METHOD AND SYSTEM FOR ENROLLING A VOICEPRINT IN A FRAUDSTER DATABASE

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ABSTRACT

Disclosed is a method for enrolling a voiceprint in a fraudster database, the method comprising: a) defining a fraud model comprising at least one hypothesis indicative of a fraudulent transaction; b) processing audio data based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster; and c) enrolling the at least one suspect voiceprint in the fraudster database.
Figure 2
Identify individuals who called multiple times

Did the individual who called multiple times use multiple caller identities?

The individual is a fraudster
Next person

Pick an unassigned audio $a_n$

Generate voice print audio/model $F_y$ for $a_n$

Screen all remaining unassigned audios against $F_y$ and determine which ones match

Mark audios which match to $F_y$

Are all $a_n$ assigned?

$y = y$

$y = 1$

$y++$

$FIG. 4$
Create (temporary) voice prints \( v_{pn} \) for each \( a_n \) \((n=1\text{ to } N)\)

Screen each \( a_n \) against each \( v_{pi} \), \((i=1\text{ to } n)\) to generate \( N^*(N-1) \) scores representing likelihood that \( a_n \) and \( v_{pi} \) are same individual.

Group \( a_n \) based on similarity to each other. Grouping results in \( Y \) groups, each group representing a unique individual.

Generate a voice print \( F_y \) for each group.

\( Y \) individuals are identified.

\textit{Figure 5}
Select transaction audio data based on selection criteria

Define a fraud model comprising at least one hypothesis indicative of a fraudulent transaction, wherein the hypothesis comprises a single voiceprint being matched to multiple identities

Process the transaction audio data based on the fraud model to identify at least one suspect voiceprint in the transaction audio data suspected of belonging to a fraudster

Is the hypothesis true?

No fraudster in the transaction audio data

Enroll each suspect voiceprint in the fraudster database

Figure 6
METHOD AND SYSTEM FOR ENROLLING A VOICEPRINT IN A FRAUDSTER DATABASE

CROSS-REFERENCE TO RELATED APPLICATION


TECHNICAL FIELD OF THE DISCLOSURE

[0002] Embodiments of the disclosure relate to a method and system to detect fraud such as credit card fraud.

BACKGROUND OF THE DISCLOSURE

[0003] Modern enterprises such as merchants, banks, insurance companies, telecommunications companies, and payments companies are susceptible to many forms of fraud. One example of fraud that is particularly pernicious is credit card fraud. With credit card fraud, a fraudster fraudulently uses a credit card or credit card credentials (name, expiration, etc.) of another to enter into a transaction for goods or services with a merchant. The merchant provides the goods or services, but since the transaction is with the fraudster the merchant runs the risk of not getting paid.

[0004] Unfortunately, the process of detecting an instance of credit card fraud committed with a particular credit card may take months after the date of transaction. For example, the credit card holder must detect the fraudulent transaction and notify the credit card company that issued the credit card. The credit card company must then process the notification and report that the credit card in question is compromised by sending a fraud report to merchants. During that time a fraudster may perpetrate further instances of credit card fraud with said credit card.

[0005] Another example of fraud that is very difficult for merchants, particularly large merchants, to detect, if at all, occurs in the job application process where an applicant has been designated as undesirable in the past—perhaps as a result of having been fired from the employ of the merchant at one location or for failing a criminal background check—fraudulently assumes a different identity and then applies for a job with the same merchant at a different location. In such cases, failure to detect the fraud could result in the rehiring of the fraudster to the detriment of the merchant. If the fraudster has assumed a new identity, background checks based on identity factors such as names or social security numbers become essentially useless. For example consider that case of a large chain store, such as, for example, Walmart. In this case, an employee can be terminated for say theft at one location, but then rehired under a different identity at another location. The employee represents a grave security risk to the company particularly since the employee, being familiar with the company’s systems and internal procedures will be able to engage in further conduct injurious to the company.

SUMMARY OF THE DISCLOSURE

[0006] In one aspect, the present disclosure provides a method for enrolling a voiceprint in a fraudster database, the method comprising: a) defining a fraud model comprising at least one hypothesis indicative of a fraudulent transaction; b) processing, audio data based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster; and c) enrolling the at least one suspect voiceprint in the fraudster database.

[0007] In another aspect, the present disclosure provides a system for enrolling a voiceprint in a fraudster database, the system comprising: a) a memory device capable of storing a fraud model comprising at least one hypothesis indicative of a fraudulent transaction; b) a voice processing engine capable of processing audio data based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster; and c) an enrollment engine capable of enrolling the at least one suspect voiceprint in the fraudster database.

