

URL ABNORMALITY POSITIONING METHOD AND DEVICE, AND SERVER AND STORAGE MEDIUM

CROSS-REFERENCE TO RELATED APPLICATIONS

[0001] The present application claims the priority to Chinese Patent Application No. 201810182571.X, entitled “URL ABNORMALITY POSITIONING METHOD AND DEVICE, AND SERVER AND STORAGE MEDIUM” filed on March 6, 2018, which is incorporated by reference herein in its entirety.

TECHNICAL FIELD

[0002] Embodiments of the present specification relate to the field of Internet technologies, and in particular, to a URL abnormality locating method and device, and a server and storage medium.

BACKGROUND

[0003] In an application scenario of the Internet, there are a large number of accesses to a URL (Uniform Resource Locator) every day. In this case, some lawbreakers attempt to perform an attack through an illegal URL access.

SUMMARY

[0004] Embodiments of the present specification provide a URL abnormality locating method and device, and a server and storage medium.

[0005] In a first aspect, an embodiment of the present specification provides a URL abnormality locating method, including: performing field segmentation on a URL to obtain a multiple-instance bag including all instances corresponding to all fields; inputting the multiple-instance bag into a multiple-instance learning-based URL abnormality locating model to predict an abnormal instance; and locating a corresponding abnormal field according to the abnormal instance.

[0006] In a second aspect, an embodiment of the present specification provides a URL abnormality locating training method, including: collecting a URL sample set including a plurality of URL samples; performing field segmentation on each URL sample in the URL sample set, and for each URL sample, obtaining a multiple-instance bag including all

instances corresponding to all fields; collecting multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; performing abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm; and obtaining the URL abnormality locating model based on the classification training.

[0007] In a third aspect, an embodiment of the present specification provides a URL abnormality locating device, including: a segmenting unit configured to perform field segmentation on a URL to obtain a multiple-instance bag including all instances corresponding to all fields; a predicting unit configured to input the multiple-instance bag into a multiple-instance learning-based URL abnormality locating model to predict an abnormal instance; and a locating unit configured to locate a corresponding abnormal field according to the abnormal instance.

[0008] In a fourth aspect, an embodiment of the present specification provides a URL abnormality locating training device, including: a sample acquiring unit configured to collect a URL sample set including a plurality of URL samples; a sample segmenting unit configured to perform field segmentation on each URL sample in the URL sample set, and for each URL sample, obtain a multiple-instance bag including all instances corresponding to all fields; an instance bag collecting unit configured to collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; and a training unit configured to perform abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm, so as to obtain the URL abnormality locating model.

[0009] In a fifth aspect, an embodiment of the present specification provides a server including a memory, a processor, and a computer program stored on the memory and executable on the processor, when the processor executes the program, steps of the method according to any of the foregoing being implemented.

[0010] In a sixth aspect, an embodiment of the present specification provides a computer-readable storage medium storing a computer program, when the program is executed by a processor, steps of the method according to any of the foregoing being implemented.

[0011] The embodiments of the present specification have beneficial effects below:

[0012] In an embodiment of the present specification, when the URL is represented as a bag including a plurality of instances corresponding to a plurality of fields, an abnormal instance is predicted by using a URL abnormality locating model, so as to locate an abnormal

field in the URL. By multiple-instance learning-based URL abnormality locating, an undiscovered potential threat in data accessed every day can be well predicted. Because an abnormality may be determined for an abnormal URL, a great support may be provided for discovering the potential threat, building a new security rule, and establishing a security system.

BRIEF DESCRIPTION OF THE DRAWINGS

[0013] FIG. 1 is a schematic diagram of a URL abnormality locating scenario, according to an embodiment of the present specification.

[0014] FIG. 2 is a flowchart of a URL abnormality locating method, according to a first aspect of an embodiment of the present specification.

[0015] FIG. 3 is a flowchart of a URL abnormality locating training method, according to a second aspect of an embodiment of the present specification.

[0016] FIG. 4 is a schematic structural diagram of a URL abnormality locating device, according to a third aspect of an embodiment of the present specification.

[0017] FIG. 5 is a schematic structural diagram of a URL abnormality locating training device, according to a fourth aspect of an embodiment of the present specification.

