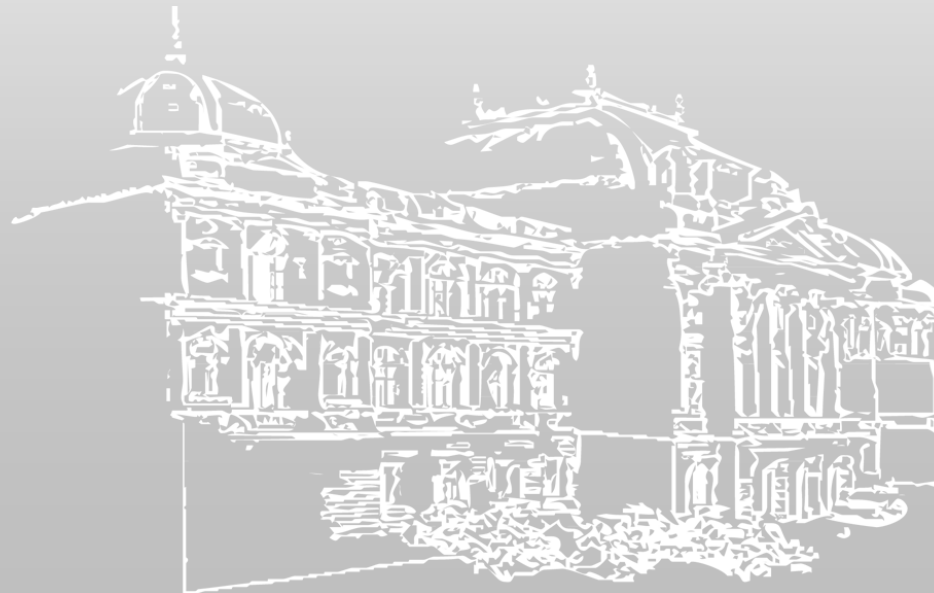


Fault localization in the light of faulty user input

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Why spreadsheets?

- Range
 - Used in nearly every company
- Fields of application
 - Forecasting
 - Financial Reporting

→ Basis for decisions
- Can be seen as programming language

→ Reuse ideas for other programming languages

To err is human.

3-5 % error rate when
writing formulas*

* Source: Panko *“Thinking is bad: Implications of human error research for spreadsheet research and practice,”* CoRR, 2008.

Outline

- 1 Running Example
- 2 Spectrum-based Fault Localization
- 3 Misclassifications
- 4 Impact Analysis
 - a Theoretical Analysis
 - b Empirical Evaluation
- 5 Conclusion

Running Example

	A	B	C	D	E	F	G	H	I	J
1	Worker	Mo	Tu	We	Th	Fr	Total	Overtime	\$/h	Total wage
2	Jones	9	9	8	8	8	42	2	20	860
3	Smith	9	9	10	7	8	43	1	24	1044
4	Total						85	3		
5										
6	Regular working time				40					
7	Working hours Jones				42					
8	Working hours Smith				43					



My wage is too low!

G	H	I	J
Total	Overtime	\$/h	Total wage
=SUM(B2:F2)	=IF(G2>E\$6;G2-E6;0)	20	=I2*(G2+H2*0,5)
=SUM(B3:F3)	=IF(G3>E\$6;G3-E7;0)	24	=I3*(G3+H3*0,5)
=SUM(G2:G3)	=SUM(H2:H3)		

Running Example

	A	B	C	D	E	F	G	H	I	J
1	Worker	Mo	Tu	We	Th	Fr	Total	Overtime	\$/h	Total wage
2	Jones	9	9	8	8	8	=SUM(B2:F2)	=IF(G2>E\$6;G2-E6;0)	20	=I2*(G2+H2*0,5)
3	Smith	9	9	10	7	8	=SUM(B3:F3)	=IF(G3>E\$6;G3-E7;0)	24	=I3*(G3+H3*0,5)
4	Total						=SUM(G2:G3)	=SUM(H2:H3)		
5										
6	Regular working time				40					
7	Working hours Jones				=G2					
8	Working hours Smith				=G3					

$$\text{CONE}(c) = c \cup \bigcup_{c' \in \rho(c)} \text{CONE}(c')$$

$\text{CONE}(J3) = \{E7, G2, G3, H3, J3\}$ $\text{CONE}(H4) = \{E7, G2, G3, H2, H3, H4\}$