[0008] In yet another aspect of the present disclosure, the present disclosure provides computer-implemented methods, computer systems and a computer readable medium containing a computer program product for enrolling a voiceprint in a fraudster database, the computer program product comprising: a) program code for defining a fraud model comprising at least one hypothesis indicative of a fraudulent transaction; b) program code for processing audio data based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster; and c) program code for enrolling the at least one suspect voiceprint in the fraudster database.

BRIEF DESCRIPTION OF THE DRAWINGS

[0009] The accompanying drawings, where like reference numerals refer to identical or functionally similar elements throughout the separate views, together with the detailed description below, are incorporated in and form part of the specification, and serve to further illustrate embodiments of concepts that include the claimed disclosure, and explain various principles and advantages of those embodiments.

[0010] FIG. 1 shows a pictorial representation of a system used for early fraud detection and early enrollment of a fraud voiceprint into a fraudster database, in accordance with an embodiment of the present disclosure;

[0011] FIG. 2 shows a high level block diagram of an Early Fraud Detection system, in accordance with one embodiment of the present disclosure;

[0012] FIG. 3 shows a high level flowchart of a method to determine whether a suspect voiceprint belongs to a fraudster, in accordance with an embodiment of the present disclosure;

[0013] FIG. 4 shows a flowchart of a method to determine whether an individual from audio data called multiple times, in accordance with an embodiment of the present disclosure;

[0014] FIG. 5 shows a flowchart of a method to determine whether an individual from audio data called multiple times, in accordance with an embodiment of the present disclosure;

[0015] FIG. 6 shows a flowchart of a method for early fraud detection and enrolling a voiceprint into a fraudster database, in accordance with an embodiment of the present disclosure; and
FIG. 7 shows hardware to implement the method disclosed herein, in accordance with an embodiment of the present disclosure. The method and system have been represented where appropriate by conventional symbols in the drawings, showing only those specific details that are pertinent to understanding the embodiments of the present disclosure so as not to obscure the disclosure with details that will be readily apparent to those of ordinary skill in the art having the benefit of the description herein.

DETAILED DESCRIPTION OF EXEMPLARY EMBODIMENTS

In the following description, for purposes of explanation, numerous specific details are set forth in order to provide a thorough understanding of the disclosure. It will be apparent, however, to one skilled in the art, that the disclosure may be practiced without these specific details. In other instances, structures and devices are shown at block diagram form only in order to avoid obscuring the disclosure.

Reference in this specification to “one embodiment” or “an embodiment” means that a particular feature, structure, or characteristic described in connection with the embodiment is included in at least one embodiment of the disclosure. The appearances of the phrase “one embodiment” in various places in the specification are not necessarily all referring to the same embodiment, nor are separate or alternative embodiments mutually exclusive of other embodiments. Moreover, various features are described which may be exhibited by some embodiments and not by others. Similarly, various requirements are described which may be requirements for some embodiments but not other embodiments.

Broadly, embodiments of the present disclosure at least reduce losses due to fraudulent transactions, for example, credit card transactions by using voice data to identify fraudsters. Specifically, the method and system disclosed herein help modern enterprises such as merchants, banks, insurance companies, telecommunications companies, and payments companies by checking whether a caller is a fraudster by checking call audios. The detected audio is enrolled into a fraudster database which may then be used to stop future fraudulent transactions in the manner described in the specifications of U.S. Ser. No. 11/404,342, U.S. Ser. No. 11/754,974, U.S. Ser. No. 11/754,975, and U.S. Ser. No. 12/352,530, each of which is incorporated herein by reference. In case of a credit card fraud, as described in the patent applications incorporated herein by reference, frauds are first detected by the credit card companies and are then enrolled into the fraudster database. However, in the present disclosure, frauds are detected at the merchant’s end and are then enrolled into the fraudster database i.e. there is an early/advance enrollment of the frauds into the fraudster database. Specifically, a fraud is enrolled into the fraudster database at an early stage i.e. before receiving a fraud report from a credit card company. Therefore, the early/advance enrollment of frauds into the fraudster database may save merchants from getting duped by alerting them at an early stage. It is to be understood that although the present disclosure explains the method and the system in context of a credit card fraud, the method and system may be used in a similar manner for other types of frauds such as loan fraud, insurance fraud, all types of identity fraud, employment fraud, etc.

