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SYSTEM AND METHOD MACHINE LEARNING-ASSISTED DETECTION OF HARMFUL SCRIPTS

Abstract

Disclosed herein are systems and methods for detecting harmful scripts. In one aspect, an exemplary method comprises, identifying a file that contains a script; determining a plurality of parameters of the script; analyzing the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; processing the identified file based on the recognized script programming language; generating a plurality of hash codes from the processed file; and comparing the plurality of generated hash codes with hash codes of known harmful files to determine if the file contains a harmful script.

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Background/Summary

CROSS REFERENCE TO RELATED APPLICATIONS [0001] The present application is a continuation of U.S. patent application Ser. No. 17/939,071 filed Sep. 7, 2022, which claims benefit of claims priority to Russian Patent Application No. 2022107038, filed on Mar. 17, 2022, which is incorporated herein by reference.

FIELD OF TECHNOLOGY

[0002] The present disclosure relates to the field of data security, more specifically to systems and methods for machine-learning-assisted detection of harmful scripts.

BACKGROUND

[0003] The number of users and corporations and the variety of their aims and activities is constantly increasing. There is a simultaneous increase in the number of websites and webpages being created.

[0004] The creation of a webpage, the simplest tool for accessing information on the internet, requires the use of the resources of a computer system. These resources are most commonly used with the aid of script programming languages (PHP, JavaScript, Python, Ruby and others).

[0005] A script (or a scenario) refers to a program code written in various interpretable programming languages. All scripts are executed with the aid of an external interpreter program. Unlike executable files, most scripts exist in the form of text files and can be read by humans. Harmful scripts are conventionally divided into two types. The first type of harmful scripts includes scripts that are embedded in webpage codes. These types of harmful codes are interpreted by a browser and execute actions inserted by malicious parties. The second type of harmful scripts includes scripts that are designed to be launched on a user's computer. These second type of harmful scripts are executed by components of the operating system and can access an API (file system, processes, etc.).

[0006] There are many methods of detecting harmful scripts. One method is to analyze the text of webpage files in detail. The known approaches may achieve partial detection of harmful scripts, but cannot effectively resolve the problem of detecting truncated and partially damaged harmful scripts transmitted by methods other than that of embedding in a webpage.

[0007] Therefore, there is a need for a method and a system for detection of harmful scripts.

SUMMARY

[0008] Aspects of the disclosure relate to for machine-learning assisted detection of harmful scripts.

[0009] A method for detecting harmful scripts, comprising: identifying a file that contains a script; determining a plurality of parameters of the script; analyzing the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; processing the identified file based on the recognized script programming language; generating a plurality of hash codes from the processed file; and comparing the plurality of generated hash codes with hash codes of known harmful files to determined if the file contains a harmful script.

[0010] In one aspect, the machine learning model is trained based on sets of characteristics of different script programming languages.

[0011] In one aspect, the machine learning model includes a plurality of decision trees for recognizing each script programming language.

[0012] In one aspect, the machine learning model further includes a plurality of language recognition rules that indicate how to combine results of the plurality of the decision trees to confirm or refute recognition of a script programming language of the script.

[0013] In one aspect, determining a plurality of parameters of the script includes: generating a summary of the script; and determining the plurality of parameters of the generated summary of the script.

[0014] In one aspect, the plurality of parameters include static and dynamic parameters.

[0015] In one aspect, the static parameters of the script include characteristics calculated based on a set of significant bytes and a number of occurrences of comments, lines containing symbolic expressions, and/or constructions containing expressions of known script programming languages.

[0016] In one aspect, the dynamic parameters of the script include characteristics calculated based on a number of occurrences of each type of construction containing a combination of symbols characteristic of each programming language in a set of significant bytes.

[0017] In one aspect, the processing of the file comprises at least one of: removing all symbols, except visible symbols, from an ASCII table, line break and space symbols and symbols from a Unicode table; removing lines containing comments and individual tagged lines characteristic of the recognized programming language; converting the text to lowercase; and indicating a start and an end of line constant characteristic of the recognized programming language.

[0018] In one aspect, generating a plurality of hash codes includes: dividing the processed file into a plurality of combinations of set number of symbols; calculating a number of occurrences of each type of combinations of symbols; determining a type of hash code based on the number of occurrences; and generating the plurality of hash codes based on the determined hash code type.

[0019] In one aspect, a system for detecting harmful scripts, comprising: a processor configure to: identify a file that contains a script; determine a plurality of parameters of the script; analyze the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; process the identified file based on the recognized script programming language; generate a plurality of hash codes from the processed file; and compare the plurality of generated hash codes with hash codes of known harmful files to determined if the file contains a harmful script.

[0020] In one aspect, a non-transitory computer readable medium storing thereon computer executable instructions for detecting harmful scripts, including instructions for: identifying a file that contains a script; determining a plurality of parameters of the script; analyzing the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; processing the identified file based on the recognized script programming language; generating a plurality of hash codes from the processed file; and comparing the plurality of generated hash codes with hash codes of known harmful files to determined if the file contains a harmful script.