$\text{CONE}(G4) = \{G2, G3, G4\}$

$\text{CONE}(J2) = \{E7, G2, G3, H3, J3\}$

Spectrum-based Fault Localization with the Ochiai Similarity Coefficient

$$SC_O(c) = \frac{a_{11}(c)}{\sqrt{(a_{11}(c) + a_{10}(c))(a_{11}(c) + a_{01}(c))}}$$

$$a_{11}(c) = |\{(c', d) \in O | c \in \text{CONE}(c') \wedge d = \text{false}\}|$$

$$a_{10}(c) = |\{(c', d) \in O | c \in \text{CONE}(c') \wedge d = \text{true}\}|$$

$$a_{01}(c) = |\{(c', d) \in O | c \notin \text{CONE}(c') \wedge d = \text{false}\}|$$

 $a_{11}, a_{10}, a_{01} \geq 0$

IF $(a_{11}(c) + a_{10}(c) = 0$ OR $a_{11}(c) + a_{01}(c) = 0)$
then $SC_O(c) = 0$

Running Example

Cell	G4	H4	J2	J3
E7		•		•
E8				
G2	•	•	•	•
G3	•	•		•
G4	•			
H2		•	•	
H3		•		•
H4		•		
J2			•	
J3				•
Classification d	true	false	true	false

Running Example

Cell	G4	H4	J2	J3	SC	Rank
E7		•		•	1.000	1
E8					0	-
G2	•	•	•	•	0.707	4
G3	•	•		•	0.816	3
G4	•				0	-
H2		•	•		0.500	7
H3		•		•	1.000	1
H4		•			0.707	4
J2			•		0	-
J3				•	0.707	4
Classification d	true	false	true	false		

To err is human.

Assuming a 3-5 % error rate

26-40 %

chance to make at least 1 mistake
when classifying 10 cell values.

Never assume **perfect** user knowledge
when **evaluating** fault localization
techniques for **end-user** software.

**What are the impacts
of a misclassification?**

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Theoretical analysis of the impact of a misclassification on the similarity coefficient

Misclassification of cell c as		
	erroneous	correct
$c_1 \in \text{CONE}(c)$	increase	
$c_2 \notin \text{CONE}(c)$		

Premises:

$$a_{11} + a_{10} > 0$$

$$a_{11} + a_{01} > 0$$

$$SC_O^\#(c_1) = \frac{a_{11} + 1}{\sqrt{(a_{11} + 1 + a_{10} - 1)(a_{11} + 1 + a_{01})}}$$

$$= \frac{a_{11} + 1}{\sqrt{(a_{11} + a_{10})(a_{11} + a_{01} + 1)}}$$

$$SC_O^\#(c_1) > SC_O^*(c_1)$$

Theoretical analysis of the impact of a misclassification on the similarity coefficient

	Misclassification of cell c as	
	erroneous	correct
$c_1 \in \text{CONE}(c)$	a_{11}^{++} a_{10}^{--}	a_{11}^{--} a_{10}^{++}
$c_2 \notin \text{CONE}(c)$	a_{01}^{++}	a_{01}^{--}

Theoretical analysis of the impact of a misclassification on the similarity coefficient

		Misclassification of cell c as	
		erroneous	correct
$c_1 \in \text{CONE}(c)$		increase	decrease or equal
$c_2 \notin \text{CONE}(c)$		decrease*	increase**