Referring to FIG. 1, a pictorial representation of a system used for early fraud detection and early enrollment of a fraud voiceprint into a fraudster database is shown, in accordance with an embodiment of the present disclosure. In one embodiment, a caller 2 may call a modern enterprise 4 using a suitable telephone network such as PSTN/Mobile/VOIP 6 for placing an order for goods or services. In one embodiment, a Private Branch Exchange (PBX) 8 may be used to receive the call. The PBX 8 may send the call audio to an audio recording device 10 which may record the call audio. In one embodiment, a call-center ‘X’ may receive and record the call on behalf of the modern enterprise 4, however, in another embodiment, the modern enterprise 4 may employ an agent (in-house or outsourced) or any other third party to receive and record the call.

The audio recording device 10 may be configured to transmit the recorded call to a transaction database 12, hereinafter database 12. The database 12 includes phone details of all possible calls received at the modern enterprise 4. In one embodiment, the phone details may include at least one of an audio conversation between the modern enterprise 4 and the caller, the amount of the transaction, type of goods or services ordered for (in case of a credit card transaction), a time of call, quantity of goods, reason for the call like placing an order or checking status, credit card credentials, a recipient of the goods, a place of shipment, billing address, caller identity such as a name and/or a social security number of the caller or agent ID (in case of an agent) or an employee ID number, a phone number from which the call is made, a phone number to which the call is made, and other transaction information. In one embodiment, the database 12 may include an audio database 14 and an order database 16. The audio database 14 is capable of storing call audios and the order database 16 is capable of storing order details.

The modern enterprise 4 may also include a fraudster database 18. The fraudster database 18 includes voice signatures or voice prints of known fraudsters. Essentially, a voice signature or voice print includes a set of voice characteristics that uniquely identify a person’s voice. In one embodiment, each voice signature in the fraudster database 18 is assigned a unique identifier (ID), which in accordance with one embodiment may include one or more incident details such as a social security number used, a name used, credit card credentials used, date and time of fraud, amount of the fraud, a type of fraud, enterprise impacted, and other details associated with the fraud incident.

In one embodiment, the phone details of certain callers may be transmitted to an Early Fraud Detection (EFD) system 20 via a file transfer server 22 using Internet/LAN 24. The EFD system 20 helps in early enrollment of a suspect voiceprint into the fraudster database 18. Specifically, the EFD system 20 processes the phone details of the callers to identify suspect voice prints present in the database 12 to enroll certain suspect voiceprints into the fraudster database 18. The suspect voice prints are enrolled into the fraudster database 18 which may then be used to stop future fraudulent transactions in the manner described in the specifications of U.S. Ser. No. 11/404,342, U.S. Ser. No. 11/754,974, U.S. Ser. No. 11/754,975, and U.S. Ser. No. 12/352,530, each of which is incorporated herein by reference. More details about the working of the EFD system 20 will be described later in this section. In one embodiment, the EFD system 20 may be a distributed system that includes components that are not all located at a single location, but instead are distributed over multiple locations. The EFD system 20 may include software to facilitate communications with the modern enterprise 4 or
the call-center ‘X’ to access the database 12. In one embodiment, the software may include a browser which is an application that facilitates communications via the Internet with the modern enterprise 4 or the call-center ‘X’ using networking protocols such as for example the Hypertext Transfer Protocol (HTTP)/the Internet Protocol (IP), the Simple Object Access Protocol (SOAP), etc. In another embodiment, the EFD system 20 may be integrated in the modern enterprise 4, thereby alleviating the need of transferring the phone details of callers.

[0025] Referring now to FIG. 2, an internal block diagram of the EFD system 20 is shown, in accordance with an embodiment of the present disclosure. The EFD system 20 includes a selection engine 26, a memory device 28, a voice processing engine 30, and an enrollment engine 32. Each of the components 26 to 32 may be implemented in hardware or software as a combination of both hardware and software. Further, it is to be understood that while the components 26 to 32 are shown as separate components based on function, some or all the components may be integrated.

[0026] In one embodiment, the EFD system 20 may use the selection engine 26 to select or choose certain audio data from the database 12. The audio data may include phone details of some callers. Selecting audio data from the database 12 is important because the database 12 includes a high volume of data and will consume a lot of time if analyzed completely. Therefore, to save time and computations, a part of the database 12 is selected for analysis. This also helps in maintaining a good speed of the EFD system 20. In one embodiment, the selection engine 26 may stream audio data via the file transfer server 22 in real time or after the call or on demand using Internet/LAN 24. Alternatively, the selection engine 26 may request for audio data in a batch using transmitting techniques such as Email/Voice Over Internet Protocol (VOIP)/HTTP/Service Control Point (SCCP)/Session Initiation Protocol (SIP) via the Internet/LAN 24.

