POSTER: Enabling Fair ML Evaluations for Security {References for Poster Timeline}

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1 LIST OF REFERENCES

Table 1 reports the list of papers that may violate constraints C1, C2 and C3 that we describe in [37]. This is an extended and updated version of the list contained in [39], and applied to the constraints of our Tesseract framework [37]. The list considers papers published in the last 10 years but is not intended to be exhaustive. Please do not hesitate to contact us if you believe if some \checkmark or \nearrow need changes.

Year	Venue	C1	C2	C3	Domain	Notes
2018	USENIX Sec [33]	X	X	-	Tunneling	We were not able to assess C3 from
						the paper.
2018	USENIX Sec [38]	X	×	 Image: A start of the start of	Android	
2018	NDSS [35]	X	X	✓	Vulnerabilities	
2018	S&P [34]	X	X	✓	Scams	
2017	NDSS [31]	1	X	X	Android	
2017	S&P [28]	X	X	✓	IP Domains	
2017	ESORICS [30]	X	X	✓	Android	
2017	CODASPY [32]	×	X	✓	Android	
2017	TDSC [29]	×	×	-	Android	C3 does not apply as they consider
						only malware clustering.
2016	CCS [27]	X	X	 Image: A start of the start of	Android	
2016	NDSS [24]	X	X	-	PDFs	We were not able to assess C3 from
						the paper.
2016	NDSS [26]	X	X	×	Bots	
2016	S&P-MoST [25]	X	×	-	Android	C3 does not apply as they consider
						only malware clustering.
2015	CCS [23]	X	X	×	Twitter	
2015	NDSS [22]	X	X	 Image: A start of the start of	Fake Accounts	
2014	NDSS [20]	X	X	 Image: A start of the start of	Android	
2014	WATeR [21]	X	×	 Image: A set of the set of the	x86	
2013	CCS [19]	X	X	 Image: A start of the start of	URLs	
2013	CCS-AISec [17]	×	×	 Image: A start of the start of	Android	
2013	ICASSP [16]	×	X	 Image: A start of the start of	x86	
2013	NDSS [18]	1	-	×	PDF	We were not able to assess C2 from
						the paper.
2012	JMLR [14]	×	×	×	x86	

Table 1. References for papers in the timeline of poster [36]. We report papers between 2009 and 2018.

Year	Venue	C1	C2	C3	Domain	Notes
2012	RAID [15]	X	X	 Image: A start of the start of	ActionScript	-
2012	ISSTA [12]	X	X	-	x86	We were not able to assess C3 from
						the paper.
2012	ACSAC [13]	1	-	 Image: A set of the set of the	PDF	We were not able to assess C2 from
						the paper.
2011	ACSAC [11]	X	X	 Image: A second s	PDF	
2011	USENIX Sec [9]	X	X	 Image: A set of the set of the	JavaScript	
2011	RAID [10]	X	X	 Image: A second s	JavaScript	
2011	WWW [8]	X	X	 Image: A second s	URLs	
2010	ESSoS [7]	X	X	 Image: A second s	x86	
2010	ACSAC [6]	X	X	 Image: A set of the set of the	JavaScript	
2010	NSDI [5]	1	-	 Image: A second s	Traffic	We were not able to assess C2 from
						the paper.
2010	WWW [4]	X	X	 Image: A start of the start of	JavaScript	
2009	MALWARE [3]	X	X	 Image: A start of the start of	x86	
2009	RAID [2]	X	X	 Image: A set of the set of the	x86	
2009	ICC [1]	X	X	X	Android	

Table 1. References for papers in the timeline of poster [36]. We report papers between 2009 and 2018.

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