*Special case $a_{11} + a_{10} = 0 \vee a_{11} + a_{01} = 0$:

$$SC_O^*(c_2) = SC_O^\#(c_2) = 0$$

**Special case $a_{11} + a_{10} = 0 \vee a_{11} + a_{01} = 1$:

$$SC_O^*(c_2) = SC_O^\#(c_2) = 0$$

Influence of Misclassification in Running Example

Cell	G4	H4	J2	J3	SC	Rank
E7		•		•	0.816	3
E8					0	-
G2	•	•	•	•	0.866	2
G3	•	•		•	1.000	1
G4	•				0.577	5
H2		•	•		0.408	8
→ H3		•		•	0.816	3
H4		•			0.577	5
J2			•		0	-
J3				•	0.577	5
Classification d	false	false	true	false		

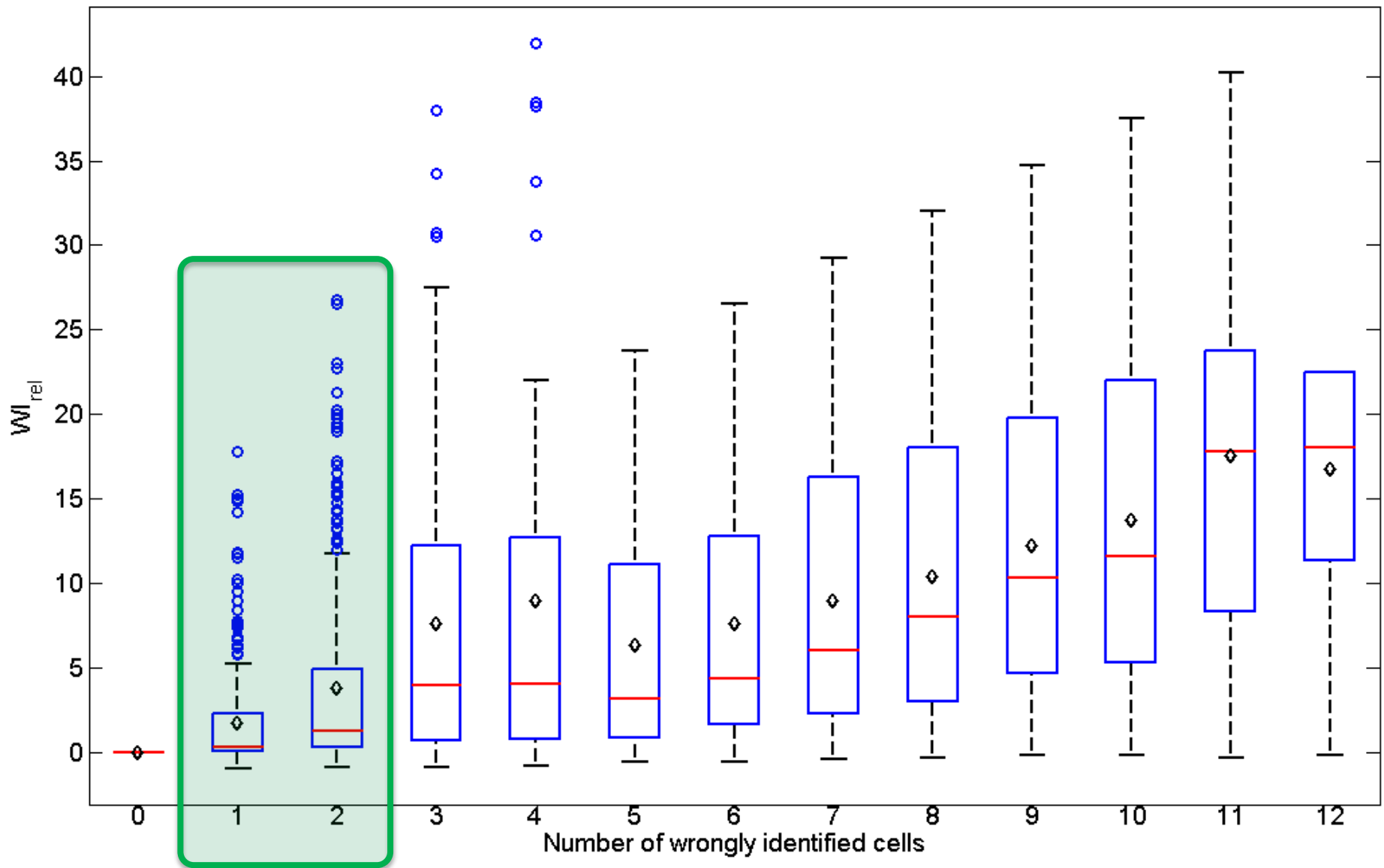
Empirical evaluation

- Integer spreadsheet corpus
 - 218 faulty spreadsheets
- Relative Worsening/Improvement

