METHOD FOR ADJUSTING RISK PARAMETER, AND METHOD AND DEVICE FOR RISK IDENTIFICATION

TECHNICAL FIELD

This application relates to the field of information technology, and in particular to a method for adjusting a risk parameter, and a method and device for risk identification.

BACKGROUND

With the popularization of on-line payment, on-line transfer and other electronic transaction methods, it is of vital importance to ensure transaction security and timely and accurate identification of transaction risks.

Generally speaking, a user can register an account and set a transaction password on a client providing electronic transaction services (e.g., an electronic transaction application installed on a mobile phone and a browser page corresponding to an electronic transaction website), and link the account to a bank card of the user. When the user logs onto the client and inputs the transaction password in the client to request a payment, the client sends a payment request to a control server. The control server performs risk identification on the received payment request. If the result of risk control identification is safe (that is, the current transaction is started by the user or by others authorized by the user), the control server performs deduction correspondingly on the bank card of the user according to the payment request. If the result of risk control identification is unsafe, that is, there is a high probability that the current transaction is started by someone who obtains, without authorization, the account and password of the user, the control server can take measures such as instructing the client to freeze the current transaction, and send risk warning messages to a mobile phone number and email address linked to the account.

Currently, there are two risk identification methods. First, various risk identification rules are artificially defined according to known risks, and the control server performs risk identification on transactions according to the risk identification rules. Second, a risk identification system is provided with artificial intelligence on the control server. Massive historical operation data (including safe operation data and unsafe operation data) are provided to the risk identification system in advance using machine learning algorithms such as random forest, logical regression and neural network. The risk identification system is trained so as to be able to obtain the risk identification rules according to the known risks and

identify the risk of a new transaction to a certain extent.

In fact, the risk identification rules defined artificially or obtained through machine learning are too much dependent on the known risks. Unauthorized users always manage to crack the risk identification rules and constantly develop new means of on-line property theft, such that the risk identification system constantly face uncontrollable unknown risks. In the existing methods for risk identification, in order to deal with the constantly emerging means of property theft, the risk identification rules need to be updated frequently.

In the existing methods for risk identification, both artificial maintenance of the risk identification rules and updating of the risk identification rules through machine learning are only remedial measures after the risk identification rules are cracked. That is to say, the updating of the risk identification rules can only be achieved through manual inputs or retraining of the risk identification system after unauthorized users have bypassed the risk identification rules using new means of property theft to cause incalculable losses to users.

In particular, on the one hand, the cost of manual inputs is too high, and negligence and omissions in the manual inputs are inevitable. On the other hand, the retraining of the risk identification system often needs to be performed offline and takes months to complete. It can be seen that the existing methods for adjusting the risk identification rules have the problems of high cost and serious delay.

SUMMARY

Embodiments of this application provide a method for adjusting a risk parameter, and a method and device for risk identification, so as to solve the problems of high cost and serious delay of existing methods for risk identification.

In order to solve the above technical problems, the embodiments of this application are implemented as follows:

The method for adjusting a risk parameter provided by the embodiments of this application includes:

receiving a payment request, wherein the payment request contains a user identifier and operation data generated in a current transaction through the user identifier;

determining, according to the operation data, at least one operation hopping sequence produced in the current transaction, wherein the operation hopping sequence contains at least one operation hopping event;

acquiring, from a set of risk parameters corresponding to the user identifier, a risk

parameter corresponding to the operation hopping event contained in the operation hopping sequence, wherein risk parameters contained in the set of risk parameters are obtained by adjusting risk parameters used for risk identification in a previous transaction according to operation data generated in the previous transaction corresponding to the user identifier; and

adjusting the acquired risk parameter according to the at least one operation hopping sequence.

The method for risk identification provided by the embodiments of this application includes:

receiving a payment request, wherein the payment request contains operation data of at least one dimension generated in a current transaction;

with respect to a respective dimension, determining a risk degree of the current transaction according to the operation data of the respective dimension; and

when different risk degrees of the current transaction corresponding to dimensions are obtained, performing risk identification on the current transaction according to the obtained different risk degrees.

The device for adjusting a risk parameter provided by the embodiments of this application includes:

a receiving module which receives a payment request, wherein the payment request contains a user identifier and operation data generated in a current transaction through the user identifier;

a determining module which determines, according to the operation data, at least one operation hopping sequence produced in the current transaction, wherein the operation hopping sequence contains at least one operation hopping event;

an acquiring module which acquires, from a set of risk parameters corresponding to the user identifier, a risk parameter corresponding to the operation hopping event contained in the operation hopping sequence, wherein risk parameters contained in the set of risk parameters are obtained by adjusting risk parameters used for risk identification in a previous transaction according to operation data generated in the previous transaction corresponding to the user identifier; and

an adjusting module which adjusts the acquired risk parameter according to the at least one operation hopping sequence.

Another device for risk identification provided by the embodiments of this application includes:

a receiving module which receives a payment request, wherein the payment request

contains operation data of at least one dimension generated in a current transaction;

a determining module which determines, with respect to a respective dimension, a risk degree of the current transaction according to the operation data of the respective dimension; and

an identifying module which performs, when different risk degrees of the current transaction corresponding to dimensions are obtained, risk identification on the current transaction according to the obtained different risk degrees.

A risk identification system provided by the embodiments of this application includes:

a smart module which receives a transaction request, acquires operation data generated in a current transaction through a user identifier, performs risk analysis according to the operation data, and sends a risk analysis result to a risk identification module, wherein the risk analysis result is obtained by the above methods for risk identification; and

the risk identification module which performs risk identification according to the received risk analysis result.

It can be seen from the technical solution provided by the embodiments of this application that in the embodiments of this application, on the one hand, operation data related to one transaction is converted into at least one operation hopping sequence, each operation hopping sequence further includes at least one operation hopping event, each operation hopping event corresponds to a risk parameter, and then, the operation data related to one transaction is actually discretized into a plurality of independent operation hopping events; and on the other hand, with respect to various user identifiers registered on a client providing electronic trading services, a set of various risk parameters in one-to-one correspondence to the various user identifiers is set, and real-time on-line updating of the various risk parameters is supported. In this way, once unauthorized users obtain the user identifier of a user and initiates a payment request to the control server on the client through the user identifier, even if the unauthorized users temporarily deceive the risk identification system and cause property losses to the user, the control server can timely dynamically update the risk parameters corresponding to each operation hopping event involved in the transaction initiated through the user identifier on line according to the operation data contained in the payment request, without waiting for manual inputs or offline retraining of the risk identification system. By means of the method for risk identification as claimed in this application, on the one hand, the high-cost manual inputs is avoided, and risk parameters can be dynamically updated on line without waiting for several months for a risk identification system to be retrained, thus completing the inputs for the risk identification

system in a timely manner.

BRIEF DESCRIPTION OF THE DRAWINGS

To describe the technical solutions in the embodiments of this application or in the prior art more clearly, the following briefly introduces the accompanying drawings required for describing the embodiments or the prior art. Apparently, the accompanying drawings in the following description show merely some embodiments of this application, and a person of ordinary skill in the art may still derive other drawings from these accompanying drawings without creative efforts.

FIG. 1 is a flowchart of a method for adjusting a risk parameter provided by the embodiments of this application;

FIG. 2 is a flowchart of a method for risk identification provided by the embodiments of this application;

FIG. 3 is a diagram of a device for adjusting a risk parameter provided by the embodiments of this application;

FIG. 4 is a diagram of a device for risk identification provided by the embodiments of this application;

FIG. 5 is a diagram of a risk identification system provided by the embodiments of this application; and

FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d) are respectively detailed diagrams of the risk identification system provided by the embodiments of this application.

DETAILED DESCRIPTION OF THE INVENTION

To enable a person skilled in the art to better understand the technical solutions of this application, the technical solutions of the embodiments of this application will be described clearly and thoroughly below with reference to the accompanying drawings of the embodiments of this application. It is apparent that the described embodiments are merely some rather than all of the embodiments of this application. All other embodiments obtained by a person of ordinary skill in the art based on the embodiments of this application without creative efforts shall fall within the protection scope of this application.

In the technical field of on-line risk identification, the combination of method 1 and method 2 as described in the background technology is usually used for on-line risk identification. That is, on the one hand, a risk identification system is trained by providing

massive historical operation data to the risk identification system and using a method of machine learning, so that the risk identification system can obtain risk identification rules; and on the other hand, the professional experience of experts is manually added to the risk identification rules. However, the method for risk identification based on the combination of method 1 and method 2 still has the defects of high cost and serious delay.

Specifically, on-line risks are like games. The risk identification rules on which the risk identification system maintained by risk control personnel rely are always determined based on known risks. After obtaining, without authorization, a user identifier of a user, unauthorized users always manage to crack the risk identification rules and develop new means of property theft. If the unauthorized users succeed, the risk control personnel can timely update the risk identification system to curb the increased harm caused by the new means of property theft, so occasional success of the unauthorized users will not cause great on-line risks.