[0027] In one embodiment, the audio data may be selected based on selection criteria. The selection criteria may help in filtering the database 12 to select the audio data in a way such that a maximum probability of identifying a fraudster is achieved. The selection criteria may be defined considering a pattern which may be followed by fraudsters. In other words, the selection criteria may help in segregating high risk audio data from the database 12. For example, in one embodiment, the audio data of last two weeks may be selected for analysis. This may be done because when a fraudster hacks certain credit cards, they generally have less time to do as many transactions as possible before the credit card stops working. Therefore, selecting the audio data of the last two weeks may be helpful to identify a fraudster. Further, the selection criteria may also include multiple calls made from a same phone number. This may be helpful to identify a fraudster because a fraudster may use a same phone line to place orders.

[0028] Other selection criteria may include multiple calls made from a specific phone number pattern. This may be helpful because the fraudster can use different public telephones in the same town to place orders. Specific phone number pattern may belong to a specific geographic location. The selection criteria may also include specific out-of-pattern orders such as an order for five LCD TVs within two days from a specific phone number pattern. The selection criteria may also include transaction profile similarities, for example all orders for a specific number and/or type of product, all wire transfers made before 9:00 A.M., similar transactions including transactions related to a specific geographic location and transactions related to a specific action such as change of address on an account, all calls for same account, random sampling, all orders that exceed a certain dollar value, all calls that have been call forwarded (all calls with similar telephony characteristics), all calls from a specific service provider (Skype or VOIP, landline, or mobile), calls for which an expected channel (landline, wireless, or VOIP) that phone number is associated with (based on phone data records) is different from a channel detected (e.g. by channel detection determined by analyzing audio signal characteristics), all calls from Automatic Number Identifications (ANIs) frequently used, etc. Therefore, based on these selection criteria, the audio data may be selected for the purpose of identifying a fraudster in the audio data.

[0029] The audio data may be processed using the voice processing engine 30. Specifically, the voice processing engine 30 processes the audio data to uniquely identify individuals who called multiple times. More specifically, the voice processing engine 30 uses voice biometrics to identify individuals in the audio data who called multiple times. With the help of voice biometrics, the voice processing engine 30 scans each voice sample in the audio data to identify individuals who called multiple times. Subsequently, the voice processing engine 30 determines whether the individuals who called multiple times used multiple caller identities. This may be done by checking the phone details which include caller identity information of all the callers. A person who called multiple times using multiple caller identities may be a fraudster. In one embodiment, when a fraudster uses a certain number of credit cards, he may also use the identity of the actual credit holder to place an order for goods or services over the phone. When the fraudster uses more than one credit card, he may use more than one caller identity. Therefore, a person who called multiple times using multiple caller identities may be a fraudster. In another embodiment, when an employee, who was fired by a merchant at one location, may use a different identity to get a job with the same merchant at a different location. Therefore, the employee who called multiple times using multiple caller identities may be a fraudster.

[0030] Based on these types of frauds, a fraud model may be defined. Specifically, the fraud model may be based on a fraudster who called multiple times using multiple identities. In one embodiment, the fraud model may be stored in the memory device 28. Further, the fraud model may include at least one hypothesis indicative of a fraudulent transaction. In the present embodiment, the hypothesis includes a single voiceprint being matched to multiple caller identities. The hypothesis may be manually coded into the memory device 28.

[0031] In the present embodiment, the voice processing engine 30 processes the audio data based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster. More specifically, the voice processing engine 30 checks the audio data to determine whether the individuals who called multiple times used multiple caller identities. In other words, the voice processing engine 30 checks the phone details of individuals who called multiple times to determine whether the hypothesis is true or not. If the hypothesis is true, i.e. if an individual is found to have called multiple times using multiple caller identities, the enrollment engine 32 enrolls the suspect voiceprint into the fraudster database 18. Otherwise, a next set of audios in the audio data is checked. In one embodiment, the EFD system 20
may periodically check the database 12 to proactively enroll a suspect voiceprint into the fraudster database 18 prior to receiving a fraud report from a credit card company or from the merchant. In another embodiment, the EFD system 20 may perform such a check on demand by a modern enterprise. Therefore, with each such check, the EFD system 20 may shrink the window between the fraud event and the modern enterprise realization that a fraud has occurred.