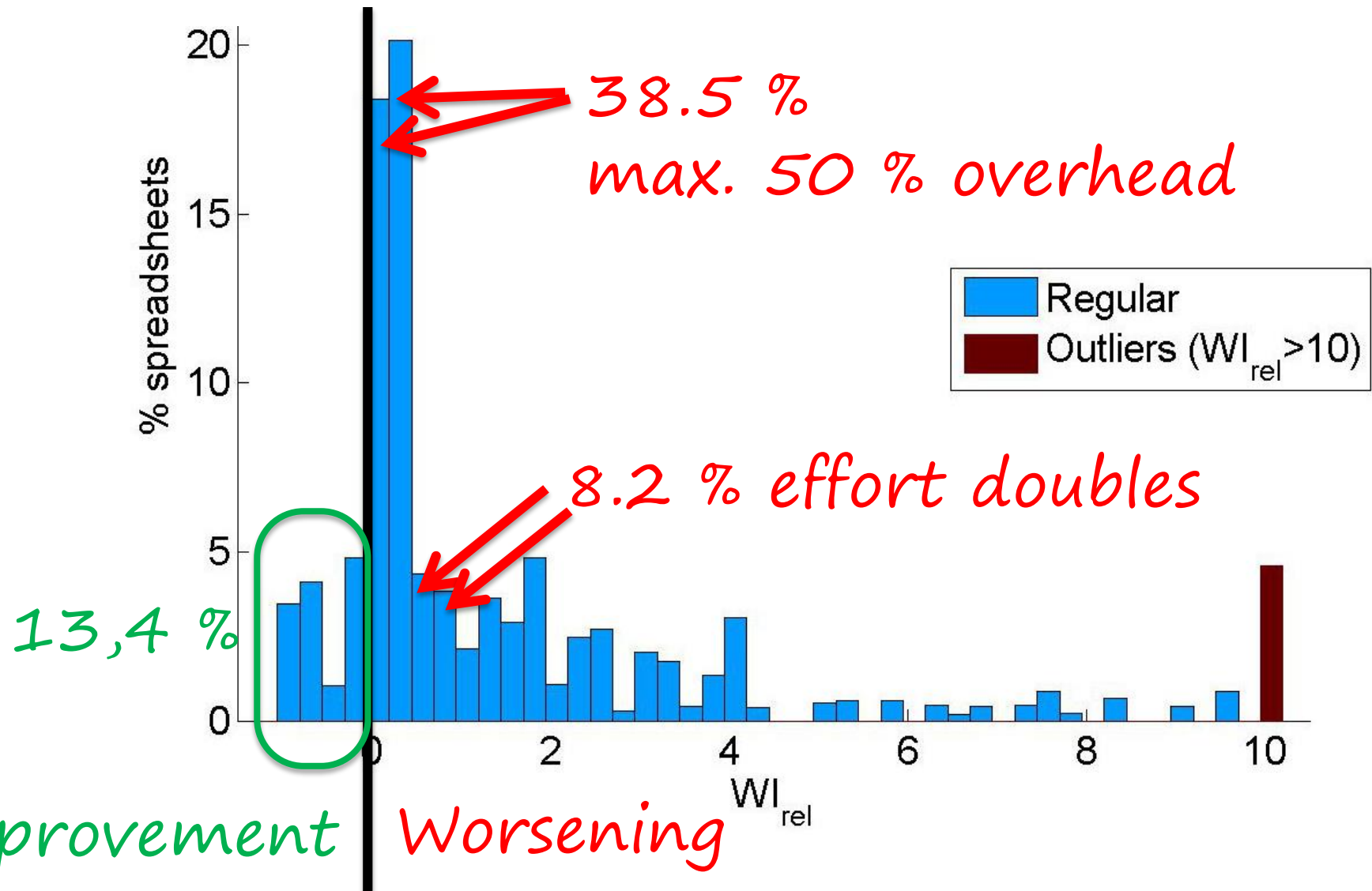
$$WI_{\text{rel}} = \frac{R_{\text{avg}}}{R_{\text{avg}}^*} - 1$$