[0018] FIG. 6 is a schematic structural diagram of a URL abnormality locating server, according to a fifth aspect of an embodiment of the present specification.

DETAILED DESCRIPTION OF THE INVENTION

[0019] To better understand the above technical solutions, the technical solutions of the embodiments of the present specification will be described in detail below with reference to the accompanying drawings and specific embodiments. It should be understood that the embodiments of the present specification and specific features in the embodiments are detailed descriptions of the technical solutions of the embodiments of the present specification and are not intended to limit the technical solutions of the present specification. The embodiments of the present specification and the technical features in the embodiments may be combined with each other without conflict.

[0020] FIG. 1 is a schematic diagram of a URL abnormality locating scenario according to an embodiment of the present specification. A client (user side) makes a URL access request to a server (network side). The server parses the URL access request to obtain a URL, and performs multiple-instance learning-based URL abnormality locating. In

particular, the server trains a URL abnormality locating model based on a multiple-instance learning algorithm in advance, predicts an abnormal instance through the URL abnormality locating model, and then determines a field at which a corresponding URL is located, so as to locate URL abnormality. In the embodiments of the present specification, the so-called “URL abnormality locating” is different from determination whether the entire URL is abnormal or not. Instead, the position of a specific abnormal field in the URL is determined, to facilitate more accurate analysis and precaution of an abnormality.

[0021] In a first aspect, an embodiment of the present specification provides a URL abnormality locating method.

[0022] Referring to FIG. 2, a URL abnormality locating method according to an embodiment of the present specification includes step S201 to step S203 below.

[0023] S201. Perform field segmentation on a URL to obtain a multiple-instance bag including all instances corresponding to all fields.

[0024] Based on multiple-instance learning (MIL), the URL is segmented to obtain a bag including a plurality of instances. Different from a traditional supervised learning, in the multiple-instance learning, data is given in a form of bag. One bag usually has a plurality of instances. It may be understood that each URL in the embodiments of the present specification corresponds to a bag. The bag includes a plurality of instances. Therefore, a “multiple-instance bag” represents one URL.

[0025] A basic URL includes a model (or protocol), a server name (or IP address), a path, and a file name, such as “protocol: // authorization / path? Query.” A complete syntax of a common uniform resource identifier with an authorization part may be shown below: Protocol: // username: password@subdomain.domain name.top-level domain: port number / directory / file name.file suffix? parameter = value # flag. In this embodiment of the present specification, field segmentation may be performed on the entire URL structure, or only a high-risk field may be segmented.

[0026] For example, only a server name field is further segmented into a plurality of instances; or a part after the # (pound sign) is segmented to obtain a plurality of instances.

[0027] The instances may be represented by feature vectors of corresponding fields. For example, a pattern, a number of characters, and a number of letters of a field are represented as the feature vectors, to obtain instances corresponding to the field.

[0028] S202. Input the multiple-instance bag into a preset URL abnormality locating model to predict an abnormal instance.

[0029] In this embodiment of the present specification, first, training is performed in

advance based on a plurality of URL samples according to the multiple-instance learning algorithm to obtain the URL abnormality locating model (referring to FIG. 3 and related description for a training process). Next, a multiple-instance bag corresponding to a to-be-predicted URL is input into the URL abnormality locating model. According to the URL abnormality locating model, a value of an abnormality label of each instance in the multiple-instance bag are predicted, so that it is predicted that whether each instance in the multiple-instance bag is an abnormal instance.

[0030] In a traditional multiple-instance learning algorithm, a label is only given in a bag level. For example, for a standard multiple-instance learning, take a binary classification as an instance, provided that one of the instances of one bag is a positive instance, the bag is a positive bag. All instances in a negative bag are negative instances. It should be noted that a label of a bag is known, and a label of an instance is unknown. Therefore, compared with a traditional supervised learning, the multiple-instance learning has less supervised information and is more difficult.

[0031] In this embodiment of the present specification, by giving labels of the instances, an abnormal instance and a normal instance (not limited to a bag-level label) are distinguished.

[0032] There is at least one positive instance in a positive bag. All instances in a negative bag are negative instances. It should be noted that, in the embodiments of the present specification, the so-called positive instance refers to an abnormal instance (for example, a value of an abnormality label being 1 or having a positive label), and the positive bag is also an abnormal bag. The so-called negative instance refers to a normal instance (for example, a value of the abnormality label being 0 or having a negative label), and the negative bag is also a normal bag.

