

Artificial Intelligence an Essential Expected Computer World Surveillance

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Abstract

A position paper toward an important and urgent discussion on how best uses the potential of Artificial Intelligence in the context of Computer World surveillance. AI is often cited in papers on Computer World surveillance. But what is meant is using pre-existing AI techniques in Computer World surveillance. AI techniques are established around applications. Computer World surveillance has never been an area of deliberation in AI. In this paper we argue that Computer World surveillance calls for new and specific AI techniques developed with that kind of application in mind. In practice, this paper is based on a broad overview of different slants, which have the budding to be game changers in Computer World surveillance. This paper focuses on web solicitation security and supporters the use of Knowledge Based Systems, probabilistic reasoning and Bayesian apprising to control the probability of false positives and false denials.

Index terms— Documentation, Security, Theory. Keywords : Bayesian updating, CSRF, probabilistic reasoning.

1 Introduction

It has been known for a long time that in Computer World surveillance, defense uses a flawed hypothesis because it leads to a strategy based on tweaking and "fixing the comprehending" with no long term vision. When new forms of attacks appear, ad hoc response are put together, which often involves making the use of the internet more cumbersome, by adding layers of authentication. "Defensive measures tend to involve complicating protocols or their implementation, making things more secure by making them more cumbersome a mentioning abstaining from using some functionality, as more often than not each new functionality provides new points of entry for malicious activities [1], [2], [3], [4]. But the introductions of new functionalities are precisely what make the internet so attractive and successful. Furthermore there is still no tolerable defence to zero day attack, as anomaly based detection has still some open problems, like the false positive probability. An ideal cyber-defense would provide full protection to users, while preserving all the functionalities. We are very far from this situation. But there is no reason why in the long run, we could not get close to such a situation.

One thing that cyber-defense can do and should is to be more intelligent. The approach to defense based on "fixing the plumbing" is inherently suboptimal. A massive paradigm shift is needed, the kind of paradigm shift that makes a much heavier use of Artificial Intelligence (AI). The idea of making heavier use of AI in Computer World surveillance is not new. In an editorial in IEEE security and Privacy [5], Carl Landwehr stated that "In their early days, computer security and artificial intelligence didn't seem to have much to say to each other. AI researchers were interested in making computers do things that only humans had been able to do, while security researchers aimed to fix the leaks in the plumbing of the computing infrastructure or design infrastructures they deemed leak proof." But the dream of retroactively make the internet secure and leak-proof, is clearly not a clever approach.

The introduction of new technologies like the proliferation of new web applications or the increasing use of wireless, have exacerbated this fact [1]. Computer World surveillance, has become the most complex threat to

2 A) AN ILLUSTRATIVE EXAMPLE: CROSS SITE REQUEST FORGERY (CSRF)

45 society. Despite years of incremental improvements in cyber-defense, it is clear that a paradigm shift is needed,
46 but at the same time difficult to imagine. This is even more true for web-application security and the need for
47 AI is even more obvious and urgent there. Against web application attacks, such as Cross Site Scripting (XSS),
48 Cross Site Request Forgery (CSRF), injection code, the present approach consists in introducing rules supposed
49 to prevent them. This is the same kind of logic that un derived the idea of same origin policy. Over the years
50 XSS and CSRF began to mean a variety of attacks. Some of them can be construed as direct circumvention of
51 the same origin policy. Same origin policy looked like a simple and efficient protection.

52 It turned out that it could be circumvented reasonably easily and was preventing some functionality to modern
53 websites. In the words of D. Crockford same origin policy (which is adopted by most browsers) "prevents useful
54 things and allows dangerous ones" [7]). Today, this policy is being revisited. For example, crosssite access control
55 [8] is an attempt to refine the security policy in such a way that one can get the benefit of cross site access
56 without the security implications.

57 In the same way, SQL injection codes were described as "extremely easy to avoid" [9]. Still recently MYSQL
58 and Oracle, both suffered such attacks. What may be construed as mistakes is also reflection of the fact that
59 those mistakes are not so easy to avoid in the increasingly complicated world of web applications. To detect web
60 application attacks such as XSS, CSRF or injection codes requires more than simple rules, but the ability to
61 some form of context dependent reasoning. Despite some work done in the past, AI does not play central role
62 in Computer World surveillance today and Computer World surveillance has not been an area of development
63 of AI as intensely pursued as robots, machine learning and the like. Typically, the use of AI in Computer World
64 surveillance has consisted in using some tools developed in AI and apply them to intrusion detection or other
65 aspects of Computer World surveillance. The approach consisting in importing some AI techniques developed in
66 totally different areas and try to apply them to Computer World surveillance, may work in a few cases, but has
67 inherent and severe limitations. AI has been developed and is organized around specific applications, many of
68 them (AI is a large field). Computer World surveillance has specific needs and to be a long term contributor to
69 Computer World surveillance, new AI techniques will have to be developed specifically.