$$R_{\text{avg}} = |\{i | SC_O(i) > SC_O(f)\}| + \\ + \frac{|\{j | SC_O(j) = SC_O(f)\}|}{2} + 0.5$$

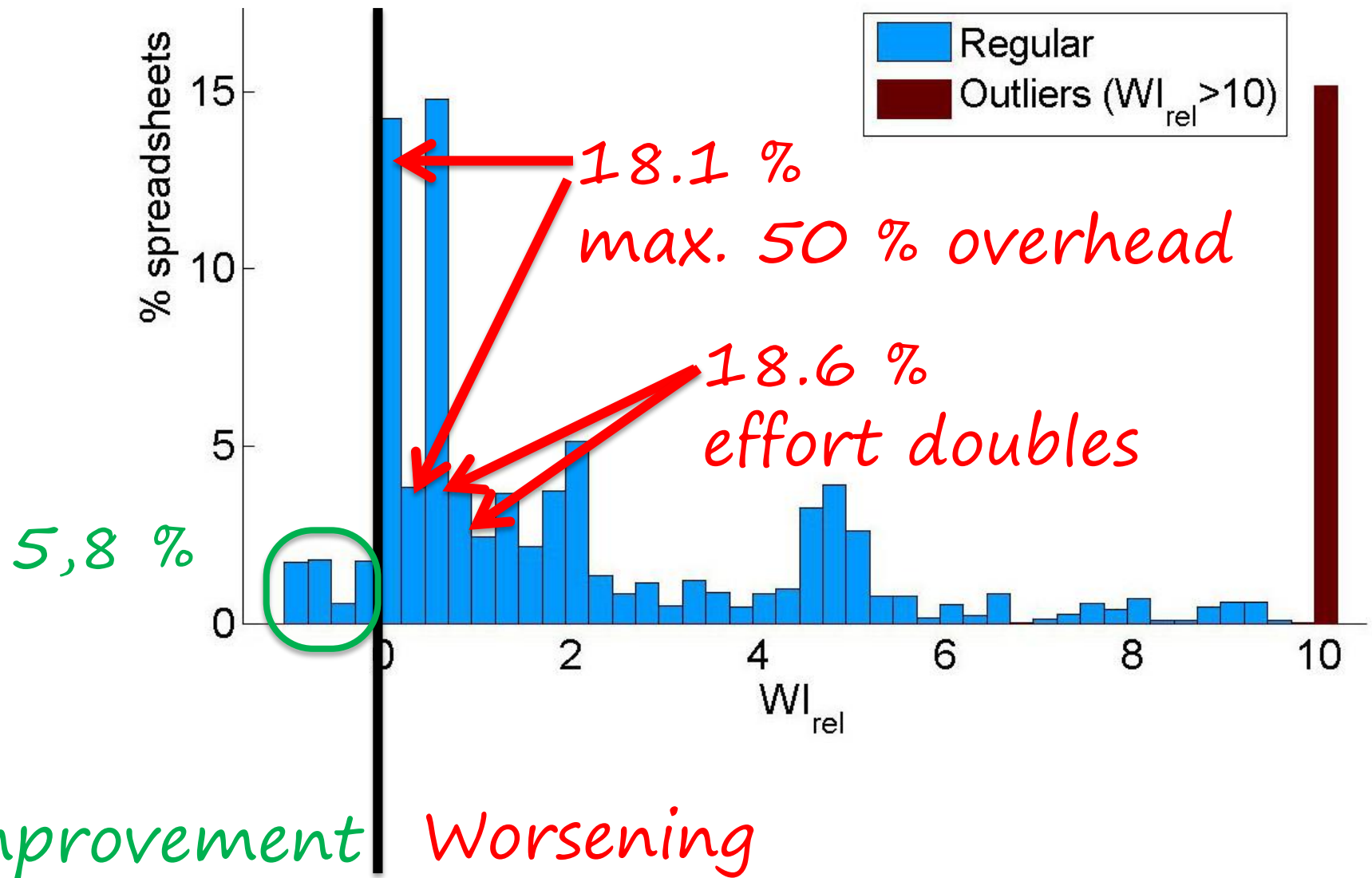