[0033] S203. Locate a corresponding abnormal field according to the abnormal instance.

[0034] Because each instance is determined to correspond to a field, after the abnormal instance is predicted, an abnormal field corresponding to the abnormal instance may be determined, that is, an abnormal position in the URL may be determined.

[0035] In an embodiment of the present specification, provided that one instance in one bag is positive, the bag is positive. All instances in a negative bag are negative. Therefore, provided that a multiple-instance bag corresponding to a URL includes one abnormal instance, the URL is determined to be an abnormal URL. If the multiple-instance bag corresponding to the URL includes no abnormal instance, the URL is determined to be a

normal URL. Therefore, if it is determined that the URL has a corresponding abnormal instance, the URL may be determined to be an abnormal URL.

[0036] FIG. 3 is a flowchart of a URL abnormality locating training method according to a second aspect of an embodiment of the present specification, including the following steps.

[0037] S301. Collect a URL sample set including a plurality of URL samples.

[0038] S302. Perform field segmentation on each URL sample in the URL sample set, and for each URL sample, obtain a multiple-instance bag including all instances corresponding to all fields.

[0039] A piece of URL original sample data is segmented. Different features may be extracted from each segmented field to represent the field with an instance. Finally, different instances of one URL are collected to be used as a bag.

[0040] S303. Collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set.

[0041] S304. Based on the multiple-instance learning algorithm, perform an abnormal instance and normal instance classification training on the multiple-instance bag set.

[0042] The instance classification training may be understood as a process where the label for each instance is attempted to be dug from each bag so as to train a classifier to classify the instances. An abnormal instance and a normal instance are distinguished through values of abnormality labels of the instances.

[0043] In an optional manner, a specific process of performing the abnormal instance and normal instance classification training on the multiple-instance bag set includes: initializing a value of an abnormality label of each instance in the multiple-instance bag set, and performing iterative learning on the value of the abnormality label to update a final value of the abnormality label of each instance.

[0044] Because an instance label is unobservable, a possible label is initialized first, and then the instance label is iteratively updated during the training process so as to optimize a result. As training progresses, the instance label is increasingly closer to a fact. Therefore, a model may gradually find a possible abnormal instance (harmful instance), for a final model to predict a possible abnormal field of a future URL.

[0045] In particular, as previously described, there is at least one positive instance in a positive bag. All instances in a negative bag are negative instances. It should be noted that, in this embodiments of the present specification, the so-called positive instance refers to an abnormal instance (for example, a value of an abnormality label being 1 or having a positive

label), and the positive bag is also an abnormal bag. The so-called negative instance refers to a normal instance (for example, a value of the abnormality label being 0 or having a negative label), and the negative bag is also a normal bag.

[0046] Therefore, in initialization, one or more instances in a positive bag are randomly selected to be provided with positive labels. The remaining instances in the positive bag are provided with negative labels. All instances in a negative bag are provided with negative labels.

[0047] Based on the initialization, each instance gets one instance label. One classifier is trained according to the instances. When the classifier is obtained, all instances may be predicted to update the labels of the instances. For example, a specific update process is below: instance labels (all negative labels) of a negative bag keep unchanged, and for an instance in a positive bag, an instance label is corrected according to a result of the classifier. Afterwards, a round of inspections is performed. That is to say, if all instances of a positive bag are provided with negative labels, an instance that is of the instances with negative labels and that has a lowest score (that is, that is determined to have a lowest confidence degree in the negative instances) is selected to be provided with a positive label. In this way, update of the instance label is completed. An updated result is given to train a new classifier.

[0048] The above process is repeated, until a label of an instance between a previous round and a next round does not change.

[0049] It should be noted that the instance label is randomly provided at the beginning. During a classification training process, the instance label is gradually corrected. The corrected instance label causes the trained classifier to be more accurate. Therefore, optimization is achieved.

[0050] S305. Based on classification training, obtain an URL abnormality locating model.

[0051] An embodiment of the present specification is illustrated through a process of abnormality locating of a specific URL below.