However, in fact, in the existing methods for risk identification, updates to a risk identification system are realized manually or through system retraining. Besides the problem of high cost, manual update also has the problem of serious delay to a certain extent since the risk control personnel will not update the risk identification rules until a new means of property theft causes harm which is severe enough to attract the attention of the risk control personnel. For system retraining, the problem of delay is even more serious, the reason is that the retraining of the risk identification system requires offline learning, and the cycle is often as long as several months; however, after several months, a new means of property theft will be developed by the unauthorized users, so the value of system retraining will be greatly reduced. It can be seen that the existing methods for risk identification have serious delay in dealing with new on-line risks, thus causing the existing risk identification system to always be "led by the nose" by the unauthorized users, which is very passive and often leads to a failure in preventing the expansion of on-line risks in time.

However, the method for risk identification as claimed in this application can realize the dynamic adjustment of various risk parameters in a timely and on-line manner in response to each challenge of the unauthorized users to the risk identification system in the on-line risk competition, which is equivalent to using the persistent trying of the unauthorized users for on-line training of the risk identification system in actual combat. Each risk parameter corresponds to an operation hopping event, and a risk identification system analyzes a risk of the operation hopping event involved in each transaction according to a set of risk parameters, and comprehensively identifies the risk of the transaction. In other words, the risk parameters

are a specific form of the risk identification rules, and the risk identification rules are updated after the risk parameters are adjusted.

In reality, even if the unauthorized users occasionally succeed, the risk identification system can be updated by adjusting the risk parameters in a timely manner (requiring one day at most), and a new means of property theft obtained by the occasional success of the unauthorized users will be identified in a next unauthorized transaction. In addition, as the risk parameters can always be dynamically adjusted on line, it is increasingly difficult for the unauthorized users to crack the risk identification system.

The core idea of this application is to discretize a transaction into a series of operation hopping events which can be independently evaluated for the possibility of occurrence (for the explanation of the operation hopping events, please see below for details). Since the operation performed by both the user and the unauthorized users after logging onto a client through the user identifier of the user can be discretized into a series of most basic operation hopping events, the evaluation of the possibility of occurrence of a current transaction can be realized comprehensively by respectively evaluating the possibility of occurrence of each operation hopping event involved in the transaction, thus realizing the risk identification for the current transaction. At the same time, by adjusting the risk parameter corresponding to a specified at least one operation hopping event in a preset set of operation hopping events corresponding to each user identifier, the expectation of the possibility of occurrence of the specified operation hopping event can be adjusted, thus realizing the adjustment of a logic or strategy of the risk identification rules of the risk identification system.

However, in the existing methods for risk identification, because the logic of the risk identification rules set up artificially or obtained by machine learning is closed, the risk identification rules cannot be adjusted, and the risk identification rules can only be added or replaced under the condition that the risk identification system is offline, which causes serious delay and high cost.

In a word, the method for risk identification as claimed in this application can realize real-time memorizing of on-line risks by dynamically adjusting the risk parameters according to the operation data of each transaction, and the risk identification system can learn adaptively and rely less on manpower; in addition, the risk identification system always maintains a "gaming" state with the unauthorized users, the accuracy of each risk parameter in the set of risk parameters is also time-tested, in this way, the risk identification system can always identify risks more accurately and even predict on-line risks to a certain extent.

The technical solution provided by various embodiments of this application will be

described in detail below with reference to the accompanying drawings.

FIG. 1 is a flowchart of a method for adjusting a risk parameter provided by the embodiments of this application, the method including the following steps:

S101: Receive a payment request.

An executive body of the method can be a control server of a service provider providing electronic transaction services. The control server interacts with a client installed on a terminal device of each user, receives a payment request sent by the client, sends an instruction to the client, such as sending an instruction to the client after freezing a transaction, and instructs the client to display the message "Transaction Frozen" on a screen of the terminal device.

In the embodiments of this application, a risk identification system may be deployed on the control server, and the risk identification system is used for risk identification according to the payment request. In addition to on-line self-learning and self-adjustment of risk parameters, the risk identification system also supports manual updating of the risk parameters by risk control personnel, as well as offline training.

In the embodiments of this application, the payment request contains a user identifier and operation data generated in a current transaction through the user identifier. The user identifier may be a user identifier for logging onto the client (an account of a user). The operation data may include the following data: the operation performed by a person performing the current transaction through the user identifier, the time of the current transaction, merchant information involved in the current transaction, logistics information, amount to be paid, data generated by interaction between a terminal device used in the current transaction and the outside world, etc. In short, all the data involved in the current transaction which can be acquired by the client can be the operation data described in this application.

A technical carrier involved in the payment in the embodiments of this application may include, for example, near field communication (NFC), WIFI, 3G/4G/5G, point of sale (POS) card swiping technology, two-dimensional code scanning technology, barcode scanning technology, Bluetooth, infrared, short message service (SMS), multimedia message service (MMS), etc.

S102: Determine, according to the operation data, at least one operation hopping sequence produced in the current transaction.

In the embodiments of this application, the operation hopping sequence may be a sequence containing at least one operation hopping event, and the operation hopping events

contained in the operation hopping sequence are logically connected. For example, the operation hopping sequence contains operation hopping events T1, T2, and T3. In one transaction, T1, T2, and T3 occur one after the other. T2 occurs only after T1, and T3 occurs only after T2.

An operation hopping event may be the occurrence of a behavior or an event of a transition between two states. For example, "clicking on a link in the client" is an operation hopping event, which can be "a behavior of clicking on a link occurs" or "transiting from a state where a link is not clicked to a state where the link is clicked." For convenience of description, the operation hopping event will be described below as an event of a transition between two states.

Assuming that Xiaoming opens a client, clicks on a purchase link in the client, and clicks on a "payment" button in a pop-up payment page, a control server will receive the payment request, which contains a user identifier of Xiaoming and operation data generated in the current transaction through the user identifier of Xiaoming. Assuming that the operation data is "Xiaoming used the iPhone of Xiaoli (Serial Number: f2154) to log onto the client with an account of Xiaoming at 17:20 on January 5, 2016, clicked on a purchase link of Nike store and requested the purchase of a pair of sneakers for 500 yuan at 18:05 on January 5, 2016," then, according to the operation data, at least one operation hopping sequence can be determined. The operation hopping sequence can actually reflect the probability that Xiaoming himself is the current user of the account of Xiaoming. For example, three operation hopping sequences can be determined, which are respectively related to how the operation was performed with the account of Xiaoming, when the operation was performed with the account of Xiaoming, and a number of times that the iPhone f2154 logs on with the account every day. From the three aspects, the probability that Xiaoming himself is the current user of the account of Xiaoming can be determined. Specifically, the operation hopping sequence 1 contains the following operation hopping event: Xiaoming clicked on the purchase link of Nike store and clicked on the payment button of a pair of sneakers; the operation hopping sequence 2 contains the following operation hopping event: Xiaoming clicked on the purchase link at 17:20 on January 5, 2016, and clicked on the payment button at 18:05 on January 5, 2016; and the hopping sequence 3 contains the following operation hopping event: the account of Xiaoli was logged out from the iPhone f2154, and the account of Xiaoming was logged onto the iPhone f2154.

For example, for the operation hopping event "Xiaoming clicked on the payment button of a pair of sneakers" contained in the operation hopping sequence 1, the operation hopping

event corresponds to the transition between two states, namely, state 1 "payment button not clicked" -> state 2 "payment button clicked". For another example, for the operation hopping event "the account of Xiaoli was logged out from the iPhone f2154" contained in the operation hopping sequence 3, the operation hopping event corresponds to the transition between two states, namely, state 1 "the account of Xiaoli was active on the iPhone f2154" -> "the account of Xiaoli was not active on the iPhone f2154".

Similarly, each operation hopping event in the operation hopping sequences can correspond to the transition between two states as above and will not be described again.

In summary, the relationship among the operation data, the operation hopping sequences and the operation hopping events (transiting from one state to another) is clear, that is, for a current transaction initiated using a certain user identifier, at least one operation hopping sequence involved in the current transaction can be determined according to operation data involved in the current transaction. Operation hopping events contained in the operation hopping sequence can be determined according to each operation hopping sequence involved in the current transaction.

In fact, the core idea of the technical solution as claimed in this application is that, on the one hand, operation data related to one transaction is converted into at least one operation hopping sequence, each operation hopping sequence further includes at least one operation hopping event, each operation hopping event corresponds to a risk parameter, and then, the operation data related to one transaction is actually discretized into a plurality of independent operation hopping events. On the other hand, with respect to various user identifiers registered on a client providing electronic trading services, a set of various risk parameters in one-to-one correspondence to the various user identifiers is set, and real-time on-line updating of the various risk parameters is supported. In this way, once unauthorized users obtain the user identifier of a user and initiates a payment request to the control server on the client through the user identifier, even if the unauthorized users temporarily deceive the risk identification system and cause property losses to the user, the control server can timely dynamically update the risk parameters corresponding to each operation hopping event involved in the transaction initiated through the user identifier on line according to the operation data contained in the payment request, without waiting for manual inputs or offline retraining of the risk identification system.