Description

BRIEF DESCRIPTION OF THE DRAWINGS

[0021] The accompanying drawings, which are incorporated into and constitute a part of this specification, illustrate one or more example aspects of the present disclosure and, together with the detailed description, serve to explain their principles and implementations.

[0022] FIG. 1 illustrates an example of a file containing the source code of a script.

[0023] FIG. 2 illustrates a block diagram of an exemplary system for detecting harmful scripts based on a set of hash codes in accordance with aspects of the present disclosure.

[0024] FIG. 3 illustrates a method for detecting harmful scripts based on a set of hash codes in accordance with aspects of the present disclosure.

[0025] FIG. 4 presents an example of a general purpose computer system on which aspects of the present disclosure can be implemented.

DETAILED DESCRIPTION

[0026] Exemplary aspects are described herein in the context of a system, method, and a computer program for detecting harmful scripts in accordance with aspects of the present disclosure. Those of ordinary skill in the art will realize that the following description is illustrative only and is not intended to be in any way limiting. Other aspects will readily suggest themselves to those skilled in the art having the benefit of the disclosure. Reference will now be made in detail to implementations of the example aspects as illustrated in the accompanying drawings. The same reference indicators will be used to the extent possible throughout the drawings and the following description to refer to the same or like items.

[0027] The following terminologies and concepts are provided in order to clearly present the aspects of the present disclosure.

[0028] A hash function refers to an algorithm that converts a line of any length (a message) into a bitmapped line of fixed length, called a hash code, a checksum or a digital fingerprint. Many security technologies use non-reversible encryption functions, also called hash functions. The main purpose of these functions is to extract from a message of arbitrary length its hash code, which is a value of fixed size. A hash code may be used as a checksum for the original message, thus enabling the integrity of the information to be checked (if a corresponding protocol is used).

[0029] A script (or a scenario) is a program (a special form of program code), usually written in a particular interpretable (not compiler-based) language and containing individual commands or instructions. For example, a script is incorporated in the text of a webpage in the form of the original code and is interpreted by a system installed on the computer of a remote user who has requested this page.

[0030] A Java script is written in a form of a text in the Java programming language, incorporated directly in a webpage and loaded together with it. It is executed by a Java interpreter that is built into almost all modern browsers.

[0031] Malicious parties use the capacity of scripts to call the resources of a computer system for harmful purposes. Existing data transmission methods make it possible to mask a script in any kind of file and to compel the user, by one method or another, to activate it. In order to detect a script in a known script programming language, or a part thereof, a system for detecting harmful script on the basis of a set of hash codes is used.

[0032] FIG. 1 illustrates an example of a file containing the source code of a script. The file containing the script 1 is written in the Bash programming language 110. The scripts of a Bash command line are sets of commands for a Linux command line collected in files and united by a common purpose. The sets of commands for a Linux command line can be entered from a keyboard. The results of the operation of the commands may either have an independent value or may act as input data for other commands and files.

[0033] FIG. 2 illustrates a block diagram of an exemplary system 200 for detecting harmful scripts based on a set of hash codes in accordance with aspects of the present disclosure. In one aspect, the system 200 comprises: a file identifier 220, a parameter calculator 230, a hash code generator 240, a detector 250, a rules database 260, and database of script hash codes 270.

[0034] In one aspect, the system 200 is implemented via a general purpose computer, as shown in FIG. 4, including a hardware processor and memory. The system 200 may comprise functional and/or hardware modules and the file identifier 220, the parameter calculator 230, the hash code

generator **240**, the detector **250**, the rules database **260**, and the database of script hash codes **270**, which in turn comprise instructions for execution on the hardware processor.

[0035] The various aspects of the file identifier **220**, the parameter calculator **230**, the hash code generator **240**, the detector **250**, the rules database **260**, and the database of script hash codes **270** are described below.

[0036] In one aspect, the file identifier **220** is designed to identify a file containing a script. The file identifier **220** identifies a file containing a script, by analyzing each file from a plurality of files **210** in which it is necessary to check for a presence of a harmful code. The plurality of files **210** may comprise various files with which the user has interacted, in one or another form. The plurality of files **210** may include website and webpage files, electronic message files, and others. The analysis is carried out using heuristic algorithms created for searching for structured file types in which the presence of scripts is unlikely (for example, executable files in PE/ELF/MACHO format, archives, individual types of document files, etc.). Said found files are excluded from the plurality of files **210** which have to be checked for the presence of harmful code. The files remaining after this exclusion are considered to be files that may contain a harmful code.

[0037] The file identifier **220** is also designed to generate a summary of the script based on the identified file, and to transmit data about the identified file and data about the generated summary of the script to the parameter calculator **230**.

[0038] In one aspect, by the file identifier **220**, the summary of the script is generated by: isolating a set of significant bytes and a set of excluded bytes from the identified file, analyzing the isolated sets, and creating the summary of the script based on the results of the analysis.