[0052] For a normal URL, for example:

`http://render.alipay.com/p/s/alipay_site/wait?mintime=3&maxtime=5&fromspanner=goldetfprod_502.`

[0053] An attacker often performs an attack by correcting any possible part of the normal URL. In an actual service, because a domain name field (such as the foregoing: `http://render.alipay.com`) is basically fixed, only a parameter request part may be corrected, an attack often results from the parameter request part. In an example below, the attacker

executes a script by manually correcting a normal URL.

[0054] For example, an abnormal URL instance is:

`http://render.alipay.com/p/s/alipay_site/wait?mintime=3>SCRiPT={Sleep}&maxtime=5&fromspanner=goldetfprod_502`

[0055] An abnormal field herein is “mintime=3>SCRiPT={Sleep}.”

[0056] In traditional technical means, the URL is attempted to be uniformly represented. That is to say, each URL is represented in a form of a feature vector with a certain dimensionality. In this manner, an abnormal URL may be found, but an abnormal field in the foregoing instance cannot be accurately located. It is anticipated that the abnormal field may be accurately founded. Therefore, this method is intended to detect the abnormal field.

[0057] In particular, this problem is formalized as a problem of multiple-instance learning. During data acquisition, it may be often known which URLs are normal, and which are abnormal. This means that one URL label is in an entire URL level, and a problem of each field is often unobservable. This just conforms to an assumption of the multiple-instance learning: In the multiple-instance learning, data is given in a form of a bag. One bag has a label and has a plurality of instances. An instance label is unknown. If one instance in one bag is a positive instance, the bag is a positive bag. If all instances are negative instances, the bag is a negative bag.

[0058] Still taking the foregoing normal URL as an example: through a segmentation of the URL, different sub-fields are obtained, that is, three instances of “mintime=3;” “Maxtime=5;” and “Fromspanner=goldetfprod_502” (values of the three instances are “3,” “5,” and “goldetfprod_502,” respectively). One URL herein is one bag, and is represented to include three instances. Further, information of each field may be extracted and represented in various patterns, for example, a value may be extracted into different patterns:: a total number of characters, a total number of letters, a total number of digits, a total number of symbols, a number of different characters, a number of different letters, a number of different digits, and a number of different symbols. Therefore, three vectors in the bag in the foregoing instance may be expressed as [1,0,1,0,1,0,1,0], [1,0,1,0,1,0,1,0], and [15,11,3,1,13,9,3,1]. Because the URL is a normal URL, a label of the bag is negative (that is, a normal URL).

[0059] Similarly, for the foregoing corrected abnormal URL, an abnormal instance may be identified through a model. A corresponding abnormal field (abnormal sub-field), that is, “SCRiPT={Sleep},” may be determined through the abnormal instance. Therefore, abnormal analysis and abnormal precaution are facilitated.

[0060] In an embodiment of the present specification, when the URL is represented as a bag including a plurality of instances corresponding to a plurality of fields, an abnormal instance is predicted by using a URL abnormality locating model, so as to locate an abnormal field in the URL. By multiple-instance learning-based URL abnormality locating, an undiscovered potential threat in data accessed every day can be well predicted. Because an abnormality may be determined for an abnormal URL, a great support may be provided for discovering the potential threat, building a new security rule, and establishing a security system.

[0061] In a third aspect, based on the same inventive concept, an embodiment of the present specification provides a URL abnormality locating device. Referring to FIG. 4, the URL abnormality locating device includes:

[0062] a segmenting unit 401 configured to perform field segmentation on a URL to obtain a multiple-instance bag including all instances corresponding to all fields;

[0063] a predicting unit 402 configured to input the multiple-instance bag into a preset URL abnormality locating model to predict an abnormal instance; and

[0064] a locating unit 403 configured to locate a corresponding abnormal field according to the abnormal instance.

[0065] In an optional manner, the URL abnormality locating device further includes:

[0066] a model training unit 404 configured to train a plurality of URL samples based on a multiple-instance learning algorithm, to obtain the URL abnormality locating model.