It can be seen that in the embodiments of this application, determining the corresponding at least one operation hopping sequence according to the operation data generated in the current transaction through the user identifier, and further determining all operation hopping

events involved in the current transaction are necessary steps for risk identification and risk parameter adjustment of the current transaction.

The embodiments of this application actually learn from the idea of "regarding a major event as at least one operation transfer sequence, that is, discretizing a major event into a plurality of operation hopping events" in the reinforcement learning theory. A plurality of operation hopping events is extracted from the operation data of a major event to be analyzed, and a machine determines the probability of occurrence corresponding to each operation hopping event through repeated trial and error, so as to make decisions on a major event according to the probability of occurrence of each operation hopping event.

However, in an embodiment provided by the technical solution as claimed in this application, inspired by the reinforcement learning theory, various events which may occur in a transaction are discretized into a series of operation hopping events. The probabilities of occurrence corresponding to each operation hopping event are initialized as initialized risk parameters, so that when a transaction occurs, the risk identification system can first analyze each operation hopping sequence involved in the transaction. So the relevant operation hopping events are determined accordingly; the probabilities of occurrence corresponding to the operation hopping events involved in the transaction are determined from the set probabilities of occurrence, and the probabilities of occurrence corresponding to the operation hopping events involved in the transaction are comprehensively considered to determine whether the transaction is reasonable and identify the risk of the transaction.

S103: Acquire, from a set of risk parameters corresponding to the user identifier, a risk parameter corresponding to the operation hopping event contained in the operation hopping sequence.

In the embodiments of this application, the risk parameter may be the probability of occurrence used in a machine reinforcement learning method, or may be a parameter obtained by risk control personnel encrypting the risk probability using some encryption algorithm in order to prevent the risk parameter from being cracked. In short, any parameter which can be used to analyze the possibility of occurrence of the operation hopping events can be the risk parameter described in this application. It is worth noting that adjustment of the risk parameters means adjustment of the risk identification rules on which the risk identification system relies.

In the embodiments of this application, before step S101, the risk identification system deployed on the control server may be initialized, for each registered user identifier, a set of operation hopping events corresponding to the user identifier may be determined, and an

initial value may be assigned to the risk parameter corresponding to each operation hopping event corresponding to the user identifier.

Specifically, different nodes (i.e. different states) may be defined, operation hopping between two different nodes (i.e. transition between two different states) may be traversed to obtain a plurality of operation hopping events, and the plurality of operation hopping events may be included in the set of operation hopping events. It is worth noting that each user identifier corresponds to a set of operation hopping events.

Then, an initial value may be randomly assigned to the risk parameter corresponding to each operation event in the set of operation hopping events, so that the risk parameter corresponding to each operation event will more conform to individual user's attributes using the corresponding user identifier in the process of the risk identification system updating the risk parameters constantly.

Of course, the initial value of each risk parameter may also be determined according to individual user's attributes using the user identifier. Specifically, historical operation data generated through the user identifier may be acquired; and for each operation hopping event contained in the set of operation hopping events, the following is executed respectively: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

For operation hopping events not included in the historical operation data, reference values may be given to the probabilities of occurrence of these operation hopping events. With the implementation of the method for risk identification, these probabilities of occurrence will be gradually adjusted to fit individual user's attributes.

When the historical operation data generated through the user identifier cannot be acquired, historical operation data generated using a set number of other user identifiers is acquired. For each operation hopping event contained in the set of operation hopping events, the following is executed respectively: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

That is, for a newly registered user identifier, since historical operation data corresponding to the user identifier cannot be obtained, historical operation data corresponding to other registered user identifiers can be used as the historical operation data corresponding to the user identifier. The probability of occurrence of each hopping event in a

set of hopping events corresponding to the user identifier can be determined according to the historical operation data.

It is worth emphasizing that after assigning initial values to the set of risk parameters in the risk operating system, to identify future risk identification for previous transactions, the risk parameters contained in the set of risk parameters may be obtained by adjusting the risk parameters used for risk identification in the previous transaction according to the operation data generated in the previous transaction corresponding to the user identifier.

In other words, in a method for risk identification for the current transaction, the set of risk parameters on which risk identification is based is the latest set of risk parameters obtained by adjusting the set of risk parameters after performing risk identification on the previous transaction. The adjusted set of risk parameters is used for risk identification for the next transaction.

S104: Adjust the acquired risk parameter according to the at least one operation hopping sequence.

In the embodiments of this application, when the operation hopping sequences involved in the current transaction are determined by analyzing the current transaction, the operation hopping events involved in the current transaction are also determined. Then, as mentioned earlier, since the risk parameters corresponding to each operation hopping event have been comprehensively established in the set of risk parameters, the risk parameters corresponding to each operation hopping event involved in the current transaction may be determined from the set of risk parameters. A risk value of the operation hopping sequence may be determined according to the risk parameters (which may be probability of occurrence) corresponding to each operation hopping event involved in the current transaction. After determining the risk value of each operation hopping sequence of the current transaction, the final risk degree of the current transaction may be determined.

In the embodiments of this application, the reason why at least one operation hopping sequence is determined according to the operation data, then the risk value of each operation hopping sequence is determined, and eventually the risk degree of the current transaction is determined according to the risk value of each operation hopping sequence. The techniques do not directly determine the risk degree according to all operation hopping events involved in the operation data. This is because the operation hopping sequence can reflect a transaction logic reflected by the operation data of the current transaction, while a single operation hopping event cannot reflect the transaction logic of the current transaction. The operation hopping sequence that includes the operation hopping events in a certain sequence has the

transaction logic. The analysis of the current transaction actually includes analyzing of the current transaction logic, and the risk identification for the current transaction is actually the judgment on whether the current transaction logic is reasonable. Therefore, even two operation hopping sequences containing the same operation hopping events may correspond to different risk values due to different occurrence sequences of the operation hopping events contained respectively.

In the embodiments of this application, the obtained risk parameters respectively correspond to the operation hopping events involved in the current transaction. The risk degree may be a numerical value, which is obtained by integrating the risk values of each operation hopping sequence involved in the current transaction, or may be a grade, which is evaluated after integrating the risk values of each operation hopping sequence involved in the current transaction.

After determining the risk degree of the current transaction, the risk identification system may determine the risk probability of the current transaction according to the risk degree, and may also classify the current transaction as safe or unsafe according to the risk degree. In a word, in the technical solution as claimed in this application, there is no specific limitation on the risk value, risk degree and an output result of risk identification.

In the embodiments of this application, after determining the risk degree of the current transaction, when it is determined that the risk degree corresponding to the current transaction is higher than a set threshold value, the acquired risk parameter may be decreased; and when it is determined that the risk degree corresponding to the current transaction is not higher than the set threshold value, the acquired risk parameter may be increased. The set threshold value may be a preset value provided by the risk control personnel as the maximum controllable risk degree. If the risk degree of the current transaction is higher than the set threshold value, it indicates that the risk of the current transaction is great, which indicates the reasonability of the occurrence of the operation hopping event involved in the current transaction should be estimated as low. That is, an operation hopping event with a low probability of occurrence is an event that the risk control personnel would consider to have a low possibility of occurrence in a safe transaction.

Similarly, if the risk degree of the current transaction is lower than the set threshold value, then the risk of the current transaction is not high, and the occurrence of the operation hopping event involved in the current transaction is reasonable and meets the established expectation. In other words, the operation hopping event involved in the current transaction is probably caused by the operation of the user rather than by the unauthorized users, and then

the risk parameter corresponding to the operation hopping event involved in the current transaction may remain unchanged or the risk parameter corresponding to the operation hopping event involved in the current transaction may be increased.

Of course, in the embodiments of this application, the acquired risk parameter may also be adjusted not according to the risk degree of the current transaction, so there is no need to wait for the determination of the risk degree. The risk parameter corresponding to each operation hopping sequence may be adjusted according to the risk value of the operation hopping sequence. For example, an operation hopping sequence with a high risk value may also indicate that when considering all the operation hopping events contained in the operation hopping sequence, the probability of occurrence is low overall.

For example, assuming that an operation hopping sequence contains operation hopping events T1, T2, T3, T4 and T5, and that the sequence of occurrence of the operation hopping events is T1, T4, T3, T5 and T2 for Xiaoming, that the probability of occurrence (risk parameter) of T1 is 0.9, that the probability of occurrence of T4 is 0.8, that the probability of occurrence of T3 is 0.1, that the probability of occurrence of T4 is 0.05, and that the probability of occurrence of T5 is 0.15, even if the probabilities of occurrence of T1 and T2 are high, the probability of T1, T4, T3, T5, T2 occurring at the same time in this order is very low. It is still determined that the risk value of the operation hopping sequence is high.