[0039] In one aspect, the set of significant bytes is isolated from the identified file by filtering the identified file using one or more filter algorithms.

[0040] In one aspect, the filter algorithms take one or more of the following actions: [0041] identifying parts of the identified file that hypothetically contains a script; [0042] removing all symbols, except visible symbols from the ASCII table; [0043] removing line break and space symbols and symbols from the Unicode table; [0044] converting the text to lowercase; [0045] removing lines containing comments in popular script programming languages; [0046] removing lines containing the expression “any text \$ any text and punctuation mark”; and removing lines containing continuous expressions (without spaces, line breaks, punctuation marks) with a length exceeding a threshold value for lengths of symbols, wherein the threshold value for the lengths of symbols is an adjustable parameter.

[0047] In one aspect, the set of excluded bytes is obtained by removing the set of significant bytes from the identified file.

[0048] In one aspect, the summary of the script comprises a data structure calculated based on parameters of the identified file, which characterize the script contained in the file.

[0049] In one aspect, the summary of the script contains at least one of the following: [0050] a set of significant bytes; [0051] a set of excluded bytes; [0052] a number of bytes in the form of comments in each pattern from the set of excluded bytes; [0053] a size of a block occupied by the set of significant bytes; [0054] a number of lines containing at least one significant byte; [0055] a number of lines containing the expression “[any text (set of one or more arbitrary symbols)]” or part thereof; [0056] a number of lines containing the expression “any text=any text” or part thereof; [0057] a number of lines containing the expression “\$=any text” or part thereof; [0058] a number of constructions containing the expression “if any text—fi any text” or part thereof; [0059] a number of constructions containing the expression “any text do any text” or part thereof; [0060] a number of constructions containing the expression “any text case—esac any text” or part thereof; [0061] a number of constructions containing the expression “if any text—fi any text” or part thereof; [0062] a number of constructions containing the expression “def any text” or part thereof; [0063] a number of constructions containing the expression “use any text” or part thereof; [0064] a number of constructions containing the expression “any number: any number” or part thereof; [0065] a number of lines beginning with a symbol from the Latin alphabet; [0066] a number of lines containing only one “.” symbol among all the symbols in the line; [0067] a number of lines in which the last symbol is the symbol “\”; [0068] a number of lines in which the last symbol is the symbol “:”; [0069] a number of lines in which the last symbol is the symbol “;”; [0070] a total number of lines; [0071] a number of significant lines; [0072] a number of bytes from the file examined; and [0073] a presence of a set of shebang symbols “#!/bin/bash”.

[0074] In one aspect, the data about the identified file include the file itself and the data obtained as a result of its analysis.

[0075] In one aspect, the parameter calculator **230** is designed to calculate static and dynamic parameters of the generated summary of the script.

[0076] In one aspect, the static parameters of the generated summary of the script include the following list of characteristics: [0077] an entropy of a block occupied by the set of significant bytes; [0078] a number of occurrences of each significant byte of the possible significant bytes in a block occupied by a set of significant bytes, divided by a total number of significant bytes and multiplied by 1024; [0079] a number of excluded comments of each type, divided by a total number of excluded comments and multiplied by 1024; [0080] a presence of a set of shebang symbols: 512 if yes, 0 otherwise; [0081] a number of lines containing the expression “[any text]” or part thereof, divided by the number of significant lines and multiplied by 1024; [0082] a number of lines containing the expression “[any text=any text]” or part thereof, divided by the number of significant lines and multiplied by 1024; [0083] a number of lines containing the expression “[\$=any text]” or part thereof, divided by the number of significant lines and multiplied by 1024; [0084] a number of constructions containing the expression “if any text-fi any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0085] a number of constructions containing the expression “any text do any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0086] a number of constructions containing the expression “any text case-esac any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0087] a number of constructions containing the expression “var any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0088] a number of constructions containing the expression “def any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0089] a number of constructions containing the expression “use any text” or part thereof, divided by the number of significant lines and multiplied by 1024; [0090] a number of constructions containing the expression “any number: any number” or part thereof, divided by the number of significant lines and multiplied by 1024; [0091] a number of lines beginning with a symbol from the Latin alphabet, divided by the number of significant lines and multiplied by 1024; [0092] a number of lines containing only one “.” symbol among all the symbols in the line, divided by the number of significant lines and multiplied by 1024; [0093] a number of lines in which the last symbol is the symbol “\”; [0094] a number of lines in which the last symbol is the symbol “\”, divided by the number of significant lines and multiplied by 1024; [0095] a number of lines in which the last symbol is the symbol “:”, divided by the number of significant lines and multiplied by 1024; [0096] a number of lines in which the last symbol is the symbol “;”, divided by the number of significant lines and multiplied by 1024; and [0097] a number of lines beginning with symbols from the Latin alphabet, divided by the number of significant lines and multiplied by 1024.

[0098] In one aspect, the dynamic parameters of the generated summary of the script include a number of occurrences of each type of construction containing a combination of symbols characteristic of each programming language in the set of significant bytes, divided by the total number of occurrences of all constructions and multiplied by 1024.

