

Additional Material

Learning to classify software defects from crowds: a novel approach

Jerónimo Hernández-González^{a,*}, Daniel Rodriguez^c, Iñaki Inza^a, Rachel
Harrison^d, Jose A. Lozano^{a,b}

^a Department of Computer Science and Artificial Intelligence, University of the Basque
Country UPV/EHU, Donostia, Spain

^b Basque Center for Applied Mathematics BCAM, Bilbao, Spain

^c Department of Computer Science, University of Alcalá, Madrid, Spain

^d Department of Computing, Oxford Brookes University, Oxford, UK

Abstract

Example of the computation of the formulae of the Section 3 of the paper.

*Corresponding author

Email addresses: jeronimo.hernandez@ehu.eus (Jerónimo Hernández-González),
daniel.rodriguez@uah.es (Daniel Rodriguez), inaki.inza@ehu.eus (Iñaki Inza),
rachel.harrison@brookes.ac.uk (Rachel Harrison), ja.lozano@ehu.eus (Jose A. Lozano)

1. Estimation of model parameters

Table 1: Example of the annotations of 3 labelers for 4 different examples
 $\mathcal{L}^1 = \{\text{Instal., Other, Instal.}\}$ $\mathcal{L}^2 = \{\text{Req., Req., Req.}\}$
 $\mathcal{L}^3 = \{\text{Req., Usab., Other}\}$ $\mathcal{L}^4 = \{\text{Usab., Usab., Req.}\}$

Table 2: Example of per-label weights (w_c^a) for the 3 annotators.
 $A1 : \{0.7, 0.8, 0.6, 0.7\}$ $A2 : \{0.8, 0.6, 0.7, 0.6\}$ $A3 : \{0.7, 0.7, 0.6, 0.6\}$

Table 3: Computation of Eq. 3 for examples of Table 1 using per-label weights of Table 2

$$\begin{aligned} F_{Inst}^{l^1} &= \frac{1 \cdot 0.7 + 0 \cdot 0.8 + 1 \cdot 0.7}{1 \cdot 0.7 + 0 \cdot 0.8 + 1 \cdot 0.7 + 0 \cdot 0.8 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 1 \cdot 0.6 + 0 \cdot 0.6} = \frac{0.7 + 0.7}{0.7 + 0.7 + 0.6} = 0.7 \\ F_{Req}^{l^1} &= \frac{0 \cdot 0.8 + 0 \cdot 0.6 + 0 \cdot 0.7}{1 \cdot 0.7 + 0 \cdot 0.8 + 1 \cdot 0.7 + 0 \cdot 0.8 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 1 \cdot 0.6 + 0 \cdot 0.6} = \frac{0.0}{0.7 + 0.7 + 0.6} = 0.0 \\ F_{Usab}^{l^1} &= \frac{0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6}{1 \cdot 0.7 + 0 \cdot 0.8 + 1 \cdot 0.7 + 0 \cdot 0.8 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 1 \cdot 0.6 + 0 \cdot 0.6} = \frac{0.0}{0.7 + 0.7 + 0.6} = 0.0 \\ F_{Other}^{l^1} &= \frac{0 \cdot 0.7 + 1 \cdot 0.6 + 0 \cdot 0.6}{1 \cdot 0.7 + 0 \cdot 0.8 + 1 \cdot 0.7 + 0 \cdot 0.8 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 0 \cdot 0.6 + 0 \cdot 0.7 + 1 \cdot 0.6 + 0 \cdot 0.6} = \frac{0.6}{0.7 + 0.7 + 0.6} = 0.3 \end{aligned}$$

c	$F_c^{l^2}$	$F_c^{l^3}$	$F_c^{l^4}$
$Inst.$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.0}{0.8+0.7+0.6} = 0$	$\frac{0.0}{0.6+0.7+0.7} = 0$
$Req.$	$\frac{0.8+0.6+0.7}{0.8+0.6+0.7} = 1$	$\frac{0.8}{0.8+0.7+0.6} = 0.38$	$\frac{0.7}{0.6+0.7+0.7} = 0.35$
$Usab.$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.7}{0.8+0.7+0.6} = 0.33$	$\frac{0.6+0.7}{0.6+0.7+0.7} = 0.65$
$Other$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.6}{0.8+0.7+0.6} = 0.29$	$\frac{0.0}{0.6+0.7+0.7} = 0$

Table 4: Example of confusion-matrix weights ($W_{cc'}^a$) for the 3 annotators. Note that the diagonals of the matrices are the same per-label weights of Table 2.

$$A1 : \begin{bmatrix} 0.7 & 0.2 & 0.1 & 0.0 \\ 0.1 & 0.8 & 0.1 & 0.0 \\ 0.1 & 0.2 & 0.6 & 0.1 \\ 0.2 & 0.0 & 0.1 & 0.7 \end{bmatrix} \quad A2 : \begin{bmatrix} 0.8 & 0.1 & 0.0 & 0.1 \\ 0.1 & 0.6 & 0.1 & 0.2 \\ 0.0 & 0.2 & 0.7 & 0.1 \\ 0.2 & 0.0 & 0.2 & 0.6 \end{bmatrix} \quad A3 : \begin{bmatrix} 0.7 & 0.2 & 0.1 & 0.0 \\ 0.1 & 0.7 & 0.1 & 0.1 \\ 0.2 & 0.2 & 0.6 & 0.0 \\ 0.1 & 0.2 & 0.1 & 0.6 \end{bmatrix}$$

