot{;} P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

# Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff \neg(y \bar{\dot{s}} \mathbf{1} \wedge xRy)$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \ll \dot{s} \mid \{\mathbf{1}\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \ll \bar{\dot{s}} \mid \{\mathbf{0}\} \rangle \quad (5)$$

where

$$\text{RR: } P \ll Q := \overline{P^\vee \circledast \bar{Q}} \quad \text{or} \quad R \subseteq P \ll Q \iff P \circledast R \subseteq Q$$

$$\text{LR: } Q // P := \overline{\bar{Q} \circledast P^\vee} \quad \text{or} \quad R \subseteq Q // P \iff R \circledast P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

## Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff x \overline{R \circ \dot{s}} \mathbf{1}$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \ll \dot{s} \mid \{\mathbf{1}\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \ll \bar{\dot{s}} \mid \{\mathbf{0}\} \rangle \quad (5)$$

where

$$\text{RR : } P \ll Q := \overline{P \circ \bar{Q}} \quad \text{or} \quad R \subseteq P \ll Q \iff P \circ R \subseteq Q$$

$$\text{LR : } Q // P := \overline{Q \circ P} \quad \text{or} \quad R \subseteq Q // P \iff R \circ P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

# Rough sets by characteristic functions

We represent  $s \subseteq \mathcal{U}$  by its characteristic relation  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$ .

We observe (pointwise):

$$x \in \llbracket R \rrbracket s \iff x R \ll \dot{s} \mathbf{1}$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \langle R \ll \dot{s} \mid \{\mathbf{1}\} \rangle \quad (4)$$

$$\langle R \rangle s := \langle R \ll \bar{\dot{s}} \mid \{\mathbf{0}\} \rangle \quad (5)$$

where

$$\text{RR: } P \ll Q := \overline{P^\vee \bar{Q}} \quad \text{or} \quad R \subseteq P \ll Q \iff P \dot{;} R \subseteq Q$$

$$\text{LR: } Q // P := \overline{Q \dot{;} P^\vee} \quad \text{or} \quad R \subseteq Q // P \iff R \dot{;} P \subseteq Q$$

with  $R$  being the biggest solution of the respective inequalities.

# Example

$\dot{c}$	<b>2</b>	
	<b>0</b>	<b>1</b>
□	1	0
■	1	0
■	0	1
●	0	1
△	0	1
◆	1	0
○	1	0

$\tilde{s}hp$	$\mathcal{U}$							
	□	■	■	●	△	◆	○	
□	1	1	1	0	0	0	0	
■	1	1	1	0	0	0	0	
■	1	1	<del>1</del>	0	0	0	0	
●	0	0	0	<del>1</del>	0	0	1	
△	0	0	0	0	1	0	0	
◆	0	0	0	0	0	1	0	
○	0	0	0	1	0	0	1	

$\tilde{s}hp \parallel \dot{c}$	<b>2</b>	
	<b>0</b>	<b>1</b>
□	0	0
■	0	0
■	0	0
●	0	0
△	0	1
◆	1	0
○	0	0

Therefore,  $\langle \tilde{s}hp \parallel \dot{c} \mid \{1\} \rangle = \langle \overline{\tilde{s}hp} \circ \bar{c} \mid \{1\} \rangle = \{\triangle\} = \llbracket \tilde{s}hp \rrbracket \{\blacksquare, \bullet, \triangle\}$ .

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff [x]_R \subseteq S\mathcal{U}$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \setminus S ) \quad (6)$$

$$\langle R \rangle s := \overline{\text{dom} ( R \setminus S^- )} \quad (7)$$



## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff \forall y : xRy \longrightarrow ySy$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \parallel S ) \quad (6)$$

$$\langle R \rangle s := \overline{\text{dom} ( R \parallel S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U}/Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \parallel C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U}/Q$  is a *target classification*.

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff \forall y : x \overline{R} y \vee y S y$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \ll S ) \quad (6)$$

$$\langle R \rangle s := \overline{\text{dom} ( R \ll S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U} / Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \ll C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U} / Q$  is a *target classification*.

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \bar{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff \forall y : \neg(y \bar{S} y \wedge x R y)$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \ll S ) \quad (6)$$

$$\langle R \rangle s := \overline{\text{dom} ( R \ll S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U}/Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \ll C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U}/Q$  is a *target classification*.

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff \forall y : x R^\smile ; \overline{S} y$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \ll S ) \quad (6)$$

$$\langle R \rangle s := \overline{\text{dom} ( R \ll S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U} / Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \ll C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U} / Q$  is a *target classification*.