[0099] For the Bash language, for example, the number of occurrences in the set of significant bytes of constructions containing a combination of symbols may be calculated from the following dynamic list of combinations of symbols: my\$; \$\$; “\$ {; \$!; \$?; \$x; \$ #; \$*; \$@; \$-; \$.; \$ (; \$0; \$1; \$2; \$3; ==; !=; <; >; =; --; +r; +w; +x; -a; -b; -c; -e; -f; -g; -h; -i; -j; -k; -l; -m; -n; -o; -p; -q; -r; -s; -t; -u; -v; -W; -x; -y; -z; >>; <<; /bin;/boot;/dev;/etc;/opt;/X11;/sgml;/xml;/home;/lib;/media;/mnt;/opt;/proc;/root;/run;/sbin;/srv;/sys;/tmp;/usr;/bin;/include;/lib;/local;/sbin;/share;/src;/var; breaksw; endsw; *); if; fi; then; case; esac; true; false; do; done; while; for; until; break; continue; select; function; if [“\$; if (“\$; if (“\$; if [“\$; if [!-d; if [!-d; if [!-f; if [!-f; iftest; then; case” \$; echo\$; echo “\$; sed-; grep; ne-; while [\$;] do; while read; test; apt; zypp; dnf; dig; chown; sync; chsh; chattr; puts; rm-; rm*; rpm; fold; rmdir; less; more; readelf; mv; pg_dump; popd; last; @echooff; @echoon; echooff; echoon; scp; pwd; logout; find; lsof; uuencode; yum; cat; clear; xmllint; stat; ssh-copy-i; Prints; printf; host; chmod; split; xxd; top; cd; od; sed; bind; whoami; ionice; touch; screen; diff; set; dirname; ifconfig; uniq; uname; cpio; eval; brew; awk; which; shopt; pkg; pdfunite; bunzip2; watch; ifconfig; bzip2; join; exec; . . . ; seq; xhost; bash; unalias; env; comm; node; ping; file; zcat; cowsay; tac; unzip; which; netstat; pstree; readlink; getent; lspci;/bin/sh; apropos; ssh; cd\$;

bar; ant; ipc; name; nslookup; tail; .sh; tar; hostname; setarch; mate; kill; cut; alias; nohup; trap; chgrp; grep; wget; lshw; rev; locate; shred; sshpass; make; ssh-keygen; nslookup; tee; mktemp; shift; objdump; gunzip; mount; cron; stdbuf; pushd; groups; qstat; xargs; tmux; sudo; mkdir; strace; timex; mvn; tree; eof; eod; exit\$;/dev/null; expect; . . . /; pkill; 2>&1; install; gcc; python; perl; java; .so; "\$@"; http; ->; class; module; import; def; std::; var; function; usestrict; export; default; jquery; const; from; namespace; public; private; protected; interface; usewarnind; uselib; foreach; my\$; eval; ==>; package; sub; usevars; useparent; \$self; <-; body {; typedef; @shift; window; microsoft; font;. PHONY; CMAKE_; PROJECT_; CTEST_; CPACK_; gem::; gem"; gemspec; docker; ENTRYPOINT; post; host::; user-agent; Accept::; DNT::; Content-Ty; Content-Le; Disass; PUSHBUTTON; COMBOBOX; LTEXT; CONTROL; EDITTEXT; EDITTEXT; IDS_; DS_; WS_; IDD_; <tr><td.

[0100] In one aspect, the combinations of symbols characteristic of a given language are revised using machine learning and mathematical statistics. Firstly, the key words, expressions, and names or parts of names of functions, classes and variables from popular libraries of a given programming language are examined. Then, the occurrences are calculated by calculating the values of a rolling hash function for the set of significant bytes. A byte sequence of 10 bytes is optimal in terms of complexity and speed of calculation. Said hash code is calculated, starting with the first byte of the byte sequence. There is then a shift of one byte and said hash code is re-calculated from the second byte of the byte sequence. This action is then repeated for the whole set of significant bytes. The number of occurrences of the hash codes from said constructions is then calculated, with an accuracy such that they fall within the Bloom probability number.

[0101] In one aspect, the parameter calculator **230** is also designed to recognize a script programming language based on the calculated parameters of the generated summary of the script, using language recognition rules obtained from the rules database **260**, and to transfer data about the identified file, data about the generated summary of the script, and data about the recognized script programming language to the hash code generator **240**.

[0102] In one aspect, the script programming language is recognized with the aid of machine learning methods. The calculated parameters are used as characteristics that are sent to the input of pre-trained machine learning models in the form of decision trees.

[0103] In one aspect, the machine learning model uses five decision tree patterns for confirming the programming language. For each programming language, a set of characteristics is generated and used for training the decision trees. As a result of the training, five decision trees are generated for one language. The result of the operation of each tree is either a confirmation or a non-confirmation of whether the examined script is written in the current programming language. A final conclusion for confirming the current programming language may be drawn from the results of the operation of all five decision trees.