In the embodiments of this application, one operation hopping event may be defined according to each operation of the user, one operation hopping event may also be defined according to the transition of every two states involved in the transaction, or one operation hopping event may also be defined according to other dimensions besides that of the user, such as the dimension of a terminal device used by the user, the dimension of a designated merchant in the current transaction, the dimension of the current location of the user, and other available dimensions. For example, "Xiaoming clicked on the purchase link" or "the purchase link was not clicked -> the purchase link was clicked" may be an operation hopping event, "the mobile phone did not scan a two-dimensional code -> the mobile phone scanned a two-dimensional code" may be an operation hopping event, and "the Nike store sold 1 pair of sneakers per minute -> the Nike store sold 10 pairs of sneakers per minute" may be an operation hopping event.

In short, this application does not limit how to define an operation hopping event. The core of this application is to discretize one transaction into a series of operation hopping events which can be independently evaluated, so as to adjust a strategy of the risk identification system by adjusting the risk parameters of each operation hopping event.

For the same reason, this application does not limit what specific algorithm or method is used to calculate and integrate a reward function value of each operation hopping event, the risk value of each operation hopping sequence, and the risk degree of the current transaction, and the risk control personnel can flexibly make a decision according to business conditions.

To sum up, through the method for risk identification shown in FIG. 1, the risk parameters corresponding to each user identifier are respectively set for each user identifier registered on the client providing electronic transaction services. On-line, real-time updating of the risk parameters is supported. In this way, once the unauthorized users obtain the user identifier of a user and initiates a payment request to the control server on the client through the user identifier, even if the unauthorized users temporarily crack the risk parameters corresponding to the user identifier and cause property losses to the user, the control server can timely dynamically update the risk parameters corresponding to the user identifier on line according to the operation data contained in the payment request, without waiting for manual inputs or offline retraining of the risk identification system. By means of the method for risk identification as claimed in this application, on the one hand, the high-cost manual inputs is avoided, and on the other hand, risk parameters can be dynamically updated on line without waiting for several months for a risk identification system to be retrained, thus completing the updates for the risk identification system in a timely manner. At the same time, self-learning of the risk identification system is realized by dynamically adjusting the risk parameters in the set of risk parameters on line.

In addition, the existing methods for risk identification also have the problem of low accuracy. The existing methods for risk identification have obvious limitations, that is, the client and the control server rely on certain risk identification rules to identify risks of transactions initiated by different users in an across-the-board manner.

Specifically, as mentioned in the background technology, in both method 1 and method 2 among the existing methods for risk identification, the risk identification rules used by the client and the control server for risk identification are common rules applicable to all users. However, each user has his/her own characteristics that the common risk identification rules would not apply to. For example, some users like overdraft consumption and often spend 80% or more of the bank card balance at a time. For another example, some users, though under 25 years old, have strong economic ability and have a high probability of transferring over RMB 5,000 yuan at a time. For these users, the existing methods for risk identification may be inaccurate.

Even if the two methods are combined, that is, both expert experience and historical

operation data are provided to the risk identification system for training, although the accuracy of risk identification may be improved to a certain extent, the problems mentioned in the background technology still exist, and the accuracy of methods for risk identification which fail to consider individual user's attributes is still not high.

That is to say, although by collecting a large amount of historical operation data generated by each user, the risk identification rules applicable to each user can be extracted from the operation data. Since the transaction habits of the users vary greatly and the daily transaction occasions of the users are not fixed, one risk identification rule applicable to different users is often rough, which is bound to omit some individualized transaction occasions of the users.

Taking the risk identification rule "transactions involving transfers of RMB 5,000 yuan or more are not safe between 12:00 a.m. and 7:00 a.m." as an example, according to experience, it does not fit to common sense to make large amounts of on-line transfers between 12:00 a.m. and 7:00 a.m., so this risk identification rule is generally accurate. However, in modern society, people have varying work-rest schedules. Some users are just active at night, so it is entirely possible to make large amounts of on-line transactions between 12:00 a.m. and 7:00 a.m.. So for these users, this risk identification rule is inaccurate, and if the transactions made by these users are frozen according to this risk identification rule, it would cause troubles to the users.

For this reason, in the technical solution as claimed in this application, an individualized risk identification strategy is customized for each user using the client providing electronic transaction services by setting a set of risk parameters for each user identifier. As the set of risk parameters corresponding to each user identifier is customized, on the one hand, the risk identification rules relied on by the risk identification system fit the individualized characteristics of each user, thus improving the accuracy of risk identification to a certain extent. On the other hand, it is more difficult for the unauthorized users to bypass the risk identification system. The unauthorized users not only need to obtain the user identifier of the user, but also must have a certain understanding of the daily transaction habits of the user to successfully conduct unauthorized transactions.

It is worth noting that experiments show that if only the individualized transaction habits of the users are examined for risk identification, the accuracy of risk identification is still not high. That is, the transaction habits of a user are analyzed according to the historical transactions of the user on the client through the corresponding user identifier in advance as the risk identification rule corresponding to the user identifier. When a trader performs

on-line trading through the user identifier, the accuracy of risk identification is still not high in determining whether the trader is the user corresponding to the user identifier that can be judged according to the trading operation this time performed through the user identifier. The reasons are as follows:

First, though the individualized risk identification rules overcomes the defects of the "across-the-board" approach and rough risk identification rules in the existing technologies, a new drawback is caused. That is, the individualized risk identification rules are poor in stability. That is to say, the risk identification rules in the existing technologies obtained by refining the common transaction habits of user groups based on business common sense and statistics, though rough and hard to adapt to the individualized trading occasions of different users, are statistically stable (the common transaction habits of most users can be refined into common risk identification rules). On the contrary, the individualized risk identification rules are obtained according to the transaction habits of a certain user. Individual transaction habits are not statistically stable like group transaction habits. The individualized risk identification rules are more specific and detailed. However, individuals often change transaction habits under the influence of emotions or unexpected events, thus causing the individualized risk identification rules to be easily distorted and unstable.

For example, Xiaoming used to eating at McDonald's downstairs of the company at noon and paying on line through a client. But occasionally he is too busy to go downstairs and has to order takeout from other restaurants through the client, or Xiaoming has gotten tired of eating western fast food recently and has eaten Chinese food for several consecutive days and paid on line through the client. The payment behavior of Xiaoming for eating Chinese food for several consecutive days is likely to be judged as unsafe trading behavior according to the risk identification rules for Xiaoming in terms of on-line trading obtained based on historical data, but in fact, the account of Xiaoming was not stolen.

Secondly, compared with mobile phones, accounts and passwords used by users on clients are more likely to be stolen. In reality, a large number of on-line property theft cases are caused by accounts and passwords being stolen. Moreover, once logging onto the client with the account and password of the user, an unauthorized user can also see the transaction records of the user and imitate the transaction habits of the user to make unauthorized payments and transfers, which are hardly covered by the risk identification rules obtained according to the transaction habits of the users.

In view of the above, in a preferred embodiment provided by the technical solution as claimed in this application, at least one other dimension of examination is required for risk

identification besides the transaction habits of the users, so as to enhance the reliability of risk identification.

Generally speaking, the dimensions involved in one transaction by the user may include: a user identifier (corresponding to individualized transaction habits), a terminal device (according to the habits of the user of using the terminal device, the law of data interaction between the terminal device and the outside world is obtained), a merchant (merchant involved in the current transaction), a location (location of the current transaction), logistics (logistics address, consignee information, etc. specified in the current transaction), a medium (current network environment), etc.

It can be seen that the method for risk identification as claimed in this application can also comprehensively identify the risk of the current transaction by examining the risk degree of the current transaction through the user identifier from at least two dimensions. As a preferred embodiment of the technical solution as claimed in this application performs risk identification in at least two dimensions, and then performs overall risk identification according to the weight of each dimension, on the one hand, the method for risk identification for the solution can realize individualized risk identification, and on the other hand, cooperation of all dimensions can also minimize errors caused by the individualization of the risk identification rules, enhance the stability of the risk identification system, and thus improve the accuracy of risk identification.

Of course, the technical solution as claimed in this application can also adopt other dimensions to examine the current transaction, and the way to examine the current transaction under the above dimensions may not be limited to the above examples. In a word, there may be many embodiments of this application, so long as the risk of the current transaction performed through the user identifier can be examined from at least two dimensions, and this application has no restriction on this. Another method for risk identification based on multiple dimensions will be described below with reference to the attached drawings.

FIG. 2 is a flowchart of a method for risk identification provided by the embodiments of this application, the method including the following steps:

S201: Receive a payment request.

In the embodiments of this application, the payment request may carry operation data of at least one dimension generated in a current transaction, and may also carry a user identifier and operation data generated in the current transaction through the user identifier.

S202: With respect to a respective dimension, determine a risk degree of the current transaction according to the operation data of the respective dimension.