[0067] In an optional manner, the model training unit 404 includes:

[0068] a sample segmenting subunit 4041 configured to perform field segmentation on each URL sample in a URL sample set, and for each URL sample, obtain a multiple-instance bag including all instances corresponding to all fields;

[0069] an instance bag collecting subunit 4042 configured to collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; and

[0070] a training subunit 4043 configured to perform an abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm, so as to obtain the URL abnormality locating model.

[0071] In an optional manner, an abnormal instance and a normal instance are distinguished through values of abnormality labels of instances.

[0072] The training subunit 4043 is specifically configured to: initialize a value of an abnormality label of each instance in the multiple-instance bag set, and perform iterative learning on the value of the abnormality label to update a final value of the abnormality label

of each instance.

[0073] In an optional manner, the predicting unit 402 is specifically configured to: predict the value of the abnormality label of each instance in the multiple-instance bag according to the URL abnormality locating model, so as to determine whether each instance is an abnormal instance.

[0074] In an optional manner, the URL abnormality locating device further includes:

[0075] an abnormal URL determining unit 405 configured to determine whether the URL is an abnormal URL: if the multiple-instance bag corresponding to the URL includes an abnormal instance, determine that the URL is the abnormal URL; and if the multiple-instance bag corresponding to the URL includes no abnormal instance, determine that the URL is a normal URL.

[0076] In an optional manner, the instances are represented by feature vectors of corresponding fields.

[0077] In an optional manner, the field is a parameter request field in the URL.

[0078] In a fourth aspect, based on the same inventive concept, an embodiment of the present specification provides a URL abnormality locating training device. Referring to FIG. 5, the URL abnormality locating training device includes:

[0079] a sample acquiring unit 501 configured to collect a URL sample set including a plurality of URL samples;

[0080] a sample segmenting unit 502 configured to perform field segmentation on each URL sample in the URL sample set, and for each URL sample, obtain a multiple-instance bag including all instances corresponding to all fields;

[0081] an instance bag collecting unit 503 configured to collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; and

[0082] a training unit 504 configured to perform an abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm, so as to obtain the URL abnormality locating model.

[0083] In an optional manner, an abnormal instance and a normal instance are distinguished through values of abnormality labels of instances.

[0084] The training unit 504 is specifically configured to: initialize a value of an abnormality label of each instance in the multiple-instance bag set, and perform iterative learning on the value of the abnormality label to update a final value of the abnormality label of each instance.

[0085] In a fourth aspect, based on the same invention concept as the URL

abnormality locating methods in the foregoing embodiments, the present application further provides a server, as shown in FIG. 6, including a memory 604, a processor 602, and a computer program stored in the memory 604 and executed on the processor 602, when being executed by the processor 602, the program implementing the steps of any one of the URL abnormality locating methods.

[0086] In FIG. 6, in a bus architecture (represented by a bus 600), the bus 600 may include any quantity of interconnected buses and bridges, and the bus 600 connects various circuits including one or more processors represented by a processor 602 and memories represented by a memory 604. The bus 600 may further connect various other circuits such as a peripheral device, a voltage stabilizer, and a power management circuit. These are known in the art, and therefore are not further described in the present specification. A bus interface 606 provides an interface between the bus 600 and a receiver 601 and a transmitter 603. The receiver 601 and the transmitter 603 may be a same element, that is, a transceiver, providing a unit configured to communicate with various other apparatuses on a transmission medium. The processor 602 is responsible for management of the bus 600 and normal processing, and the memory 604 may be configured to store data used when the processor 602 performs an operation.

[0087] In a sixth aspect, based on the same inventive concept as the URL abnormality locating methods in the foregoing embodiments, the present application further provides a computer-readable storage medium, storing a computer program, when being executed by a processor, the program implementing the steps of any one of the URL abnormality locating methods.

[0088] This specification is described with reference to the flowchart and/or block diagram of the method, device (system), and computer program product of the embodiments of the present specification. It should be understood that computer program instructions can implement each procedure and/or block in the flowcharts and/or block diagrams and a combination of procedures and/or blocks in the flowcharts and/or block diagrams. These computer program instructions may be provided for a general-purpose computer, a special-purpose computer, an embedded processor, or a processor of any other programmable data processing device to generate a machine, so that the instructions executed by a computer or a processor of any other programmable data processing device generate a device for implementing a specific function in one or more processes in the flowcharts and/or in one or more blocks in the block diagrams.