Relative Worsening/Improvement



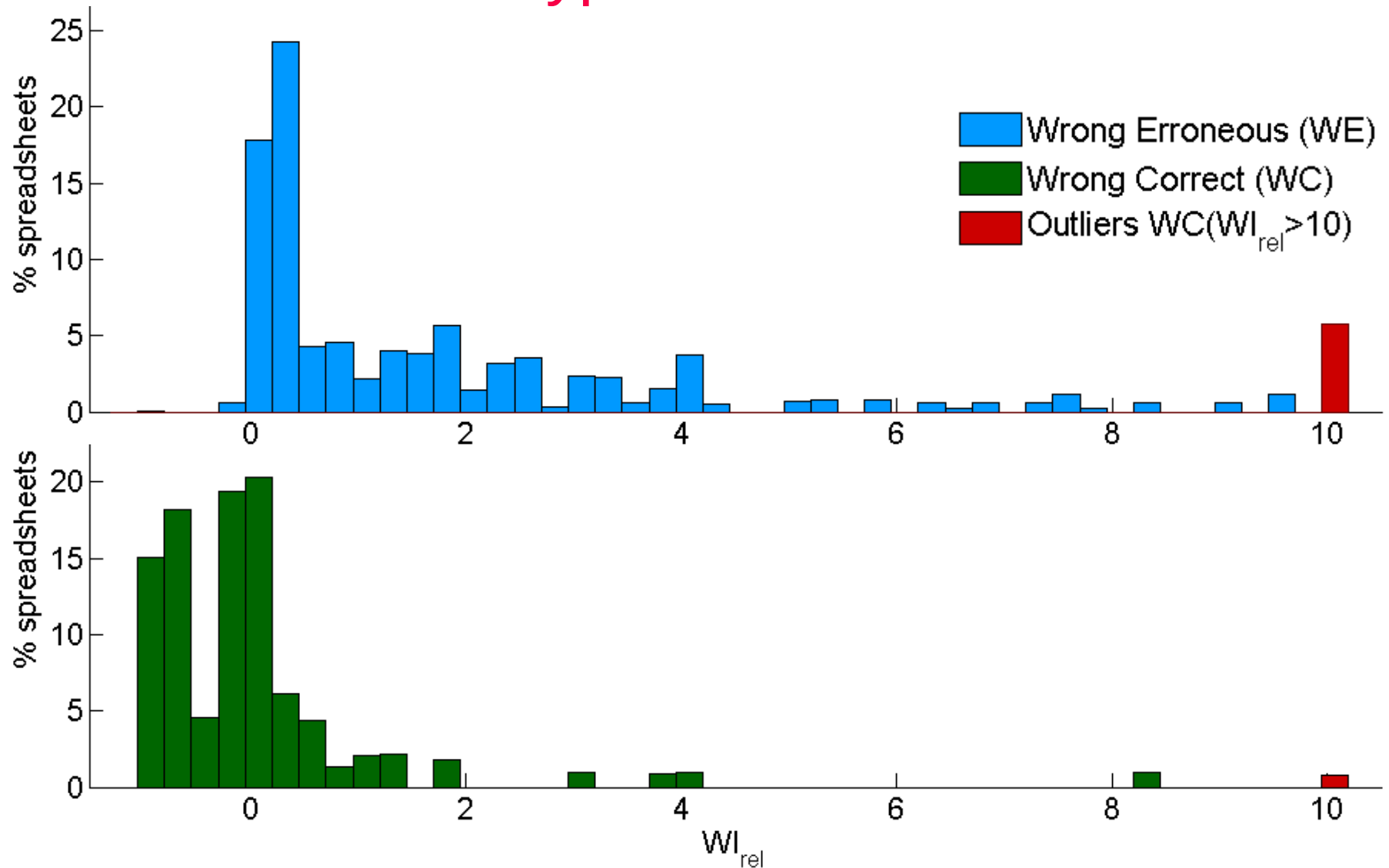
Effect histogram for 1 misclassification