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff \forall y : x R \ll S y$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \ll S ) \quad (6)$$

$$\langle \llbracket R \rrbracket s \rangle := \overline{\text{dom} ( R \ll S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U} / Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \ll C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U} / Q$  is a *target classification*.

## Rough sets by subidentities

We represent  $s \subseteq \mathcal{U}$  by its subidentity  $S = 1 \cap (s \times s)$  and define  $S^- := 1 \cap \overline{S}$ . We observe:

$$x \in \llbracket R \rrbracket s \iff x \in R \ll S \mathcal{U}$$

**Def.** Upper and Lower Approximations, again.

$$\llbracket R \rrbracket s := \text{dom} ( R \ll S ) \quad (6)$$

$$\langle \langle R \rangle \rangle s := \overline{\text{dom} ( R \ll S^- )} \quad (7)$$

**Def.** Approximations of classifications

$$\llbracket R \rrbracket \mathcal{U} / Q := \{ \llbracket R \rrbracket c : c \in \mathfrak{c} \} = \{ \text{dom} ( R \ll C ) : c \in \mathfrak{c} \} \quad (8)$$

where  $\mathfrak{c} = \mathcal{U} / Q$  is a *target classification*.

# Rough sets by KAD

## Preimages, domains and tests

With KAT we are given

$$\langle R | := \min \{ X \in \mathbf{U} : R \subseteq X \circ R \} \quad (9)$$

$$- \langle R | := \max \{ X \in \mathbf{U} : X \circ R \subseteq \overline{C} \} \quad (10)$$

Here,  $C = \top$  (i.e. not a relative, but absolute complement).

Then, by domain laws,

$$\langle R | S = \langle R \circ S | = \langle \overline{R \setminus S} | = \langle R \rangle_s \quad (11)$$

and, canonically,

$$[R | \top := - \langle R | \top^- = \langle R \setminus \top | = \llbracket R \rrbracket t. \quad (12)$$

# Rough sets by GC

... are for free.

With  $\llbracket \cdot \rrbracket$  /  $\langle \cdot \rangle$  being defined by  $[\cdot] \cdot$  /  $\langle \cdot | \cdot$ , and reading sets as subidentities,

$$\langle R | s \subseteq t \iff \langle R | t^- \subseteq s^- \iff s \subseteq -\langle R | t^- \iff s \subseteq [R] t. \quad (13)$$

What we get for free

- ▶  $\llbracket R \rrbracket \bar{s} = \overline{\langle R \rangle s}$  proving (3).
- ▶  $\llbracket R \rrbracket s = s \iff \langle R \rangle s = s$  proving (17, 18) in the paper.
- ▶ ...
- ▶ a nice metaphor: “RST is just a GC, with only equivalences”.



# Positive regions and refinement

How good is  $R$  to ...

- ▶ describe a classification  $U/Q$  (or  $s/Q$ )...
- ▶ compared to a relation  $P$ ?

**Def.** Positive regions

$$\llbracket R \triangleleft P \rrbracket_{s/Q} := \bigcup_{c \in s/Q} \llbracket R \triangleleft P \rrbracket_c := \bigcup_{c \in s/Q} \bigcup_{t \in c/P} \llbracket R \rrbracket_t. \quad (14)$$

$\llbracket R \triangleleft P \rrbracket_{s/Q}$  is the *largest set of elements*  $x \in s$  s.t.  $[x]_R \subseteq [x]_P \subseteq [x]_Q$ .

# Positive regions and refinement

How good is  $R$  to ...

- ▶ describe a classification  $U/Q$  (or  $s/Q$ )...
- ▶ compared to a relation  $P$ ?

**Def.** Positive regions

$$\llbracket R \triangleleft P \rrbracket_{s/Q} := \bigcup_{c \in s/Q} \llbracket R \triangleleft P \rrbracket_c := \bigcup_{c \in s/Q} \bigcup_{t \in c/P} \llbracket R \rrbracket t. \quad (14)$$

Positive regions by residuals

$$\llbracket R \triangleleft P \rrbracket_{s/Q} = \langle (R \parallel P) \parallel Q \mid s = \text{dom}((R \parallel P) \parallel Q \circ S) \rangle \quad (15)$$

$$\begin{array}{l} \text{s.t.} \quad \llbracket R \rrbracket_s = \text{dom}((R \parallel \tilde{s}) \parallel S) = \text{dom}(R \parallel S) \\ \text{but} \quad \bigcup \llbracket R \rrbracket_{(s/Q)} = R \parallel Q \cdot s = \text{dom}(R \parallel Q \circ S) \end{array}$$

$R$  is better than  $P$ , if...