70 Obviously AI has made a lot of accomplishments and there is a lot to be learned relevant for Computer World
71 surveillance and many new techniques suited for Computer World surveillance could be inspired from existing
72 ones in AI. In a sense this has happened already. As observed by C. Landwehr [5]: "A branch of AI that has
73 been connected with computer security from relatively early days is automated reasoning, particularly as applied
74 to programs and systems. [...] Although it wasn't identified as AI at the time, Dan Farmer and Wietse Venema's
75 SATAN program, released in 1995, automated a process for finding vulnerabilities in system configurations that
76 had previously required much more human effort." S. Forrest [10] has proposed a system of inductive reasoning,
77 which belongs to the realm of AI. And one can interpret the work of the group of Vigna et al in UCSB [11], [12],
78 [13], [14], [15]) as proceeding from a somewhat similar philosophy. Arguably Web application firewalls (WAFs)
79 or more generally firewalls using deep packet inspections can be construed as a kind of instantiation of AI in
80 security. Firewalls have been part of the arsenal of cyber-defense for many years. Although more sophisticated
81 techniques are also used, in most cases the filtering is based on port numbers. WAFs cannot rely on port number
82 as most web applications use the same port as the rest of the web traffic. Deep packet inspection is the only
83 option for WAFs to be able to tell a malicious application from a legitimate one. The idea of filtering at the
84 application layer was introduced already in the third generation of firewalls in the 1990's. WAFs can be construed
85 as a special case of application layer firewalls. The modest success of those technologies reflects the need for far
86 more work on AI before they can make significant difference Computer World surveillance.

87 But those examples show that AI has a natural role to play in Computer World surveillance. They also show
88 that using AI is inherently difficult. Introducing tools whose reaction to attacks is not totally predictable as they
89 involve some context dependent reasoning will force attackers to potentially change completely their strategy.
90 The increasing role of AI is in the logic of the modernization of the web and the internet. Web 2.0, the semantic
91 web, the use of first order logic, the development of OWL are as many evidences of that. These developments
92 will raise the level of complexity of Computer World surveillance and force it to be much more sophisticated.
93 Computer World surveillance may turn out to be one of the best areas of applications of AI.

94 2 a) An Illustrative Example: Cross Site Request Forgery 95 (CSRF)

96 This paragraph is meant to make the following discussion a bit less "high level" or purely abstract, by providing
97 an example around which the rest of the discussion can be organized. The example is Cross Site Request forgery
98 (CSRF). CSRF is not new. In 1988 it was known as "confused deputy". Dubbed a "sleeping giant", it came to
99 prominence (i.e. enter in the OWASP top ten list) in the last few years. In the same way that XSS covers a
100 large spectrum of attacks or vulnerabilities, some of which justifies to be called "cross site scripting", CSRF has
101 grown into an open ended class of attacks. What all these attacks seem to have in common is that they hijack
102 some credentials from a user and use them for the advantage of the attackers. Of this large class of attacks,
103 what follows applies only to a subclass: it is when the attack takes place within the browser of the user. In
104 fact the following considerations would apply to all situations where the attacks take place within the browser
105 of the user, such as the man in the browser. Would an expert monitoring each HTTP request be able in real

106 time to realize that a CSRF attack is unfolding? Some context dependent analysis is of essence. The decision of
107 whether malicious activity is taking place or not does not need to be the result of only one measurement. The
108 analysis can be protracted and based on a succession of observations. An AI machine using a Bayesian algorithm
109 could potentially do exactly the same. It is known that in a situation where the probability of false positive
110 and negative of individual measurements is not so small, a shrewd Bayesian algorithm, well implemented can
111 dramatically reduce the overall probability of false positive, while maintaining the probability of false negative
112 low [16]. In other words, it is(D D D D) D 2012

113 Year possible with a shrewd use of multi-observations to maintain the probability of mis-determinations very
114 low [16].

115 If one takes the example of the CSRF attack described in the classic paper of Felten ??17]. A user while
116 in a trusted session with his bank goes to a malicious website and click to download an image. The HTML
117 request is crafted in such a way that through the browser, the query ends going to the bank website and instead
118 of downloading a picture gives instructions (like transferring money or creating a new account) to the bank on
119 behalf of the victim user. A security tool monitoring the steps would find suspicious to have to find an image
120 at a bank using the trusted session. It would assign a reasonably small probability of false positive for that
121 (which could have been determined statistically before). Parsing the request, it would notice that it carries an
122 executable. This too would raise serious suspicion. Then it could figure in the executable corresponds to what
123 the bank requests for financial transaction. The probability of those three occurrences happening within the
124 same query is small enough to trigger an alert that has a very very small probability of being a false positive. At
125 each step the probability of false positive will be small, but different.