As mentioned earlier, there are at least four dimensions involved in one transaction by a user: the user identifier (corresponding to individualized transaction habits), the terminal device (according to the habits of the user of using the terminal device, the law of data interaction between the terminal device and the outside world is obtained), the merchant (merchant involved in the current transaction), the location (location of the current transaction), the logistics (logistics address, consignee information, etc. specified in the current transaction), and the medium (current network environment).

For example, the operation data of at least one dimension of the user identifier may be the operation data of the user identifier under the terminal device dimension involved in the current transaction, such as a number of accounts logged on through the terminal device every day, or the operation data under the user identifier dimension involved in the current transaction, such as the daily purchase habits of Xiaoming, and a series of action habits that Xiaoming usually performs after logging onto the client, or the operation data under the merchant and location dimensions involved in the current transaction, such as the characteristics of customers traded with the merchant every day, and the characteristics of users who often trade near the location.

The user identifier (account) dimension can reflect the individualized transaction habits of the user. When unauthorized users obtain the account of the user for trading, a risk identification system can identify the risk from the user identifier dimension. However, for the two reasons mentioned above, the accuracy of risk identification performed only from the user identifier dimension is not high.

In addition, the terminal device dimension can be used as a good supplement to improve the accuracy of risk identification. In contemporary society, terminal devices, such as mobile phones, are so popular that people almost never put them down, so they can hardly be stolen. Moreover, even if a mobile phone is stolen, the unauthorized users can hardly have access to the client in an operating system of the mobile phone to make unauthorized payments or transfers due to a password-locked screen interface of the mobile phone. Therefore, the unauthorized users often obtain the user identifier of the user by means of the network hacker technology and use the user identifier on other terminal devices. That is to say, even if the unauthorized users can use the stolen user identifier to imitate the transaction habits of the user on other terminal devices for unauthorized transactions, the unauthorized users do not use the terminal device often used by the user, thus being exposed easily. The transactions are identified as abnormal by the risk identification system from the terminal device dimension.

On the other hand, tests prove that by combining the terminal device dimension and the

user identifier dimension and according to business experience, giving different weights in risk identification to the two dimensions to form "double insurance" can effectively improve the accuracy of risk identification.

For example, the habit of Xiaoming of using his iPhone (assuming that the Serial Number is A) is that a number of accounts logged onto the client on the iPhone A daily does not exceed two. Since the unauthorized users often use their own mobile phone to log on with different accounts frequently for unauthorized transactions, if the unauthorized users obtain the account of Xiaoming, log on with the account of Xiaoming with their own iPhone B, imitate the transaction habits of Xiaoming, and spend money at merchants where Xiaoming often go, then even from the user identifier dimension, the risk identification system fails to identify the risk of this transaction. But from the terminal device dimension, the risk identification system can identify that the iPhone B through which the account of Xiaoming is logged on in the current transaction is not the iPhone A usually used by Xiaoming, and there are many accounts logged on through the iPhone B every day. Thus it is judged that the iPhone B is probably a terminal device used by the unauthorized users to conduct unauthorized transactions, and this transaction is identified as a high-risk transaction from the terminal device dimension. In addition, the risk identification system can adjust a set of risk parameters of the terminal device dimension corresponding to the user identifier of Xiaoming according to the operation data of the current transaction performed by the unauthorized users, so that the proportion of the risk parameter "a number of accounts logged onto the client on the iPhone A daily does not exceed two" in risk identification is increased.

In addition, the merchants involved in unauthorized transactions are often merchants to whom the unauthorized users sell stolen goods, which makes merchants as a dimension to help to judge the probability that a merchant involved in the current transaction is the merchant to whom the unauthorized users sell stolen goods. For example, the average daily sales amount of a jewelry store A is RMB 500,000 yuan, but the daily sales amount is stable at RMB 2 million yuan for a certain period of time, and it is suspected that there are criminals selling stolen goods on line in a concentrated way. If the account of Xiaoming is stolen by the same gang of criminals during this period of time, the risk identification system may determine that the account of Xiaoming is at a high risk of theft from the merchant dimension when detecting that the merchant involved in the current transaction is jewelry store A. That is, a high risk degree is output from the merchant dimension.

Similarly, from the logistics dimension, it can be identified whether the difference between the logistics information specified by the current transaction and the logistics

information specified by the user at ordinary times is too large, and from the media dimension, it can be identified whether the network environment of the current transaction is WIFI or mobile data, whether the service provider of the mobile data is China Mobile, China Unicom or China Telecom, and whether the difference between this information and the usual media information of the user is too large.

In the embodiments of this application, for each of the above dimensions, the risk degree of the current transaction can be determined according to the operation data of the dimension. Specifically, the method for adjusting the risk parameter shown in FIG. 1 and the method for risk identification can be adopted to determine the risk degree according to the operation data and perform risk identification. That is, at least one operation hopping sequence of the device dimension generated in the current transaction can be determined according to the operation data of the dimensions (any one of device dimension, user dimension, merchant dimension and location dimension), and the operation hopping sequence of the device dimension contains at least one operation hopping event of the device dimension; a risk parameter corresponding to the operation hopping event of the device dimension contained in the operation hopping sequence of the device dimension is determined according to the at least one operation hopping sequence of the device dimension. According to the risk parameter, the risk degree of the current transaction is determined.

S203: When different risk degrees of the current transaction corresponding to dimensions are obtained, perform risk identification on the current transaction according to the obtained different risk degrees.

In the embodiments of this application, the different risk degrees obtained can be calculated according to specific strategy rules to obtain a risk coefficient of the current transaction, and risk identification can be performed on the current transaction according to the determined risk coefficient.

The specific strategy rules may be to set different weights for different risk degrees and integrate the risk degrees to obtain the risk coefficient, or may simply add the risk degrees to obtain the risk coefficient. In short, the specific strategy rules may integrate different risk degrees to reflect the risk of the current transaction holistically.

It is worth noting that since the method shown in FIG. 2 and the method shown in FIG. 1 are based on the same inventive concept. The way to determine the risk degree in each dimension in the method shown in FIG. 2 is exactly the way to determine the risk degree of the current transaction in the method shown in FIG. 1, except that in the method shown in FIG. 2, the risk degree of the current transaction can be determined from more than one

dimension, and the risk degrees in different dimensions can be further integrated for risk identification.

When risk identification is performed on the current transaction, the more dimensions examined, the higher the accuracy of risk identification. Eventually, the risk identification system can identify the risk of the current transaction holistically according to the determined risk results of each dimension.

In the method for risk identification shown in FIG. 2, when the server receives the payment request of the current transaction, the method for risk identification shown in FIG. 1 is executed once in each of at least two dimensions. The risk degree in each dimension is output, and eventually, the risk result of the current transaction can be determined according to the risk degree in each dimension (for example, the risk degrees in different dimensions are normalized to determine the risk probability of the current transaction), thus realizing the risk identification for the current transaction. The risk parameter of each dimension may also be adjusted according to the transaction characteristics of the respective dimension.

Based on the method for adjusting the risk parameter shown in FIG. 1, the embodiments of this application also correspondingly provide a device for adjusting a risk parameter, as shown in FIG. 3, including:

a receiving module 301 which receives a payment request, wherein the payment request contains a user identifier and operation data generated in a current transaction through the user identifier;

a determining module 302 which determines, according to the operation data, at least one operation hopping sequence produced in the current transaction, wherein the operation hopping sequence contains at least one operation hopping event;

an acquiring module 303 which acquires, from a set of risk parameters corresponding to the user identifier, a risk parameter corresponding to the operation hopping event contained in the operation hopping sequence, wherein risk parameters contained in the set of risk parameters are obtained by adjusting risk parameters used for risk identification in a previous transaction according to operation data generated in the previous transaction corresponding to the user identifier; and

an adjusting module 304 which adjusts, according to the at least one operation hopping sequence, the acquired risk parameter.

The device further includes an identifying module 305 which determines a risk degree of the current transaction according to the acquired risk parameter and the at least one operation hopping sequence, and performs risk identification on the current transaction according to the

risk degree.

The adjusting module 304 decreases the acquired risk parameter when determining that the risk degree corresponding to the current transaction is higher than a set threshold value, and increases the acquired risk parameter when determining that the risk degree corresponding to the current transaction is not higher than the set threshold value, and the adjusted risk parameter is used for risk identification for the next transaction corresponding to the user identifier.

The device further includes an initializing module 306 which determines a set of operation hopping events corresponding to the user identifier and an initial value for each of the risk parameters of a corresponding operation hopping event contained in the set of operation hopping events before receiving the payment request when a risk identification system is initialized.

The initializing module 306 defines different nodes, and traverses operation hopping between the different nodes to obtain a plurality of operation hopping events. The plurality of operation hopping events involve two of the nodes. According to the plurality of operation hopping events, the set of operation hopping events is obtained.