**Def.** Refinement

$R$  refines  $P$  w.r.t  $H$  on  $s/Q$ , iff

$$\begin{aligned} R \underset{s/Q}{\overset{H}{\succeq}} P &:\iff \llbracket R \triangleleft H \rrbracket(s/Q) \supseteq \llbracket P \triangleleft H \rrbracket(s/Q) & (16) \\ &\iff \text{dom}(((R \setminus H) \setminus Q);S) \supseteq \text{dom}(((P \setminus H) \setminus Q);S) \end{aligned}$$

**Def.**  $\mathbf{R}$  is a ( $H$ -relative) reduct of  $\mathbf{P}$  (on  $s/Q$ ), if...

1.  $\mathbf{R} \subseteq \mathbf{P}$  (hence  $\tilde{\mathbf{P}} \subseteq \tilde{\mathbf{R}}$  and  $\mathbf{P} \succeq \mathbf{R}$ )
2.  $\mathbf{R}$  is a *largest* subset of  $\mathbf{P}$  s.t.  $\mathbf{R} \succeq \mathbf{P}$ .

$\text{Red}(\mathbf{P})$  is the set of all reducts of  $\mathbf{P}$ ;  $\text{Cor}(\mathbf{P}) := \bigcap \text{Red}(\mathbf{R})$ .

$R$  is better than  $P$ , if...

**Def.** Refinement

$R$  refines  $P$  w.r.t  $H$  on  $s$ , iff

$$\begin{aligned} R \underset{s}{\overset{H}{\succeq}} P & \iff \llbracket R \triangleleft H \rrbracket_s \supseteq \llbracket P \triangleleft H \rrbracket_s & (16) \\ & \iff \text{dom}((R \setminus\!\!\setminus H);S) \supseteq \text{dom}((P \setminus\!\!\setminus H);S) \end{aligned}$$

**Def.**  $\mathbf{R}$  is a ( $H$ -relative) reduct of  $\mathbf{P}$  (on  $s/Q$ ), if...

1.  $\mathbf{R} \subseteq \mathbf{P}$  (hence  $\tilde{\mathbf{P}} \subseteq \tilde{\mathbf{R}}$  and  $\mathbf{P} \succeq \mathbf{R}$ )
2.  $\mathbf{R}$  is a *largest* subset of  $\mathbf{P}$  s.t.  $\mathbf{R} \succeq \mathbf{P}$ .

$\text{Red}(\mathbf{P})$  is the set of all reducts of  $\mathbf{P}$ ;  $\text{Cor}(\mathbf{P}) := \bigcap \text{Red}(\mathbf{R})$ .

$R$  is better than  $P$ , if...

**Def.** Refinement

$R$  refines  $P$  w.r.t  $H$  on  $\mathcal{U}$ , iff

$$\begin{aligned} R \stackrel{H}{\succeq} P &: \iff \llbracket R \triangleleft H \rrbracket \mathcal{U} \supseteq \llbracket P \triangleleft H \rrbracket \mathcal{U} & (16) \\ &\iff \text{dom}(R \setminus H) \supseteq \text{dom}(P \setminus H) \end{aligned}$$

**Def.**  $\mathbf{R}$  is a ( $H$ -relative) reduct of  $\mathbf{P}$  (on  $s/Q$ ), if...

1.  $\mathbf{R} \subseteq \mathbf{P}$  (hence  $\tilde{\mathbf{P}} \subseteq \tilde{\mathbf{R}}$  and  $\mathbf{P} \succeq \mathbf{R}$ )
2.  $\mathbf{R}$  is a *largest* subset of  $\mathbf{P}$  s.t.  $\mathbf{R} \succeq \mathbf{P}$ .

$\text{Red}(\mathbf{P})$  is the set of all reducts of  $\mathbf{P}$ ;  $\text{Cor}(\mathbf{P}) := \bigcap \text{Red}(\mathbf{R})$ .

## We need to make a few points

It is nice to know that  $\mathbf{R} \in \text{Red}(\mathbf{P})$ ,  
but the definition does not help finding such  $\mathbf{R}$ .

**Def.** “ $\dot{s}$ ” is an arbitrarily chosen, fixed element (point) of  $s$ .

## Sets and Partitions

### ► Sets

► characteristic relations:  $\dot{s} : \mathcal{U} \rightarrow \mathbf{2}$

$$s = \langle \dot{s} | 1 \rangle$$

► subidentities:  $S := 1 \cap (s \times s)$

$$s = \langle S | \mathcal{U} \rangle$$

### ► Classifications

► quotients  $s/R$ :

$$s/R = \{ \langle R | x : x \in s \rangle \}$$

► pointwise representation:  $r \subseteq s$

$$s/R = \{ \langle [c] : c \in r \rangle \}$$

# Concrete Tasks

... sadly require concrete data (points)

**Goal:** Construct (efficiently) some  $\mathbf{R}$  in  $\text{Red}(\mathbf{P})$

Gedankenexperiment

$(\mathbf{Q} := \mathbf{R} - \{R\})$

Suppose,  $R \in \text{Cor}(\mathbf{P})$ .