[0032] Referring to FIG. 3, a high level flowchart of a method to determine whether a suspect voiceprint belongs to a fraudster is shown, in accordance with an embodiment of the present disclosure. At 300, a set of audios of the audio data having suspect voiceprints is processed. At 302, individuals who called multiple times are identified. At 304, it is determined whether the individuals who called multiple times used multiple caller identities. If yes, then the individual is declared a fraudster at 306 and the suspect voiceprint is enrolled into the fraudster database 18. Otherwise, a next set of audios of the audio data is processed at 300.

[0033] Referring to FIG. 4, a flowchart of a method to determine whether an individual from the audio data called multiple times is shown, in accordance with an embodiment of the present disclosure. In one embodiment, to determine this, let us assume that the number of audios in the audio data are \( N \) i.e. from \( a_1 \) to \( a_N \) and the number of callers who called multiple times are \( Y \) i.e. from \( F_1 \) to \( F_Y \), where \( Y \leq N \). In the flowchart, at 400, a first person is chosen. At 402, an unassigned audio \( a_i \) is picked up. At 404, a voice print model \( F_{\text{v}} \) is generated for \( a_i \). At 406, all the remaining audios of the set \( a_i \) to \( a_N \) are screened against the voice print \( F_{\text{v}} \) and it is determined which audios from the \( a_i \) to \( a_N \) match voice print \( F_{\text{v}} \). At 408, all the audios from the set of \( a_i \) to \( a_N \) that match the voiceprint \( F_{\text{v}} \) are marked. At 410, it is determined whether all audios i.e. \( a_i \) are assigned. If no, then a next person is chosen to run the flowchart on. If yes, then at 412 it is determined that \( Y=\phi \). This means that the corresponding individuals called multiple times.

[0034] Referring to FIG. 5, a flowchart of a method to determine whether an individual from the audio data called multiple times is shown, in accordance with an embodiment of the present disclosure. In one embodiment, to determine this, let us assume that the number of audios in the audio data are \( N \) i.e. from \( a_1 \) to \( a_N \) and the number of callers who called multiple times are \( Y \) i.e. from \( F_1 \) to \( F_Y \), where \( Y \leq N \). In the flowchart, at 500, temporary voice prints \( V_{\text{p}} \) are created for each an (1 to \( N \)). At 502, each \( a_i \) is screened against \( V_{\text{p}} \) to generate \( N*(N-1) \) scores representing likelihood that an and \( V_{\text{p}} \) are same individual. At 504, similar audios are grouped resulting in \( Y \) groups, each group representing a unique individual. At 506, a voiceprint \( F_{\text{v}} \) is generated for each group. At 508, \( Y \) individuals are identified.

[0035] Referring now to FIG. 6, a flowchart of a method for early fraud detection and enrolling a voiceprint into a fraudster database 18 is shown, in accordance with an embodiment of the present disclosure. At 600, audio data may be selected based on selection criteria. Further, at 602, a fraud model is defined. The fraud model includes at least one hypothesis indicative of a fraudulent transaction. In one embodiment, the hypothesis indicative of the fraudulent transaction includes a single voiceprint being matched to multiple caller identities. At 604, the audio data is processed based on the fraud model to identify at least one suspect voiceprint in the audio data suspected of belonging to a fraudster. The suspect voiceprint is considered identified when the hypothesis is true. At 606, it is determined whether the hypothesis is true or not. If the hypothesis is true, then at 608, the suspect voiceprint is enrolled into a fraudster database 18. However, if the hypothesis is not true, then at 610, it is assumed that there is no fraudster in the audio data.

[0036] Referring now FIG. 7, system 70 to implement the method disclosed herein is shown, in accordance with an embodiment of the present disclosure. The EFD system 20 and the server system have, thus far, been described in terms of their respective functions. By way of example, each of the EFD system 20 and server system of the present disclosure may be implemented using the system 70 of FIG. 7. The system 70 typically includes at least one processor 72 coupled to a memory 74. The processor 72 may represent one or more processors (e.g., microprocessors), and the memory 74 may represent random access memory (RAM) devices comprising a main storage of the system 70, as well as any supplemental levels of memory e.g., cache memories, non-volatile or backup memories (e.g. programmable or flash memories), read-only memories, etc. In addition, the memory 74 may be considered to include memory storage physically located elsewhere in the system 70, e.g. any cache memory in the processor 72, as well as any storage capacity used as a virtual memory, e.g., as stored on a mass storage device 80.