126 The point is that the tool would have to be able to do those inferences. It should be somewhat intelligent. We
127 now turn to the questions: How far is AI from being able to produce tools like that? And what can be done now
128 to facilitate the development of such tools? b) A Very Brief Discussion of AI Methods Although AI can be called a
129 discipline, it is so organized around many different applications that it almost looks like a vast fragmented world.
130 The example of CSRF points toward some form of probabilistic reasoning, as being a more natural approach to
131 deal with that kind of situation than an approach requiring a lot of data for statistical learning for example.

132 When it comes to probabilistic learning, it seems difficult to avoid contact with Bayesianism. Bayesian
133 reasoning is not without pitfalls, as Judea Pearl intimated in a recent presentation: "I turned Bayesian in 1971,
134 as soon as I began reading Savage's monograph The Foundations of Statistical Inference [18]. The arguments
135 were unassailable: (i) It is plain silly to ignore what we know, (ii) It is natural and useful to cast what we know in
136 the language of probabilities, and (iii) If our subjective probabilities are erroneous, their impact will get washed
137 out in due time, as the number of observations increases.

138 Thirty years later, I am still a devout Bayesian in the sense of (i), but I now doubt the wisdom of (ii) and I
139 know that, in general, (iii) is false." But (iii) may be false for humans, but not necessarily for machines, because
140 machines do not need to be prejudiced. Algorithms based on Bayesian updating work better with machines
141 than humans.... An additional reason to use a Bayesian approach, is that as we saw in the previous paragraph,
142 through Bayesian updating, it is possible to reduce the probability of false positive and negative. The decision
143 of whether malicious activity is taking place or not does not need to be the result of only one measurement. It
144 can be the result of analyzing a succession of steps.

145 This vision of a system able to process fast a lot of information, maintaining the rate of false positive and
146 false negative small, despite the fact that the individual components themselves can have a high rate of false
147 positive or false negative is reminiscent of the original idea of von Neumann discussed in his 1956 paper entitled:
148 "Probabilistic logics and the synthesis of reliable organisms from unreliable components" [19].

149 As can be seen in the CSRF example, this probabilistic reasoning supposes the ability of some autonomous
150 context dependent decision capability, i.e. needs some form of model of the environment. In the words of Judea
151 Pearl [25]: "An intelligent system attempting to build a workable model of its environment cannot rely exclusively
152 on pre-programmed causal knowledge", but must be able to interpret autonomously direct observations. This
153 is where AI differs from more traditional approach to security, which tends to be based on rules. But it is also
154 where the use of AI looks more challenging.

155 The amount of knowledge virtually present is very much greater than the amount of knowledge explicitly
156 present. The extra knowledge is the result of query-time inference, which can require a lot of computation. And
157 yet, we humans routinely perform this kind of inference quickly and in a way that seems almost effortless." One
158 problem is to access this "virtual knowledge".

159 We humans also have a remarkable ability that is central to all recognition tasks: we begin with a set of
160 observed features, a set of expectations, and a vast collection of stored descriptions; the problem is to find the
161 stored description that best matches these features and expectations. This is a computationally demanding task
162 that we humans do frequently, quickly, and with no sense of mental effort," but far more challenging for(D D D
163 D) D

164 machines. This is where the AI formidable challenge of identifying algorithms and world representation enters.

165 The attraction of Knowledge Based systems (KBS) [21] is their ability to make context dependent inferences.
166 KBS tends to be specialized. Knowledge (virtual or real) is acquired or introduced in a variety of ways. But a
167 lot rides on the way knowledge is stored and represented.

168 Those systems can reason and make inferences. Although they have been developed with different applications

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169 in mind, in principle they could be able to make autonomous determination of whether a malicious attack is
170 unfolding.