- ▶ Then,  $R \in \mathbf{R}$  for all  $\mathbf{R} \in \text{Red}(\mathbf{P})$ .
- ▶ Hence,  $\mathbf{R} \succ^{\mathbf{P}} \mathbf{Q}$  (strictly!).
- ▶ By definition,  $\exists x : x \notin [(\mathbf{Q} \triangleleft \mathbf{P})]_s \subset [(\mathbf{R} \triangleleft \mathbf{P})]_s \ni x$ .
- ▶ Then,  $[x]_{\mathbf{R}}^{\approx} \subseteq [x]_{\mathbf{P}}^{\approx}$  but  $[x]_{\mathbf{Q}} \not\subseteq [x]_{\mathbf{P}}^{\approx}$ .
- ▶ Therefore, there is a representation  $r$  of  $s/\tilde{\mathbf{P}}$ , s.t.

$$\exists r_1, r_2 : r_1 \overline{R} r_2 \wedge [r_1]_{\mathbf{Q}}^{\approx} = [r_2]_{\mathbf{Q}}^{\approx} \quad (17)$$

# Concrete Tasks

... sadly require concrete data (points)

**Goal:** Construct (efficiently) some  $\mathbf{R}$  in  $\text{Red}(\mathbf{P})$

Gedankenexperiment

$(\mathbf{Q} := \mathbf{R} - \{R\})$

Suppose,  $R \in \text{Cor}(\mathbf{P})$ .

- ▶ Then,  $R \in \mathbf{R}$  for all  $\mathbf{R} \in \text{Red}(\mathbf{P})$ .
- ▶ Hence,  $\mathbf{R} \stackrel{\mathbf{P}}{\succ} \mathbf{Q}$  (strictly!).
- ▶ By definition,  $\exists x : x \notin [(\mathbf{Q} \triangleleft \mathbf{P})]_s \subset [(\mathbf{R} \triangleleft \mathbf{P})]_s \ni x$ .
- ▶ Then,  $[x]_{\mathbf{R}} \approx \subseteq [x]_{\mathbf{P}} \approx$  but  $[x]_{\mathbf{Q}} \not\subseteq [x]_{\mathbf{P}} \approx$ .
- ▶ Therefore, there is a representation  $r$  of  $s/\tilde{\mathbf{P}}$ , s.t.

$$\exists r_1, r_2 : r_1 \bar{R} r_2 \wedge r_1 \tilde{\mathbf{Q}} r_2 \quad (17)$$



# Concrete Tasks

... sadly require concrete data (points)

**Goal:** Construct (efficiently) some  $\mathbf{R}$  in  $\text{Red}(\mathbf{P})$

Gedankenexperiment

$(\mathbf{Q} := \mathbf{R} - \{R\})$

Suppose,  $R \in \text{Cor}(\mathbf{P})$ .

- ▶ Then,  $R \in \mathbf{R}$  for all  $\mathbf{R} \in \text{Red}(\mathbf{P})$ .
- ▶ Hence,  $\mathbf{R} \succ^{\mathbf{P}} \mathbf{Q}$  (strictly!).
- ▶ By definition,  $\exists x : x \notin \llbracket (\mathbf{Q} \triangleleft \mathbf{P}) \rrbracket_s \subset \llbracket (\mathbf{R} \triangleleft \mathbf{P}) \rrbracket_s \ni x$ .
- ▶ Then,  $[x]_{\mathbf{R}}^{\approx} \subseteq [x]_{\mathbf{P}}^{\approx}$  but  $[x]_{\mathbf{Q}} \not\subseteq [x]_{\mathbf{P}}^{\approx}$ .
- ▶ Therefore, there is a representation  $r$  of  $s/\tilde{\mathbf{P}}$ , s.t.

$$\exists r_1, r_2 : r_1 \bar{R} r_2 \wedge \forall Q \in \mathbf{R} : Q \neq R \longrightarrow r_1 Q r_2 \quad (17)$$

# A visual example

## "Discernability matrices"

$\Delta \mathbf{P}$	□	■	●	△	◆	○
□		ZC	CES	ZES	CES	ES
■			ZCES	CES	ZS	ZCES
●				ZCES	CES	C
△					ZCES	ZES
◆						CES
○						

1. No entry is nil — hence,  $\tilde{\mathbf{P}} = \bigcap \{ \tilde{siz}, \tilde{col}, \tilde{edg}, \tilde{shp} \} = 1$
2. In any  $\mathbf{R} \in \text{Red}(\mathbf{P})$ ,  $\tilde{shp}$  is indispensable: it is essential
3.  $\tilde{col} \in \text{Cor}(\mathbf{P})$

# Discernability of representatives

- ▶ If we have to make points, try making as few as possible.
- ▶ Speeding up exhaustive pointwise processes.