[0104] In one aspect, in addition to the sets of trees for confirming the script programming language, decision trees may also be created for confirming alternative languages, such as a configuration file language (ini-file) or a PostScript programming language (pdf-file). These additional decision trees may be used to reduce false positives.

[0105] A language recognition rule includes a set of results of the operation of all the programming language confirmation decision trees for the recognition of all languages. The presence of a result confirming to the language recognition rule is used to detect a script that is written in a specific programming language.

[0106] An example of a language recognition rule is the following set of results of the operation of decision trees: [0107] a minimum of three out of five decision trees for confirming one programming language have yielded a positive result; and [0108] the decision trees for confirming other languages have yielded not more than two positive results for any language.

[0109] When said set of conditions are satisfied, it is considered that the detected script is written in the programming language for the recognition of which three out of five decision trees have yielded a positive result.

[0110] Another example of a language recognition rule is the following set of results of the operation of decision trees: [0111] a minimum of three out of five decision trees for confirming one programming language have yielded a positive result; [0112] the decision trees for confirming other languages have yielded not more than two positive results for any language; and [0113] a minimum of four out of five decision trees for confirming an alternative language have given a positive result.

[0114] When said set of conditions are satisfied, it is considered that the detected script is written in the alternative language, for the recognition of which four out of five decision trees have yielded a positive result.

[0115] Another example of a recognition rule is the following set of results of the operation of decision trees: [0116] a minimum of three out of five decision trees for confirming one language have yielded a positive result; and [0117] a minimum of three out of five decision trees for recognizing another language have yielded a positive result.

[0118] When said set of conditions are satisfied, it is considered that a further decision tree must be used to refine the result of the decision trees that have yielded a positive result.

[0119] In one aspect, the hash code generator **240** is designed to process the identified file based on the data about the recognized programming language, thus converting the identified file into a specific form.

[0120] In one aspect, the hash code generator **240** processes the identified file by: [0121] removing all the symbols, except visible symbols from the ASCII table; line break and space symbols and symbols from the Unicode table; lines containing comments and individual tagged lines characteristic of the recognized programming language; [0122] converting the text to lowercase; and indicating the start and end of line constants characteristic of the recognized programming language.

[0123] In one aspect, the hash code generator **240** is also designed to: generate a set of hash codes based on the processed file, use the rules for generating hash codes from the rules database **260**; and transmit the generated set of hash codes to the detector **250**.

[0124] In one aspect, the hash code generator **240** generates the set of hash codes by executing the following list of actions: [0125] 1) dividing the processed file into constructions containing a combination of symbols with a length of 4 to 10 symbols; [0126] 2) calculating the number of occurrences of all the detected constructions containing a combination of symbols with a length of 4 to 10 symbols in the processed file; for example, if the processed file is represented in the form of the symbols "colacola", then the division into constructions results in the following block: {cola: 2, colacola: 1, olacola: 1, lacula: 1, acola: 1, colac: 1, colaco: 1, colacol: 1}; [0127] 3) determining the type of hash code used, with the aid of the hash code generation rules; and [0128] 4) generating a set of hash codes from the hash codes of a specified type.

[0129] In one aspect, the number of occurrences of each type of construction, from all the programming languages, containing a combination of symbols with a length of 4 to 10 symbols, in the processed file is calculated by calculating the rolling hash function formed from the parts of the file, in the form of a byte sequence with a size of 10 bytes, where the number of symbols in the combination of symbols has a length of 4 to 10 symbol. Then, the number of occurrences of the hash codes from said constructions is calculated with an accuracy such that they fall within the Bloom probability number (the size of which coincides with the size of the Bloom probability number used in the language recognition step).

[0130] In one aspect, the rule for generating hash codes includes a set of conditions. For example, based on whether or not the rules for generating hash codes are satisfied, a first given hash code or a second given hash code may be used.

[0131] In one aspect, the creation of hash code generation rules is mainly based on the calculated number of occurrences. An example of a rule for generating hash codes is the case where the number of occurrences is less than a threshold value. If this set of conditions is satisfied, when the set of hash codes is generated, an abbreviated type of hash code (described below) is used. Another example of a rule for generating hash codes is the case where the number of occurrences is equal to or greater than a threshold value. If this set of conditions is satisfied, a full-size type of hash code (described below) is used when the set of hash codes is generated. The threshold value of the number of occurrences may be determined statistically or empirically, or may be specified by an expert. In one aspect, the threshold value of the number of occurrences is established as a result of the operation of machine learning algorithms.

[0132] In one aspect, the abbreviated type of hash code includes a 64-bit value obtained from a combination of two halves of the 128-bit

rules for generating hash codes based on the rules database **260**. In one aspect, by the parameter calculator **230**, method **300** transmits data about the identified file, data about the generated summary of the script, and data about the recognized script programming language to the hash code generator **240**.