Table 5: Computation of Eq. 4 for examples of Table 1 using confusion-matrix weights of Table 4

c	$F_c^{l^2}$	$F_c^{l^3}$	$F_c^{l^4}$
$Inst.$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.0}{0.8+0.2+0.2+0.1+0.7+0.1+0.0+0.1+0.6} = 0$	$\frac{0.0}{0.2+0.2+0.7+0.6+0.7+0.1} = 0$
$Req.$	$\frac{0.8+0.6+0.7}{0.8+0.6+0.7} = 1$	$\frac{0.8+0.2+0.2}{0.8+0.2+0.2+0.1+0.7+0.1+0.0+0.1+0.6} = 0.43$	$\frac{0.2+0.2+0.7}{0.2+0.2+0.7+0.6+0.7+0.1} = 0.44$
$Usab.$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.1+0.7+0.1}{0.8+0.2+0.2+0.1+0.7+0.1+0.0+0.1+0.6} = 0.32$	$\frac{0.6+0.7+0.1}{0.2+0.2+0.7+0.6+0.7+0.1} = 0.56$
$Other$	$\frac{0.0}{0.8+0.6+0.7} = 0$	$\frac{0.0+0.1+0.6}{0.8+0.2+0.2+0.1+0.7+0.1+0.0+0.1+0.6} = 0.25$	$\frac{0.0}{0.2+0.2+0.7+0.6+0.7+0.1} = 0$

2. Estimation of reliability weights for the annotators

Table 6: Computation of per-label weights (w_c^a , Eq. 5) for annotations of Table 1.

c	L_1	L_2	L_3
<i>Inst.</i>	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{1} = 0.5$	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$
<i>Req.</i>	$\frac{1 \cdot \frac{1}{2} \cdot (1+1) + 1 \cdot \frac{1}{2} \cdot (0+0)}{2} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (1+1)}{1} = 1.0$	$\frac{1 \cdot \frac{1}{2} \cdot (1+1) + 1 \cdot \frac{1}{2} \cdot (0+0)}{2} = 0.5$
<i>Usab.</i>	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (0+0) + 1 \cdot \frac{1}{2} \cdot (1+0)}{2} = 0.25$	0.0
<i>Other</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (0+0)}{1} = 0.0$	$\frac{1 \cdot \frac{1}{2} \cdot (0+0)}{1} = 0.0$

Table 7: Computation of confusion-matrix weights ($W_{cc'}^a$, Eq. 6) for annotations of Table 1.

L_1 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{1} = 0.5$	0.0	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$
<i>Req.</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+1) + 1 \cdot \frac{1}{2} \cdot (0+0)}{2} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{2} = 0.25$	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{2} = 0.25$
<i>Usab.</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{1} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$	0.0
<i>Other</i>	0.0	0.0	0.0	0.0

L_2 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	0.0	0.0	0.0	0.0
<i>Req.</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+1)}{1} = 1.0$	0.0	0.0
<i>Usab.</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+0) + 1 \cdot \frac{1}{2} \cdot (0+1)}{2} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{2} = 0.25$	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{2} = 0.25$
<i>Other</i>	$\frac{1 \cdot \frac{1}{2} \cdot (1+1)}{1} = 1.0$	0.0	0.0	0.0

L_3 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$	0.0	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{1} = 0.5$
<i>Req.</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+1) + 1 \cdot \frac{1}{2} \cdot (0+0)}{2} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (1+1)}{2} = 0.5$	0.0
<i>Usab.</i>	0.0	0.0	0.0	0.0
<i>Other</i>	0.0	$\frac{1 \cdot \frac{1}{2} \cdot (1+0)}{1} = 0.5$	$\frac{1 \cdot \frac{1}{2} \cdot (0+1)}{1} = 0.5$	0.0

3. Re-estimating weights

Table 8: Example of probability distributions ($p_{\mathbb{M}}(c|\mathbf{x})$) and predictions ($\arg \max_c p_{\mathbb{M}}(c|\mathbf{x})$) given by a model \mathbb{M} for examples of Table 1.

\mathbf{x}	Prediction	$p_{\mathbb{M}}(c \mathbf{x})$			
		Inst.	Req.	Usab.	Other
\mathbf{x}^1	Inst.	0.7	0.1	0.0	0.2
\mathbf{x}^2	Req.	0.1	0.6	0.1	0.2
\mathbf{x}^3	Usab.	0.1	0.3	0.4	0.2
\mathbf{x}^4	Req.	0.2	0.5	0.3	0.0

Table 9: Computation of per-label weights (w_c^a , Eq. 7) for annotations of Table 1 and model results of Table 8.