171 Computer World surveillance may in fact be an area very appropriate for the application of constructs like
172 the KBS Scone (developed at CMU by Scott ?ahlman [22]). Like other KBS, Scone provides support for
173 representing symbolic knowledge about the world: general common-sense knowledge or knowledge about some
174 specific application domain. But Scone is designed to be used as a component in a wide range of software
175 applications. Therefore, a primary emphasis has been put on Scone's expressiveness, ease of use, scalability, and
176 on the efficiency of the most commonly used operations for search and inference. A feature of Scone, which makes
177 it attractive in the context of Computer World surveillance tools is that unlike other KBS (such as Cyc [23],
178 Owl [24], for example, and most Description Logic systems), the emphasis is on the ability to do a lot of simple
179 inference very quickly, not the ability to prove deep theorems or to solve complex logic puzzles. In the system of
180 trade-offs that underlie the development of KBS's, the priority was put on "expressiveness" and "scalability". At
181 this stage, it is far too premature to exclude or recommend any approach. c) What AI Could Bring to Computer
182 World surveillance CSRF is meant only as an example. It is only one in an already large and increasing number of
183 webapplications vulnerabilities. Many popular websites are known to have exploitable cross-site scripting (XSS)
184 [2] or cross site request forgery (CSRF, [24], [1]), "ClickJacking" vulnerabilities [4].

185 Most existing defenses against CSRF are ad hoc. Since CSRF involves in general hijacking a trusted session
186 between a user and a website, a natural approach is to make such hijacking more difficult. One possibility is to
187 not rely excessively on cookies to build trust, but add additional identifiers, at the cost of making trusted sessions
188 more burdensome. A minor consideration in the security community, but the cumulative effect of making every
189 "critical" interaction cumbersome is to project the impression that the logic of the culture of security is just the
190 opposite of the logic of the technological innovations taking place in the internet and the web.

191 Since users for their security should not have to rely on website designers to anticipate all forms of attacks,
192 tools protecting directly the users are intrinsically more attractive. The user side proposals tend to be based
193 on rules tailored for each known scenario of attack. An example is to treat HTTP POST requests as more
194 dangerous, because they are the one that changes the state of the server and forbid them in some circumstances.
195 In addition to interfere with some useful functionality, this is not sufficient since it is possible (using Java script
196 code injection) to make GET requests accomplish the same thing as POST requests. Disallowing or limiting the
197 use of Java script has been suggested. This makes sense in some specific cases, but that kind of approach is an
198 exacerbated version of security standing on the way of functionalities.

199 The commercial tool Request Rodeo is a commercial tool, which precludes scenarios on the basis of rules, that
200 could lead to CSRF attacks. Its rules are so strict that it precludes proper interactions with a large number of
201 modern websites. Relaxing the rules on the other hand would reduce the degree of protection, demonstrating if
202 need be that an ideal tool would have to be more intelligent than simply applying rules.

203 Web application security generates a very new type of challenges compared with the world of worms, buffer
204 overflows, etc.. The two worlds overlap, but they call for very different kinds of security tools. In fact there is
205 no good security tool or even paradigm yet for web application. From the perspective of Computer World
206 surveillance, the world of web applications is very complicated as it seems to offer an infinite numbers of
207 opportunities for abuse.

208 Some exploitable vulnerabilities are difficult to understand or anticipate as they result from technical details
209 of protocols, implementation of application or are consequences of abusing functionalities which otherwise are
210 very useful or valuable. This is at a time where web applications are proliferating fast and playing an increasingly
211 central role in many critical operations performed in the internet. Some exploitable vulnerabilities are difficult
212 to understand or anticipate as they result from technical details of protocols, implementation of application or
213 are consequences of abusing functionalities which otherwise are very useful or valuable. Most of web application
214 vulnerabilities stems from what makes HTTP, HTML, Java-Script and the like so efficient to support web activity.
215 The controversy around the "same origin policy" illustrates the complication of web application security. All this
216 to show that web application security calls for intelligent tools.

217 Approaches to defense deliberately relying on AI may not deliver quick results. But they offer the perspective of
218 a future very different and far more attractive than what the present approach based on "tweaking the plumbing"
219 offers. In the same way that the co-evolution of pathogens and defenses from biological Year organisms has led
220 to the emergence of the immune system, and AI-based approach to cyberdefense can be seen as the natural next
221 step.

222 To the legitimate concern that AI-based tools may be very large and suck a lot of CPUs (the immune system
223 has as many cells as the nervous system: it is a huge organ), one can point to the fact that there is not much
224 alternative. One immediate predictable benefit of approaching Computer World surveillance from and AI point
225 of view will be to raise the level of the debate. Instead of being lost in the details of the implementation of old
226 ideas, it will be forward looking.

227 On the other hand, it has to be recognized also that letting security be the organizing principle of modernization
228 of the internet, will have a stifling effect on innovations, i.e. one of the main reasons of the success of the internet.



211

Figure 1: Figure 2 . 1 . 1 :



231

Figure 2: Figure 2 . 3 . 1 :

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229 But if innovation and Computer World surveillance begin to use the common perspective of AI, both at the same
230 time rely increasingly on AI, the logic of the interaction will be dramatically different. ¹ ²

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