[0151] In one aspect, the at least one language recognition rule includes a set of results of operations of at least one decision tree, where a presence of a result of the at least one decision tree is used to indicate that a script written in a specific programming language has been detected.

[0152] In one aspect, the static parameters of the generated summary of the script include at least one of: a list of characteristics calculated based on a set of significant bytes and a number of occurrences of comments, lines containing symbolic expressions, and constructions containing expressions of known script programming languages.

[0153] In one aspect, the dynamic parameters of the generated summary of the script include at least a list of characteristics calculated based on a number of occurrences of each type of construction containing a combination of symbols characteristic of each programming language in a set of significant bytes.

[0154] In step **315**, by the hash code generator **240**, method **300** processes the identified file based on the data about the recognized script programming language.

[0155] In one aspect, the processing of the identified file comprises: removing all symbols, except visible symbols, from the ASCII table, line break and space symbols and symbols from the Unicode table; removing lines containing comments and individual tagged lines characteristic of the recognized programming language; converting the text to lowercase; and indicating a start and an end of line constant characteristic of the recognized programming language.

[0156] In step **316**, by the hash code generator **240**, method **300** generates a set of hash codes based on a processed file using rules for generating hash codes. In one aspect, the rules for generating hash codes are obtained from the rules database **260**. In one aspect, by the hash code generator **240**, method **300** transmits the generated set of hash codes to the detector **250**.

[0157] In one aspect, the set of hash codes is generated by: dividing the processed file into constructions containing a combination of symbols with a length of 4 to 10 symbols; calculating a number of occurrences of all detected constructions in the processed file; determining a type of hash code used using hash code generation rules; and generating the set of hash codes from hash codes of a specified type.

[0158] In one aspect, the hash code generation rules include a set of conditions, wherein, when the set of conditions is satisfied, a specified type of hash code is used.

[0159] In step **317**, by the detector **250**, method **300** determines whether or not the generated set of hash codes is similar to known harmful sets of hash codes. If a similarity to known harmful sets of hash codes is not present, method **300** proceeds to step **318**. If a similarity to known harmful sets of hash codes is present, method **300** proceeds to step **319**. In step **318**, the system **200** terminated method **300**. In step **319**, by the detector **250**, method **300** detects the harmful script.

[0160] In one aspect, the known harmful sets of hash codes are obtained from a database of script hash codes **270**.

[0161] FIG. **4** is a block diagram illustrating a computer system **20** on which aspects of systems and methods for detecting harmful scripts based on a set of hash codes may be implemented. The computer system **20** can be in the form of multiple computing devices, or in the form of a single computing device, for example, a desktop computer, a notebook computer, a laptop computer, a mobile computing device, a smart phone, a tablet computer, a server, a mainframe, an embedded device, and other forms of computing devices.

[0162] As shown, the computer system **20** includes a central processing unit (CPU) **21**, a system memory **22**, and a system bus **23** connecting the various system components, including the memory associated with the central processing unit **21**. The system bus **23** may comprise a bus memory or bus memory controller, a peripheral bus, and a local bus that is able to interact with any other bus architecture. Examples of the buses may include PCI, ISA, PCI-Express, HyperTransport™, InfiniBand™, Serial ATA, I2C, and other suitable interconnects. The central processing unit **21** (also referred to as a processor) can include a single or multiple sets of processors having single or multiple cores. The processor **21** may execute one or more computer-executable code implementing the techniques of the present disclosure. The system memory **22** may be any memory for storing data used herein and/or computer programs that are executable by the processor **21**. The system memory **22** may include volatile memory such as a random access memory (RAM) **25** and non-volatile memory such as a read only memory (ROM) **24**, flash memory, etc., or any combination thereof. The basic input/output system (BIOS) **26** may store the basic procedures for transfer of information between elements of the computer system **20**, such as those at the time of loading the operating system with the use of the ROM **24**.

[0163] The computer system **20** may include one or more storage devices such as one or more removable storage devices **27**, one or more non-removable storage devices **28**, or a combination thereof. The one or more removable storage devices **27** and non-removable storage devices **28** are connected to the system bus **23** via a storage interface **32**. In an aspect, the storage devices and the corresponding computer-readable storage media are power-independent modules for the storage of computer instructions, data structures, program modules, and other data of the computer system **20**. The system memory **22**, removable storage devices **27**, and non-removable storage devices **28** may use a variety of computer-readable storage media. Examples of computer-readable storage media include machine memory such as cache, SRAM, DRAM, zero capacitor RAM, twin transistor RAM, eDRAM, EDO RAM, DDR RAM, EEPROM, NRAM, RRAM, SONOS, PRAM; flash memory or other memory technology such as in solid state drives (SSDs) or flash drives; magnetic cassettes, magnetic tape, and magnetic disk storage such as in hard disk drives or floppy disks; optical storage such as in compact disks (CD-ROM) or digital versatile disks (DVDs); and any other medium which may be used to store the desired data and which can be accessed by the computer system **20**.