[0089] These computer program instructions may be stored in a computer-readable

memory that can instruct the computer or any other programmable data processing device to work in a specific manner, so that the instructions stored in the computer-readable memory generate an artifact that includes an instruction device. The instruction device implements a specific function in one or more processes in the flowcharts and/or in one or more blocks in the block diagrams.

[0090] These computer program instructions may also be loaded into a computer or another programmable data processing device, so that a series of operation steps are performed on the computer or another programmable data processing device to generate processing implemented by a computer, and instructions executed on the computer or another programmable data processing device provide steps for implementing functions specified in one or more procedures in the flowcharts and/or one or more blocks in the block diagrams.

[0091] Although the exemplary embodiments of the present specification have been described, persons skilled in the art may make alterations and modifications to these embodiments once persons skilled in the art learn the basic inventive concept. Therefore, the appended claims are intended to be construed as including the exemplary embodiments and all alterations and modifications falling within the scope of the present specification.

[0092] Apparently, persons skilled in the art may make various modifications and variations to the present specification without departing from the spirit and scope of the present specification. If these modifications and variations of the present specification belong to the scope of the claims of the present specification and equivalent technologies thereof, the present specification is also intended to cover these modifications and variations.

CLAIMS

What is claimed is:

1. A URL abnormality locating method, comprising:
 - performing field segmentation on a URL to obtain a multiple-instance bag comprising all instances corresponding to all fields;
 - inputting the multiple-instance bag into a multiple-instance learning-based URL abnormality locating model to predict an abnormal instance, wherein the URL abnormality locating model is trained based on a plurality of URL samples by:
 - segmenting each of the plurality of URL samples to obtain a corresponding multiple-instance bag comprising all instances of all fields in the URL sample;
 - for an abnormal URL sample in the plurality of URL samples, randomly selecting one or more instances in the multiple-instance bag corresponding to the abnormal URL sample and assigning positive labels to the one or more instances;
 - for a normal URL sample in the plurality of URL samples, assigning negative labels to all instances in the multiple-instance bag corresponding to the normal URL sample; and
 - training the URL abnormality locating model based on the labeled instances in the multiple-instance bags of the plurality of URL samples by:
 - updating labels of instances in the multiple-instance bags based on predictions generated by a classifier; and
 - training a new classifier based on the updated labels of the instances in the multiple-instance bags
 - locating a corresponding abnormal field according to the abnormal instance.
2. The method according to claim 1, further comprising:
 - training the URL abnormality locating model based on a multiple-instance learning algorithm, to obtain the URL abnormality locating model.
3. The method according to claim 2, wherein the training the URL abnormality locating model based on the multiple-instance learning algorithm, to obtain the URL abnormality locating model comprises:
 - performing abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm; and

obtaining the URL abnormality locating model based on the classification training.

4. The method according to claim 1, wherein the inputting the multiple-instance bag into a preset URL abnormality locating model to predict an abnormal instance comprises:

predicting the value of the abnormality label of each instance in the multiple-instance bag according to the URL abnormality locating model, so as to determine whether each instance in the multiple-instance bag is an abnormal instance.

5. The method according to any of claims 1 to 4, further comprising:

determining whether the URL is abnormal:

if the multiple-instance bag corresponding to the URL comprises an abnormal instance, determining that the URL is abnormal; and

if the multiple-instance bag corresponding to the URL comprises no abnormal instance, determining that the URL is normal.

6. The method according to any of claims 1 to 4, wherein the instance is represented by a feature vector of a corresponding field.

7. The method according to any of claims 1 to 4, wherein the field is a parameter request field in the URL.

8. A URL abnormality locating training method, comprising:

collecting a URL sample set comprising a plurality of URL samples;

performing field segmentation on each URL sample in the URL sample set, and for each URL sample, obtaining a multiple-instance bag comprising all instances corresponding to all fields;

collecting multiple-instance bags of all of the URL samples;

for an abnormal URL sample in the plurality of URL samples, randomly selecting one or more instances in the multiple-instance bag corresponding to the abnormal URL sample and assigning positive labels to the one or more instances;

for a normal URL sample in the plurality of URL samples, assigning negative labels to all instances in the multiple-instance bag corresponding to the normal URL sample;

performing abnormal instance and normal instance classification training on the labeled instances in the multiple-instance bags based on a multiple-instance learning algorithm; and obtaining the URL abnormality locating model based on the classification training.