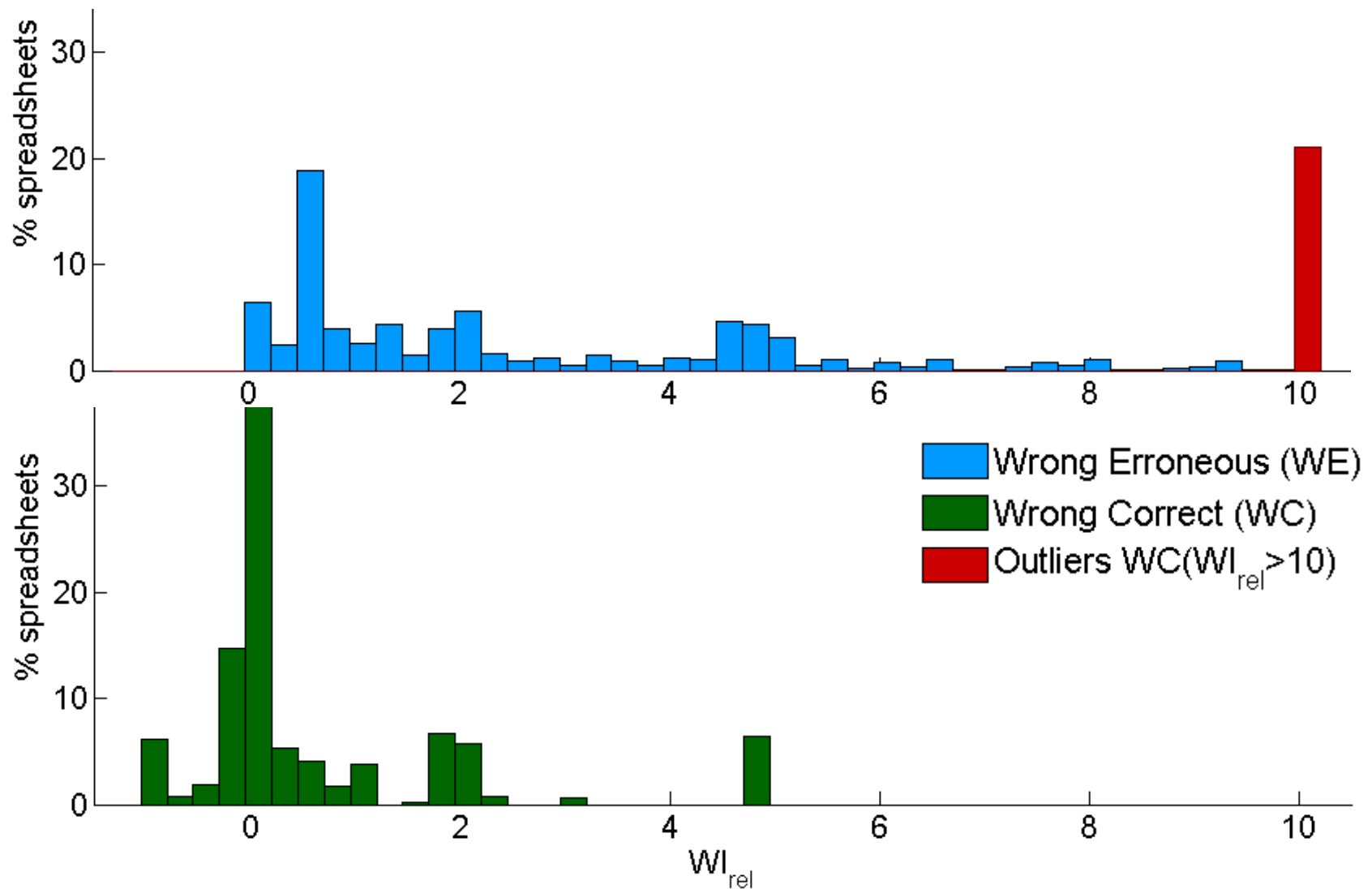
Effect histogram for 2 misclassifications