[0164] The system memory **22**, removable storage devices **27**, and non-removable storage devices **28** of the computer system **20** may be used to store an operating system **35**, additional program applications **37**, other program modules **38**, and program data **39**. The computer system **20** may include a peripheral interface **46** for communicating data from input devices **40**, such as a keyboard, mouse, stylus, game controller, voice input device, touch input device, or other peripheral devices, such as a printer or scanner via one or more I/O ports, such as a serial port, a parallel port, a universal serial bus (USB), or other peripheral interface. A display device **47** such as one or more monitors, projectors, or integrated display, may also be connected to the system bus **23** across an output interface **48**, such as a video adapter. In addition to the display devices **47**, the computer system **20** may be equipped with other peripheral output devices (not shown), such as loudspeakers and other audiovisual devices.

[0165] The computer system **20** may operate in a network environment, using a network connection to one or more remote computers **49**. The remote computer (or computers) **49** may be local computer workstations or servers comprising most or all of the aforementioned elements in describing the nature of a computer system **20**. Other devices may also be present in the computer network, such as, but not limited to, routers, network stations, peer devices or other network nodes. The computer system **20** may include one or more network interfaces **51** or network adapters for communicating with the remote computers **49** via one or more networks such as a local-area computer network (LAN) **50**, a wide-area computer network (WAN), an intranet, and the Internet. Examples of the network interface **51** may include an Ethernet interface, a Frame Relay interface, SONET interface, and wireless interfaces.

[0166] Aspects of the present disclosure may be a system, a method, and/or a computer program product. The computer program product may include a computer readable storage medium (or media) having computer readable program instructions thereon for causing a processor to carry out aspects of the present disclosure.

[0167] The computer readable storage medium can be a tangible device that can retain and store program code in the form of instructions or data structures that can be accessed by a processor of a computing device, such as the computing system **20**. The computer readable storage medium may be an electronic storage device, a magnetic storage device, an optical storage device, an electromagnetic storage device, a semiconductor storage device, or any suitable combination thereof. By way of example, such computer-readable storage medium can comprise a random access memory (RAM), a read-only memory (ROM), EEPROM, a portable compact disc read-only memory (CD-ROM), a digital versatile disk (DVD), flash

memory, a hard disk, a portable computer, a memory stick, a floppy disk, or even a punch-cards or raised structures in a groove having instructions recorded thereon. As used herein, a computer readable storage medium is not to be construed as being transitory signals per se, such as radio waves or other freely propagating electromagnetic waves, electromagnetic waves propagating through a waveguide or transmission media, or electrical signals transmitted through a wire.

[0168] Computer readable program instructions described herein can be downloaded to respective computing devices from a computer readable storage medium or to an external computer or external storage device via a network, for example, the Internet, a local area network, a wide area network and/or a wireless network. The network may comprise copper transmission cables, optical transmission fibers, wireless transmission, routers, firewalls, switches, gateway computers and/or edge servers. A network interface in each computing device receives computer readable program instructions from the network and forwards the computer readable program instructions for storage in a computer readable storage medium within the respective computing device.

[0169] Computer readable program instructions for carrying out operations of the present disclosure may be assembly instructions, instruction-set-architecture (ISA) instructions, machine instructions, machine dependent instructions, microcode, firmware instructions, state-setting data, or either source code or object code written in any combination of one or more programming languages, including an object oriented programming language, and conventional procedural programming languages. The computer readable program instructions may execute entirely on the user's computer, partly on the user's computer, as a stand-alone software package, partly on the user's computer and partly on a remote computer or entirely on the remote computer or server. In the latter scenario, the remote computer may be connected to the user's computer through any type of network, including a LAN or WAN, or the connection may be made to an external computer (for example, through the Internet). In some aspects, electronic circuitry including, for example, programmable logic circuitry, field-programmable gate arrays (FPGA), or programmable logic arrays (PLA) may execute the computer readable program instructions by utilizing state information of the computer readable program instructions to personalize the electronic circuitry, in order to perform aspects of the present disclosure.

[0170] In various aspects, the systems and methods described in the present disclosure can be addressed in terms of modules. The term "module" as used herein refers to a real-world device, component, or arrangement of components implemented using hardware, such as by an application specific integrated circuit (ASIC) or FPGA, for example, or as a combination of hardware and software, such as by a microprocessor system and a set of instructions to implement the module's functionality, which (while being executed) transform the microprocessor system into a special-purpose device. A module may also be implemented as a combination of the two, with certain functions facilitated by hardware alone, and other functions facilitated by a combination of hardware and software. In certain implementations, at least a portion, and in some cases, all, of a module may be executed on the processor of a computer system (such as the one described in greater detail in FIG. 4, above). Accordingly, each module may be realized in a variety of suitable configurations, and should not be limited to any particular implementation exemplified herein.