9. The method according to claim 8, wherein

the performing abnormal instance and normal instance classification training on the labeled instances in the multiple-instance bags comprises:

performing iterative learning on the value of the abnormality label to update a final value of the abnormality label of each instance.

10. A URL abnormality locating device, comprising:

a segmenting unit configured to perform field segmentation on a URL to obtain a multiple-instance bag comprising all instances corresponding to all fields;

a predicting unit configured to input the multiple-instance bag into a multiple-instance learning-based URL abnormality locating model to predict an abnormal instance, wherein the URL abnormality locating model is trained based on a plurality of URL samples by:

segmenting each of the plurality of URL samples to obtain a corresponding multiple-instance bag comprising all instances of all fields in the URL sample;

for an abnormal URL sample, randomly selecting one or more instances in the multiple-instance bag corresponding to the abnormal URL sample and assigning positive labels to the one or more instances;

for a normal URL sample, assigning negative labels to all instances in the multiple-instance bag corresponding to the normal URL sample;

training the URL abnormality locating model based on the labeled instances in the multiple-instance bags of the plurality of URL samples by:

updating labels of instances in the multiple-instance bags based on predictions generated by a classifier; and

training a new classifier based on the updated labels of the instances in the multiple-instance bags; and

a locating unit configured to locate a corresponding abnormal field according to the abnormal instance.

11. The device according to claim 10, further comprising:

a model training unit configured to train the URL abnormality locating model based on a

multiple-instance learning algorithm, to obtain the URL abnormality locating model.

12. The device according to claim 11, wherein the model training unit comprises:

a sample segmenting subunit configured to perform field segmentation on each URL sample in the plurality of URL samples, and for each URL sample, obtain a multiple-instance bag comprising all instances corresponding to all fields;

an instance bag collecting subunit configured to collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; and

a training subunit configured to perform abnormal instance and normal instance classification training on the multiple-instance bag set based on the multiple-instance learning algorithm, so as to obtain the URL abnormality locating model.

13. The device according to claim 10, wherein the predicting unit is specifically configured to: predict the value of the abnormality label of each instance in the multiple-instance bag according to the URL abnormality locating model, so as to determine whether each instance is an abnormal instance.

14. The device according to any of claims 10 to 13, further comprising:

an abnormal URL determining unit configured to determine whether the URL is abnormal: if the multiple-instance bag corresponding to the URL comprises an abnormal instance, determine that the URL is abnormal; and if the multiple-instance bag corresponding to the URL comprises no abnormal instance, determine that the URL is normal.

15. The device according to any of claims 10 to 13, wherein the instance is represented by a feature vector of a corresponding field.

16. The device according to any of claims 10 to 13, wherein the field is a parameter request field in the URL.

17. A URL abnormality locating training device, comprising:

a sample acquiring unit configured to collect a URL sample set comprising a plurality of URL samples;

a sample segmenting unit configured to perform field segmentation on each URL sample in the URL sample set, and for each URL sample, obtain a multiple-instance bag comprising

all instances corresponding to all fields;

an instance bag collecting unit configured to collect multiple-instance bags of all of the URL samples to obtain a multiple-instance bag set; and

a training unit configured to:

for an abnormal URL sample in the plurality of URL samples, randomly select one or more instances in the multiple-instance bag corresponding to the abnormal URL sample and assign positive labels to the one or more instances;

for a normal URL sample in the plurality of URL samples, assign negative labels to all instances in the multiple-instance bag corresponding to the normal URL sample; and

train the URL abnormality locating model based on the labeled instances in the multiple-instance bags of the plurality of URL samples by:

updating labels of instances in the multiple-instance bags based on predictions generated by a classifier; and

training a new classifier based on the updated labels of the instances in the multiple-instance bags.

18. A server comprising a memory, a processor, and a computer program stored on the memory and executable on the processor, when the processor executes the program, steps of the method according to any of claims 1 to 9 being implemented.

19. A computer-readable storage medium storing a computer program, when the program is executed by a processor, steps of the method according to any of claims 1 to 9 being implemented.