A (last) motivational example

$$\mathcal{U} = \{\square, \blacksquare, \bullet, \triangle, \blacklozenge, \circ\}$$

1. Let  $\mathbf{P} = \{\tilde{col}, \tilde{shp}\}$ .
2. Is  $\tilde{col}$  indispensable in  $\mathbf{P}$  w.r.t. 1?
  - 2.1 Choose wisely  $r_{\mathbf{P}-\{\tilde{col}\}} = r_{\tilde{shp}} = \{\blacksquare, \blacklozenge, \circ, \triangle\}$ .
  - 2.2 Consider  $\blacksquare$ : Then,  $[\blacksquare]_{\tilde{\mathbf{P}}} = \{\blacksquare\}$ .
  - 2.3 But:  $\blacksquare \notin [\{\tilde{shp}\} < \mathbf{P}] \mathcal{U}$   
because  $[\blacksquare]_{\tilde{shp}} = \{\blacksquare, \square\} \not\subseteq \{\blacksquare\} = [\blacksquare]_{\tilde{\mathbf{P}}}$
3. Hence,  $\tilde{col}$  is essential

## Finding Reducts (Skowron's exhaustive approach)

1. (Compute  $c := s/Q$ .)
2. (Compute  $h := \llbracket \mathbf{R} \triangleleft Q \rrbracket s$ .)
3. (Compute  $\mathbf{R} \setminus h$ .)
4. For every  $x, y \in \mathcal{U}$ , compute  $\{R : R \in \mathbf{R} \wedge x\overline{R}y\}$
5.  $\text{Cor}_c(\mathbf{R}) := \{R : \exists x, y \in r : x\overline{\bigcap \mathbf{R} - \{R\}}y\}$
6. For every  $\mathbf{P} \subseteq \mathbf{R} - \text{Cor}_c(\mathbf{R})$  :
  - $\mathbf{Q} := \mathbf{P} \cup \text{Cor}_c(\mathbf{R})$  is a  $Q$ -reduct of  $\mathbf{R}$  w.r.t.  $s$ , iff:
  - $\mathbf{Q} \underset{s}{\preceq}^Q \mathbf{R}$  and  $\mathbf{Q}$  is not a superset of any other reduct.

## Finding Reducts (A not so exhaustive approach)

1. (Compute  $c := s/Q$ .)
2. (Compute  $h := \llbracket \mathbf{R} \triangleleft Q \rrbracket s$ .)
3. **Guess a suitable**  $r := \{c : c \in h/Q\}$ .
4. For every  $x, y \in r$ , compute  $\{R : R \in \mathbf{R} \wedge x\overline{R}y\}$
5.  $\text{Cor}_c(\mathbf{R}) := \{R : \exists x, y \in r : x\overline{\bigcap \mathbf{R} - \{R\}}y\}$
6. For every  $\mathbf{P} \subseteq \mathbf{R} - \text{Cor}_c(\mathbf{R})$  **ordered by voodoo**:  
     $\mathbf{Q} := \mathbf{P} \cup \text{Cor}_c(\mathbf{R})$  is a  $Q$ -reduct of  $\mathbf{R}$  w.r.t.  $s$ , iff:  
     $\mathbf{Q} \underset{s}{\prec}^Q \mathbf{R}$  and  $\mathbf{Q}$  is not a superset of any other reduct.

# Conclusion

## Summary

- ▶ Rough sets by residuals
- ▶ Rough sets by GC
- ▶ An algorithm for reduct construction

## Prospects

- ▶ Rough sets by KAT (if not done yet)
- ▶ Rough sets by morphology (if not done yet)
- ▶ Rough sets by formal concept analysis (if not done yet)
- ▶ ... automatically create a searchable space of logic programs with  $\top$  being a program s.t.  $P \not\vdash E^-$  and  $\perp$  s.t.  $Q \vdash E^+$  with  $\approx$  as p.o. (not done yet)

$you \in \text{dom} ( \textit{Thanks}; \text{ATTENTION} ) .$