The initializing module 306 acquires historical operation data generated through the user identifier. For each operation hopping event contained in the set of operation hopping events, the following is executed respectively: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

When the historical operation data generated through the user identifier cannot be acquired, the initializing module 306 acquires historical operation data generated using a set number of other user identifiers. For each operation hopping event contained in the set of operation hopping events, the following is executed respectively: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

The identifying module 305 determines the operation hopping events contained in each operation hopping sequence, and determines the risk value of the operation hopping sequence according to the risk parameter corresponding to each operation hopping event. After obtaining the risk value of each operation hopping sequence, determines the risk degree of the current transaction according to the risk value of each operation hopping sequence.

Based on the method for risk identification shown in FIG. 2, this application also correspondingly provides a device for risk identification, as shown in FIG. 4, including:

a receiving module 401 which receives a payment request, wherein the payment request contains operation data of at least one dimension generated in a current transaction;

a determining module 402 which determines, with respect to a respective dimension, a risk degree of the current transaction according to the operation data of the respective dimension; and

an identifying module 403 which performs, when different risk degrees of the current transaction corresponding to dimensions are obtained, risk identification on the current transaction according to the obtained different risk degrees.

If the operation data of the respective dimension is the operation data of a device dimension, the determining module 402 determines at least one operation hopping sequence of the device dimension generated in the current transaction according to the operation data of the device dimension, wherein the operation hopping sequence of the device dimension contains at least one operation hopping event of the device dimension; determines a risk parameter corresponding to the operation hopping event of the device dimension contained in the operation hopping sequence of the device dimension according to the at least one operation hopping sequence of the device dimension; and determines the risk degree of the current transaction according to the risk parameter.

If the operation data of the respective dimension is the operation data of a user dimension, the determining module 402 determines at least one operation hopping sequence of the user dimension generated in the current transaction according to the operation data of the user dimension, wherein the operation hopping sequence of the user dimension contains at least one operation hopping event of the user dimension; determines a risk parameter corresponding to the operation hopping event of the user dimension contained in the operation hopping sequence of the user dimension according to the at least one operation hopping sequence of the user dimension; and determines the risk degree of the current transaction according to the risk parameter.

If the operation data of the respective dimension is the operation data of a merchant dimension, the determining module 402 determines at least one operation hopping sequence of the merchant dimension generated in the current transaction according to the operation data of the merchant dimension, wherein the operation hopping sequence of the merchant dimension contains at least one operation hopping event of the merchant dimension; determines a risk parameter corresponding to the operation hopping event of the merchant

dimension contained in the operation hopping sequence of the merchant dimension according to the at least one operation hopping sequence of the merchant dimension; and determines the risk degree of the current transaction according to the risk parameter.

If the operation data of the respective dimension is the operation data of a location dimension, the determining module 402 determines at least one operation hopping sequence of the location dimension generated in the current transaction according to the operation data of the location dimension, wherein the operation hopping sequence of the location dimension contains at least one operation hopping event of the location dimension; determines a risk parameter corresponding to the operation hopping event of the location dimension contained in the operation hopping sequence of the location dimension according to the at least one operation hopping sequence of the location dimension; and determines the risk degree of the current transaction according to the risk parameter.

The identifying module 403 calculates the different risk degrees obtained according to specific strategy rules to obtain a risk coefficient of the current transaction, and performs risk identification on the current transaction according to the determined risk coefficient.

FIG. 5 is a diagram of a risk identification system provided by the embodiments of this application, and the system includes:

a smart module 501 which acquires operation data generated in a current transaction through a user identifier, performs risk analysis according to the operation data, and sends a risk analysis result to a risk identification module, wherein a number of the smart modules is at least one, and the risk analysis result is obtained by the above method for risk identification; and

the risk identification module 502 which performs risk identification according to the received risk analysis result.

The smart module 501 specifically includes:

an acquiring unit 5011 which acquires the operation data generated in the current transaction through the user identifier, determines at least one operation hopping sequence according to the operation data, and sends the at least one operation hopping sequence to an analyzing unit;

the analyzing unit 5012 which performs risk analysis according to the received operation hopping sequence, and outputs the risk analysis result to a sending unit; and

the sending unit 5013 which determines the risk degree according to the received risk analysis result and sends the risk degree to the risk identification module.

FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d) are detailed diagrams of the risk

identification system provided by the embodiments of this application. It is worth emphasizing that a number of the smart modules of the risk identification system shown in FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d) is at most 4, respectively corresponding to four dimensions of user, terminal device, merchant and location, but this does not constitute a limitation to this application.

As can be seen from FIG. 6(a), the risk identification system can collect operation data of one dimension and identify whether there is risk in the current transaction according to the collected operation data. The specific identification method has been described in the above embodiment. The operation data of one dimension in FIG. 6(a) includes the operation data of the user dimension as an example.

As can be seen from FIG. 6(b), the risk identification system can collect operation data of two dimensions and identify whether there is risk in the current transaction according to the collected operation data. The specific identification method has been described in the above embodiments. The operation data of two dimensions in FIG. 6(b) includes the operation data of the user dimension and the operation data of the device dimension as an example.

As can be seen from FIG. 6(c), the risk identification system can collect operation data of three dimensions and identify whether there is risk in the current transaction according to the collected operation data. The specific identification method has been described in the above embodiment. The operation data of three dimensions in FIG. 6(c) includes the operation data of the user dimension, the operation data of the device dimension and the operation data of the merchant dimension as an example.

As can be seen from FIG. 6(d), the risk identification system can collect operation data of four dimensions and identify whether there is risk in the current transaction according to the collected operation data. The specific identification method has been described in the above embodiment. The operation data of four dimensions in FIG. 6(d) includes the operation data of the user dimension, the operation data of the device dimension, the operation data of the merchant dimension and the operation data of the location dimension as an example.

In the risk identification system shown in FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d), the analyzing unit determines a risk value for each received operation hopping sequence, and outputs the risk value to the sending unit as a risk analysis result. The sending unit can determine the risk degree according to each received risk analysis result, send the risk degree to the risk identification module for final decision, and output a risk evaluation (e.g., risk probability) for the current transaction.

In the risk identification system shown in FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d), the smart modules can communicate and link with each other. For example, when unauthorized users sell stolen goods to a certain merchant. After finding something unusual, the smart module corresponding to the merchant dimension can inform other smart modules of this new risk, and other smart modules can adjust their risk identification strategies in time according to this new risk. As the risk identification system has this feedback mechanism between modules, the set of risk parameters of each module can be kept up to date all the time, and can respond to the latest on-line risks more quickly and achieve risk identification more accurately.

The risk identification system shown in FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d) has strong robustness. That is to say, since the risk identification system may include more than one smart module, when an abnormality occurs in one smart module, other smart modules will not be affected. The risk identification module may still receive the risk degree sent by at least one smart module for risk identification.

In addition, the risk identification system shown in FIG. 6(a), FIG. 6(b), FIG. 6(c), and FIG. 6(d) may also have a corresponding test system, i.e., adopting an A/B Test on-line mode. The test system and the risk identification system have the same architecture. The test system only performs risk identification on transactions corresponding to a few user identifiers. When new risks occur, updates may be performed on the test system first. After observing whether the performance of the test system is stable, risk control personnel can decide whether to perform the same updates on the risk identification system.

In the 1990s, improvements of a technology can be clearly distinguished between hardware improvements (for example, improvements to a circuit structure such as a diode, a transistor, a switch, etc.) and software improvements (improvements to a method procedure). However, with the development of technology, improvements of many method procedures can be considered as direct improvements of hardware circuit structures. Designers almost all program an improved method procedure to a hardware circuit, to obtain a corresponding hardware circuit structure. Therefore, it does not mean that the improvement of a method procedure cannot be implemented by using a hardware entity module. For example, a programmable logic device (PLD) such as a field programmable gate array (FPGA) is a type of integrated circuit whose logic function is determined by a user by programming the device. The designers perform voluntary programming to "integrate" a digital system into a single PLD without requiring a chip manufacturer to design and prepare a dedicated integrated circuit chip. In addition, instead of making an integrated circuit chip manually, the

programming is mostly implemented by using "logic compiler software, which is similar to the software compiler used to write programs. Original code before compiling is also written in a specific programming language, which is referred to as Hardware Description Language (HDL). There are many types of HDLs, such as Advanced Boolean Expression Language (ABEL), Altera Hardware Description Language (AHDL), Confluence, Cornell University Programming Language (CUPL), HDCal, Java Hardware Description Language (JHDL), Lava, Lola, MyHDL, PALASM, Ruby Hardware Description Language (RHDL), etc. Currently, Very-High-Speed Integrated Circuit Hardware Description Language (VHDL) and Verilog are most commonly used. A person skilled in the art should also understand that as long as a method procedure is logically programmed and then programmed to an integrated circuit by using the foregoing hardware description languages, a hardware circuit that implements the logical method procedure can be easily obtained.