Effect histogram for 1 misclassification for different types of misclassifications



Effect histogram for 2 misclassifications for different types of misclassifications



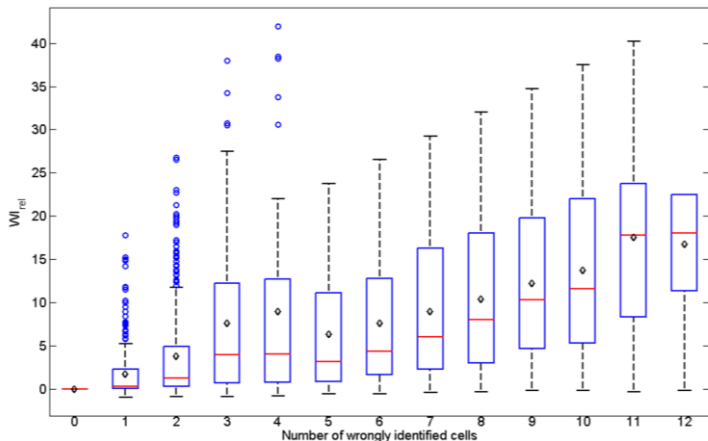
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3-5 % error rate when writing formulas*

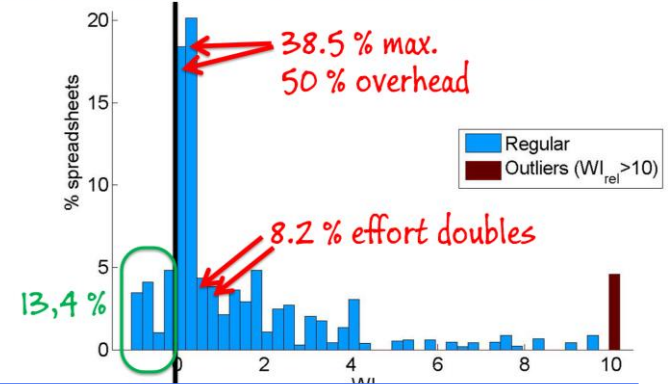
Theoretical analysis of the impact of a misclassification on the similarity coefficient

Misclassification of cell c as		
	erroneous	correct
$c_1 \in \text{CONE}(c)$	increase	decrease or equal
$c_2 \notin \text{CONE}(c)$	decrease*	increase**

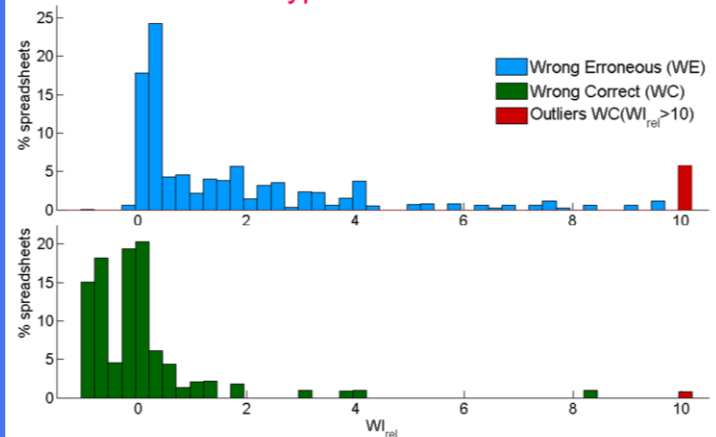
Relative Worsening/Improvement



Effect histogram for 1 misclassification



Effect histogram for 1 misclassification for different types of misclassifications



Thank you!
Questions?