[0171] In the interest of clarity, not all of the routine features of the aspects are disclosed herein. It would be appreciated that in the development of any actual implementation of the present disclosure, numerous implementation-specific decisions must be made in order to achieve the developer's specific goals, and these specific goals will vary for different implementations and different developers. It is understood that such a development effort might be complex and time-consuming, but would nevertheless be a routine undertaking of engineering for those of ordinary skill in the art, having the benefit of this disclosure.

[0172] Furthermore, it is to be understood that the phraseology or terminology used herein is for the purpose of description and not of restriction, such that the terminology or phraseology of the present specification is to be interpreted by the skilled in the art in light of the teachings and guidance presented herein, in combination with the knowledge of those skilled in the relevant art(s). Moreover, it is not intended for any term in the specification or claims to be ascribed an uncommon or special meaning unless explicitly set forth as such.

[0173] The various aspects disclosed herein encompass present and future known equivalents to the known modules referred to herein by way of illustration. Moreover, while aspects and applications have been shown and described, it would be apparent to those skilled in the art having the benefit of this disclosure that many more modifications than mentioned above are possible without departing from the inventive concepts disclosed herein.

Claims

1. A method for detecting harmful scripts, comprising: identifying a file that contains a script; determining a plurality of parameters of the script; analyzing the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; processing the identified file based on the recognized script programming language; generating a plurality of hash codes from the processed file; and comparing the plurality of generated hash codes with hash codes of known harmful files to determine if the file contains a harmful script.
2. The method of claim 1, wherein the machine learning model is trained based on sets of characteristics of different script programming languages.
3. The method of claim 1, wherein the machine learning model includes a plurality of decision trees for recognizing each script programming language.
4. The method of claim 3, wherein the machine learning model further includes a plurality of language recognition rules that indicate how to combine results of the plurality of the decision trees to confirm or refute recognition of a script programming language of the script.
5. The method of claim 1, wherein determining a plurality of parameters of the script includes: generating a summary of the script; and determining the plurality of parameters of the generated summary of the script.
6. The method of claim 1, wherein the plurality of parameters include static and dynamic parameters.
7. The method of claim 6, wherein the static parameters of the script include characteristics calculated based on a set of significant bytes and a number of occurrences of comments, lines containing symbolic expressions, and/or constructions containing expressions of known script programming languages.
8. The method of claim 6, wherein the dynamic parameters of the script include characteristics calculated based on a number of occurrences of each type of construction containing a combination of symbols characteristic of each programming language in a set of significant bytes.
9. The method of claim 1, wherein the processing of the file comprises at least one of: removing all symbols, except visible symbols, from an ASCII table, line break and space symbols and symbols from a Unicode table; removing lines containing comments and individual tagged lines characteristic of the recognized programming language; converting the text to lowercase; and indicating a start and an end of line constant characteristic of the recognized programming language.
10. The method of claim 1, wherein generating a plurality of hash codes includes: dividing the processed file into a plurality of combinations of set number of symbols; calculating a number of occurrences of each type of combinations of symbols; determining a type of hash code based on the number of occurrences; and generating the plurality of hash codes based on the determined hash code type.
11. A system for detecting harmful scripts, comprising: a processor configured to: identify a file that contains a script; determine a plurality of parameters of the script; analyze the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; process the identified file based on the recognized script programming language; generate a plurality of hash codes from the processed file; and compare the plurality of generated hash codes with hash codes of known harmful files to determine if the file contains a harmful script.
12. The system of claim 11, wherein the machine learning model is trained based on sets of characteristics of different script programming languages.
13. The system of claim 11, wherein the machine learning model includes a plurality of decision trees for recognizing each script programming

language.

14. The system of claim 13, wherein the machine learning model further includes a plurality of language recognition rules that indicate how to combine results of the plurality of the decision trees to confirm or refute recognition of a script programming language of the script.

15. The system of claim 11, wherein determining a plurality of parameters of the script includes: generating a summary of the script; and determining the plurality of parameters of the generated summary of the script.

16. The system of claim 11, wherein the plurality of parameters include static and dynamic parameters.

17. The system of claim 16, wherein the static parameters of the script include characteristics calculated based on a set of significant bytes and a number of occurrences of comments, lines containing symbolic expressions, and/or constructions containing expressions of known script programming languages.

18. The system of claim 16, wherein the dynamic parameters of the script include characteristics calculated based on a number of occurrences of each type of construction containing a combination of symbols characteristic of each programming language in a set of significant bytes.

19. A non-transitory computer readable medium storing thereon computer executable instructions for detecting harmful scripts, including instructions for: identifying a file that contains a script; determining a plurality of parameters of the script; analyzing the plurality of parameters of the script using a machine learning model trained to recognize a script programming language of said script; processing the identified file based on the recognized script programming language; generating a plurality of hash codes from the processed file; and comparing the plurality of generated hash codes with hash codes of known harmful files to determine if the file contains a harmful script.

20. The non-transitory computer readable medium of claim 19, wherein the machine learning model includes a plurality of decision trees for recognizing each script programming language, and a plurality of language recognition rules that indicate how to combine results of the plurality of the decision trees to confirm or refute recognition of a script programming language of the script.