The controller can be implemented in any suitable manner, for example, the controller can take the form of, for example, a microprocessor or processor and a computer-readable medium storing computer-readable program code (for example, software or firmware) executable by the processor, a logic gate, a switch, an application-specific integrated circuit (ASIC), a programmable logic controller and an embedded microcontroller. Examples of the controller include, but are not limited to, the following microcontrollers: ARC 625D, Atmel AT91SAM, Microchip PIC18F26K20 and Silicone Labs C8051F320. The memory controller can also be implemented as part of the memory control logic. A person skilled in the art will also appreciate that, in addition to implementing the controller in the form of pure computer-readable program code, it is also possible to implement the controller in the form of a logic gate, switch, application-specific integrated circuit, programmable logic controller, and embedded microcontroller and other forms to achieve the same function. Such a controller can thus be considered as a hardware component and apparatuses included therein for implementing various functions can also be considered as structures inside the hardware component. Alternatively, apparatuses configured to implement various functions can be considered as both software modules implementing the method and structures inside the hardware component.

The system, the apparatus, the module or the unit described in the foregoing embodiments can be specifically implemented by a computer chip or an entity or implemented by a product having a certain function. A typical implementation device is a computer. Specifically, the computer can be, for example, a personal computer, a laptop computer, a cellular phone, a camera phone, a smartphone, a personal digital assistant, a

media player, a navigation device, an email device, a game console, a tablet computer, a wearable device, or a combination of any of these devices.

For ease of description, when the apparatus is described, the apparatus is divided into units according to functions, which are separately described. Certainly, in implementation of this application, the function of the units may be implemented in a same piece of or multiple pieces of software and/or hardware.

A person skilled in the art should understand that the embodiments of the present invention may be provided as a method, a system, or a computer program product. Therefore, the present invention may use a form of hardware only embodiments, software only embodiments, or embodiments with a combination of software and hardware. Moreover, the present invention may use a form of a computer program product that is implemented on one or more computer-usable storage media (including but not limited to a disk memory, a CD-ROM, an optical memory, and the like) that include computer-usable program code.

The present invention is described with reference to the flowcharts and/or block diagrams of the method, the device (system), and the computer program product according to the embodiments of the present invention. It should be understood that computer program instructions may be used for implementing each process and/or each block in the flowcharts and/or the block diagrams and a combination of a process and/or a block in the flowcharts and/or the block diagrams. These computer program instructions may be provided for a general-purpose computer, a dedicated 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 an apparatus for implementing a specific function in one or more processes in the flowcharts and/or in one or more blocks in the block diagrams.

These computer program instructions may further 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 apparatus. The instruction apparatus implements a specified function in one or more processes in the flowcharts and/or in one or more blocks in the block diagrams.

These computer program instructions may further be loaded onto a computer or another programmable data processing device, so that a series of operations and steps are performed on the computer or another programmable device, thereby generating computer-implemented processing. Therefore, the instructions executed on the computer or another programmable

device provide steps for implementing a specific function in one or more processes in the flowcharts and/or in one or more blocks in the block diagrams.

In a typical configuration, the computer device includes one or more processors (CPUs), an input/output interface, a network interface, and a memory.

The memory may include, among computer readable media, a non-persistent memory such as a random access memory (RAM) and/or a non-volatile memory such as a read-only memory (ROM) or a flash memory (flash RAM). The memory is an example of the computer-readable medium.

The computer-readable medium includes persistent, non-persistent, movable, and unmovable media that may implement information storage by using any method or technology. Information may be a computer-readable instruction, a data structure, a program module, or other data. Examples of computer storage media include but are not limited to a phase change memory (PRAM), a static random access memory (SRAM), a dynamic random access memory (DRAM), other type of random access memory (RAM), a read-only memory (ROM), an electrically erasable programmable read-only memory (EEPROM), a flash memory or other memory technology, a compact disc read-only memory (CD-ROM), a digital versatile disc (DVD) or other optical storage, a cassette magnetic tape, tape and disk storage or other magnetic storage device or any other non-transmission media that may be configured to store information that a computing device can access. Based on the definition in the present disclosure, the computer-readable medium does not include transitory computer readable media (transitory media), such as a modulated data signal and a carrier.

It should also be noted that the terms "include", "comprise" and any other variants mean to cover the non-exclusive inclusion. Thereby, the process, method, article, or device which include a series of elements not only include those elements, but also include other elements which are not clearly listed, or include the inherent elements of the process, method, article and device. Without further limitation, the element defined by a phrase "include one....." does not exclude other same elements in the process, method, article or device which include the element.

A person skilled in the art should understand that the embodiments of this application may be provided as a method, a system, or a computer program product. Therefore, this application may use a form of hardware only embodiments, software only embodiments, or embodiments with a combination of software and hardware. Moreover, this application may use a form of a computer program product that is implemented on one or more computer-usable storage media (including but not limited to a disk memory, a CD-ROM, an

optical memory, and the like) that include computer-usable program code.

This application can be described in the general context of computer executable instructions executed by a computer, for example, a program module. Generally, the program module includes a routine, a program, an object, a component, a data structure, and the like for executing a particular task or implementing a particular abstract data type. This application can also be practiced in a distributed computing environment in which tasks are performed by remote processing devices that are connected through a communication network. In a distributed computing environment, the program module may be located in both local and remote computer storage media including storage devices.

The embodiments of the present disclosure are all described in a progressive manner. For same or similar parts in the embodiments, reference may be made to these embodiments. Each embodiment focuses on a difference from other embodiments. Especially, a system embodiment is basically similar to a method embodiment, and therefore is described briefly; for related parts, reference may be made to partial descriptions in the method embodiment.

The foregoing descriptions are merely embodiments of this application and are not intended to limit this application. For a person skilled in the art, various modifications and variations can be made to this application. Any modification, equivalent replacement, or improvement made without departing from the spirit and principle of this application shall fall within the scope of the claims of this application.

CLAIMS

What is claimed is:

1. A method for adjusting a risk parameter, comprising:

receiving a payment request, wherein the payment request contains a user identifier and operation data generated in a current transaction through the user identifier;

determining, according to the operation data, at least one operation hopping sequence produced in the current transaction, wherein the operation hopping sequence contains at least one operation hopping event;

acquiring, from a set of risk parameters corresponding to the user identifier, a risk parameter corresponding to the operation hopping event contained in the operation hopping sequence, wherein risk parameters contained in the set of risk parameters are obtained by adjusting risk parameters used for risk identification in a previous transaction according to operation data generated in the previous transaction corresponding to the user identifier; and

adjusting the acquired risk parameter according to the at least one operation hopping sequence.

2. The method according to claim 1, further comprising:

determining a risk degree of the current transaction according to the acquired risk parameter and the at least one operation hopping sequence; and

performing risk identification on the current transaction according to the risk degree.

3. The method according to claim 2, wherein adjusting the acquired risk parameter according to the at least one operation hopping sequence comprises:

decreasing the acquired risk parameter when determining that the risk degree corresponding to the current transaction is higher than a set threshold value, and

increasing the acquired risk parameter when determining that the risk degree corresponding to the current transaction is not higher than the set threshold value, wherein

the adjusted risk parameter is used for risk identification for a next transaction corresponding to the user identifier.

4. The method according to claim 2, wherein before receiving the payment request, the method further comprises:

determining a set of operation hopping events corresponding to the user identifier and an

initial value for each of the risk parameters of a corresponding operation hopping event contained in the set of operation hopping events when a risk identification system is initialized.

5. The method according to claim 4, wherein determining the set of operation hopping events comprises:

defining different nodes;

traversing operation hopping between the different nodes to obtain a plurality of operation hopping events, wherein the plurality of operation hopping events involve two of the nodes; and

obtaining the set of operation hopping events according to the plurality of operation hopping events.

6. The method according to claim 4, wherein determining the initial value for each of the risk parameters of a corresponding operation hopping event contained in the set of operation hopping events comprises:

acquiring historical operation data generated through the user identifier;

for each operation hopping event contained in the set of operation hopping events, executing:

determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and

using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

7. The method according to claim 4, wherein determining the initial value for each of the risk parameters of a corresponding operation hopping event contained in the set of operation hopping events comprises:

when the historical operation data generated through the user identifier cannot be acquired, acquiring historical operation data generated using a set number of other user identifiers;

for each operation hopping event contained in the set of operation hopping events, executing:

determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and

using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

8. The method according to claim 6 or 7, wherein determining the risk degree of the current transaction according to the acquired risk parameter and the at least one operation hopping sequence comprises:

determining operation hopping events contained in each operation hopping sequence, and determining a risk value of the operation hopping sequence according to the risk parameter corresponding to each operation hopping event; and

after obtaining the risk value of each operation hopping sequence, determining the risk degree of the current transaction according to the risk value of each operation hopping sequence.

9. A method for risk identification, comprising:

receiving a payment request, wherein the payment request contains operation data of at least one dimension generated in a current transaction;

with respect to a respective dimension, determining a risk degree of the current transaction according to the operation data of the respective dimension; and

when different risk degrees of the current transaction corresponding to dimensions are obtained, performing risk identification on the current transaction according to the obtained different risk degrees.