Accuracy-based strategy:

c	L_1	L_2	L_3
Inst.	$\frac{1}{1} = 1.0$	0.0	$\frac{1}{1} = 1.0$
Req.	$\frac{1}{2} = 0.5$	$\frac{1}{1} = 1.0$	$\frac{2}{2} = 1.0$
Usab.	$\frac{0}{1} = 0.0$	$\frac{1}{2} = 0.5$	0.0
Other	0.0	$\frac{0}{1} = 0.0$	$\frac{0}{1} = 0.0$

Probability-based strategy:

c	L_1	L_2	L_3
Inst.	$\frac{0.7}{1} = 0.7$	0.0	$\frac{0.7}{1} = 0.7$
Req.	$\frac{0.6+0.3}{2} = 0.45$	$\frac{0.6}{1} = 0.6$	$\frac{0.6+0.5}{2} = 0.55$
Usab.	$\frac{0.3}{1} = 0.3$	$\frac{0.4+0.3}{2} = 0.35$	0.0
Other	0.0	$\frac{0.2}{1} = 0.2$	$\frac{0.2}{1} = 0.2$

Table 10: Computation with *accuracy-based strategy* of confusion-matrix weights ($W_{cc'}^a$, Eq. 8) for annotations of Table 1 and model results of Table 8.

$L_1:$				
$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{1}{1} = 1.0$	0.0	0.0	0.0
<i>Req.</i>	0.0	$\frac{1}{2} = 0.5$	$\frac{1}{2} = 0.5$	0.0
<i>Usab.</i>	0.0	$\frac{1}{1} = 1.0$	$\frac{0}{1} = 0.0$	0.0
<i>Other</i>	0.0	0.0	0.0	0.0

$L_2:$				
$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	0.0	0.0	0.0	0.0
<i>Req.</i>	0.0	$\frac{1}{1} = 1.0$	0.0	0.0
<i>Usab.</i>	0.0	$\frac{1}{2} = 0.5$	$\frac{1}{2} = 0.5$	0.0
<i>Other</i>	$\frac{1}{1} = 1.0$	0.0	0.0	$\frac{0}{1} = 0.0$

$L_3:$				
$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{1}{1} = 1.0$	0.0	0.0	0.0
<i>Req.</i>	0.0	$\frac{2}{2} = 1.0$	0.0	0.0
<i>Usab.</i>	0.0	0.0	0.0	0.0
<i>Other</i>	0.0	0.0	$\frac{1}{1} = 1.0$	$\frac{0}{1} = 0.0$

Table 11: Computation with *probability-based strategy* of confusion-matrix weights ($W_{cc'}^a$, Eq. 8) for annotations of Table 1 and model results of Table 8.

L_1 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{0.7}{1} = 0.7$	$\frac{0.1}{1} = 0.1$	$\frac{0.0}{1} = 0.0$	$\frac{0.2}{1} = 0.2$
<i>Req.</i>	$\frac{0.1+0.1}{2} = 0.1$	$\frac{0.6+0.3}{2} = 0.45$	$\frac{0.1+0.4}{2} = 0.25$	$\frac{0.2+0.2}{2} = 0.2$
<i>Usab.</i>	$\frac{0.2}{1} = 0.2$	$\frac{0.5}{1} = 0.5$	$\frac{0.3}{1} = 0.3$	$\frac{0.0}{1} = 0.0$
<i>Other</i>	0.0	0.0	0.0	0.0

L_2 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	0.0	0.0	0.0	0.0
<i>Req.</i>	$\frac{0.1}{1} = 0.1$	$\frac{0.6}{1} = 0.6$	$\frac{0.1}{1} = 0.1$	$\frac{0.2}{1} = 0.2$
<i>Usab.</i>	$\frac{0.1+0.2}{2} = 0.15$	$\frac{0.3+0.5}{2} = 0.4$	$\frac{0.4+0.3}{2} = 0.35$	$\frac{0.2+0.0}{2} = 0.1$
<i>Other</i>	$\frac{0.7}{1} = 0.7$	$\frac{0.1}{1} = 0.1$	$\frac{0.0}{1} = 0.0$	$\frac{0.2}{1} = 0.2$

L_3 :

$c \setminus c'$	<i>Inst.</i>	<i>Req.</i>	<i>Usab.</i>	<i>Other</i>
<i>Inst.</i>	$\frac{0.7}{1} = 0.7$	$\frac{0.1}{1} = 0.1$	$\frac{0.0}{1} = 0.0$	$\frac{0.2}{1} = 0.2$
<i>Req.</i>	$\frac{0.1+0.2}{2} = 0.15$	$\frac{0.6+0.5}{2} = 0.55$	$\frac{0.1+0.3}{2} = 0.2$	$\frac{0.2+0.0}{2} = 0.1$
<i>Usab.</i>	0.0	0.0	0.0	0.0
<i>Other</i>	$\frac{0.1}{1} = 0.1$	$\frac{0.3}{1} = 0.3$	$\frac{0.4}{1} = 0.4$	$\frac{0.2}{1} = 0.2$