10. The method according to claim 9, wherein determining the risk degree of the current transaction according to the operation data of the respective dimension specifically comprises:

if the operation data of the respective dimension is the operation data of a device dimension, determining at least one operation hopping sequence of the device dimension generated in the current transaction according to the operation data of the device dimension, wherein the operation hopping sequence of the device dimension contains at least one operation hopping event of the device dimension;

determining a risk parameter corresponding to the operation hopping event of the device dimension contained in the operation hopping sequence of the device dimension according to the at least one operation hopping sequence of the device dimension; and

determining the risk degree of the current transaction according to the risk parameter.

11. The method according to claim 10, wherein determining the risk degree of the current transaction according to the operation data of the respective dimension further comprises:

if the operation data of the respective dimension is the operation data of a user dimension, determining at least one operation hopping sequence of the user dimension generated in the current transaction according to the operation data of the user dimension, wherein the operation hopping sequence of the user dimension contains at least one operation hopping event of the user dimension;

determining a risk parameter corresponding to the operation hopping event of the user dimension contained in the operation hopping sequence of the user dimension according to the at least one operation hopping sequence of the user dimension; and

determining the risk degree of the current transaction according to the risk parameter.

12. The method according to claim 11, wherein determining the risk degree of the current transaction according to the operation data of the respective dimension further comprises:

if the operation data of the respective dimension is the operation data of a merchant dimension, determining at least one operation hopping sequence of the merchant dimension generated in the current transaction according to the operation data of the merchant dimension, wherein the operation hopping sequence of the merchant dimension contains at least one operation hopping event of the merchant dimension;

determining a risk parameter corresponding to the operation hopping event of the merchant dimension contained in the operation hopping sequence of the merchant dimension according to the at least one operation hopping sequence of the merchant dimension; and

determining the risk degree of the current transaction according to the risk parameter.

13. The method according to claim 12, wherein determining the risk degree of the current transaction according to the operation data of the respective dimension further comprises:

if the operation data of the respective dimension is the operation data of a location dimension, determining at least one operation hopping sequence of the location dimension generated in the current transaction according to the operation data of the location dimension, wherein the operation hopping sequence of the location dimension contains at least one

operation hopping event of the location dimension;

determining a risk parameter corresponding to the operation hopping event of the location dimension contained in the operation hopping sequence of the location dimension according to the at least one operation hopping sequence of the location dimension; and

determining the risk degree of the current transaction according to the risk parameter.

14. The method according to claim 9, wherein performing risk identification on the current transaction according to the obtained different risk degrees further comprises:

calculating the different risk degrees obtained according to specific strategy rules to obtain a risk coefficient of the current transaction; and

performing risk identification on the current transaction according to the determined risk coefficient.

15. A device for adjusting a risk parameter, comprising:

a receiving module which receives a payment request, wherein the payment request contains a user identifier and operation data generated in a current transaction through the user identifier;

a determining module which determines, according to the operation data, at least one operation hopping sequence produced in the current transaction, wherein the operation hopping sequence contains at least one operation hopping event;

an acquiring module which acquires, from a set of risk parameters corresponding to the user identifier, a risk parameter corresponding to the operation hopping event contained in the operation hopping sequence, wherein risk parameters contained in the set of risk parameters are obtained by adjusting risk parameters used for risk identification in a previous transaction according to operation data generated in the previous transaction corresponding to the user identifier; and

an adjusting module which adjusts the acquired risk parameter according to the at least one operation hopping sequence.

16. The device according to claim 15, further comprising an identifying module, wherein

the identifying module determines a risk degree of the current transaction according to the acquired risk parameter and the at least one operation hopping sequence, and performs risk identification on the current transaction according to the risk degree.

17. The device according to claim 16, wherein,

the adjusting module decreases the acquired risk parameter when determining that the risk degree corresponding to the current transaction is higher than a set threshold value, and increases the acquired risk parameter when determining that the risk degree corresponding to the current transaction is not higher than the set threshold value, and the adjusted risk parameter is used for risk identification for a next transaction corresponding to the user identifier.

18. The device according to claim 16, further comprising an initializing module, wherein,

the initializing module determines a set of operation hopping events corresponding to the user identifier and an initial value for each of the risk parameters of a corresponding operation hopping event contained in the set of operation hopping events before receiving the payment request when a risk identification system is initialized.

19. The device according to claim 19, wherein

the initializing module defines different nodes; traverses operation hopping between the different nodes to obtain a plurality of operation hopping events, wherein the plurality of operation hopping events involve two of the nodes; and obtains the set of operation hopping events according to the plurality of operation hopping events.

20. The device according to claim 19, wherein

the initializing module acquires historical operation data generated through the user identifier; and for each operation hopping event contained in the set of operation hopping events, the initializing module executes: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

21. The device according to claim 19, wherein

when the historical operation data generated through the user identifier cannot be acquired, the initializing module acquires historical operation data generated using a set number of other user identifiers; and for each operation hopping event contained in the set of

operation hopping events, the initializing module executes: determining a probability of occurrence of a respective operation hopping event according to the historical operation data; and using the probability of occurrence as the initial value of the risk parameter for the respective operation hopping event.

22. The device according to claim 19 or 20, wherein

the identifying module determines the operation hopping events contained in each operation hopping sequence, and determines a risk value of the operation hopping sequence according to the risk parameter corresponding to each operation hopping event; and after obtaining the risk value of each operation hopping sequence, determines a risk degree of the current transaction according to the risk value of each operation hopping sequence.

23. A device for risk identification, comprising:

a receiving module which receives a payment request, wherein the payment request contains operation data of at least one dimension generated in a current transaction;

a determining module which determines, with respect to a respective dimension, a risk degree of the current transaction according to the operation data of the respective dimension; and

an identifying module which performs, when different risk degrees of the current transaction corresponding to dimensions are obtained, risk identification on the current transaction according to the obtained different risk degrees.

24. The device according to claim 23, wherein if the operation data of the respective dimension is the operation data of a device dimension, the determining module determines at least one operation hopping sequence of the device dimension generated in the current transaction according to the operation data of the device dimension, wherein the operation hopping sequence of the device dimension contains at least one operation hopping event of the device dimension; determines a risk parameter corresponding to the operation hopping event of the device dimension contained in the operation hopping sequence of the device dimension according to the at least one operation hopping sequence of the device dimension; and determines the risk degree of the current transaction according to the risk parameter.

25. The device according to claim 24, wherein if the operation data of the respective dimension is the operation data of a user dimension, the determining module determines at

least one operation hopping sequence of the user dimension generated in the current transaction according to the operation data of the user dimension, wherein the operation hopping sequence of the user dimension contains at least one operation hopping event of the user dimension; determines a risk parameter corresponding to the operation hopping event of the user dimension contained in the operation hopping sequence of the user dimension according to the at least one operation hopping sequence of the user dimension; and determines the risk degree of the current transaction according to the risk parameter.

26. The device according to claim 25, wherein if the operation data of the respective dimension is the operation data of a merchant dimension, the determining module determines at least one operation hopping sequence of the merchant dimension generated in the current transaction according to the operation data of the merchant dimension, wherein the operation hopping sequence of the merchant dimension contains at least one operation hopping event of the merchant dimension; determines a risk parameter corresponding to the operation hopping event of the merchant dimension contained in the operation hopping sequence of the merchant dimension according to the at least one operation hopping sequence of the merchant dimension; and determines the risk degree of the current transaction according to the risk parameter.

27. The device according to claim 26, wherein if the operation data of the respective dimension is the operation data of a location dimension, the determining module determines at least one operation hopping sequence of the location dimension generated in the current transaction according to the operation data of the location dimension, wherein the operation hopping sequence of the location dimension contains at least one operation hopping event of the location dimension; determines a risk parameter corresponding to the operation hopping event of the location dimension contained in the operation hopping sequence of the location dimension according to the at least one operation hopping sequence of the location dimension; and determines the risk degree of the current transaction according to the risk parameter.

28. The device according to claim 23, wherein the identifying module calculates the different risk degrees obtained according to specific strategy rules to obtain a risk coefficient of the current transaction, and performs risk identification on the current transaction according to the determined risk coefficient.

29. A risk identification system, comprising:

at least one smart module which receives a transaction request, acquires operation data generated in a current transaction through a user identifier, performs risk analysis according to the operation data, and sends a risk analysis result to a risk identification module, wherein the risk analysis result is obtained by the method according to claims 1 to 8; and

the risk identification module which performs risk identification according to the received risk analysis result.

30. The system according to claim 29, wherein the smart module specifically comprises:

an acquiring unit which acquires the operation data generated in the current transaction through the user identifier, determines at least one operation hopping sequence according to the operation data, and sends the at least one operation hopping sequence to an analyzing unit;

the analyzing unit which performs risk analysis according to the received operation hopping sequence, and outputs the risk analysis result to a sending unit; and

the sending unit which determines a risk degree according to the received risk analysis result and sends the risk degree to the risk identification module.