[0037] The system 70 also typically receives a number of inputs and outputs for communicating information externally. For interface with a user or operator, the system 70 may include one or more user input devices 76 (e.g.: a keyboard, a mouse, etc.) and a display 78 (e.g., a Liquid Crystal Display (LCD) panel).

[0038] For additional storage, the system 70 may also include one or more mass storage devices 80, e.g., a floppy or other removable disk drive, a hard disk drive, a Direct Access Storage Device (DASD), an optical drive (e.g. a Compact Disk (CD) drive, a Digital Versatile Disk (DVD) drive, etc.) and/or a tape drive, among others. Furthermore, the system 70 may include an interface with one or more networks 82 (e.g., a local area network (LAN), a wide area network (WAN), a wireless network, and/or the Internet among others) to permit the communication of information with other computers coupled to the networks. It should be appreciated that the system 70 typically includes suitable analog and/or digital interfaces between the processor 72 and each of the components 74, 76, 78 and 82 as is well known in the art.

[0039] The system 70 operates under the control of an operating system 84, and executes various computer software applications, components, programs, objects, modules, etc. to perform the respective functions of the EFD system 20 and server system of the present disclosure. Moreover, various applications, components, programs, objects, etc. may also execute on one or more processors in another computer coupled to the system 70 via a network 82, e.g. in a distributed computing environment, whereby the processing required to implement the functions of a computer program may be allocated to multiple computers over a network.

[0040] In general, the routines executed to implement the embodiments of the present disclosure, may be implemented as part of an operating system or a specific applications component, program, object, module or sequence of instructions referred to as "computer programs." The computer programs typically comprise one or more instructions set at various times in various memory and storage devices in a computer, and that, when read and executed by one or more processors in a computer, cause the computer to perform operations
necessary to execute elements involving the various aspects of the present disclosure. Moreover, while the disclosure has been described in the context of fully functioning computers and computer systems, those skilled in the art will appreciate that the various embodiments of the present disclosure are capable of being distributed as a program product in a variety of forms, and that the present disclosure applies equally regardless of the particular type of machine or computer-readable media used to actually effect the distribution. Examples of computer-readable media include but are not limited to recordable type media such as volatile and non-volatile memory devices, floppy and other removable disks, hard disk drives, optical disks (e.g., Compact Disk Read-Only Memory (CD ROMS), Digital Versatile Disks, (DVDs), etc.), among others, and transmission type media such as digital and analog communication links.

One advantage of the techniques and systems described herein is that fraud detection is based on a fraudster's voice, which being biometric in nature is linked to the fraudster. Further, the method and system enables a person to detect fraud much before a confirmed fraud report is issued by a credit card company thereby minimizing the amount of damage that could be caused.

21. A method for enrolling a voiceprint into a fraudster database, the method comprising:
   storing a fraud model in a memory device;
   receiving at a fraud detection engine a batch of transactions before detection of a fraud for transactions in the batch, each transaction including audio data and a caller identity;
   processing the batch of transactions to extract a subset of the batch of transactions based on the fraud model, the processing performed using a selection engine;
   processing the audio data of each of the extracted transactions using a voice processing engine to identify a group of transactions in the extracted transactions, the identified group of transactions including audio data characteristic of a unique individual who has called multiple times;
   processing the caller identities of the identified group of transactions using the voice processing engine to determine that the unique individual has used multiple caller identities;
   generating a voiceprint from the audio data of the identified group of transactions using the voice processing engine;
   and
   enrolling the generated voiceprint into the fraudster database using an enrollment engine.

22. The method according to claim 21, wherein the fraud model comprises transactions related to a phone number pattern belonging to a specific geographical location.

23. The method according to claim 21, wherein the fraud model comprises transactions related to a specific action requested by the caller.

24. The method according to claim 23, wherein the specific action is change of address.

25. The method according to claim 21, wherein processing the audio data of each of the extracted transactions comprises generating a score for each transaction and comparing scores to determine a likelihood that the two callers are the same individual.

26. The method according to claim 21, wherein fraud model comprises at least one of a predetermined time period, random sampling, calls made to a same phone number, calls received from a same phone number, calls made to a specific phone number pattern, all calls for a same account, calls received from a specific phone number pattern, call forwarded calls, calls from a specific service provider, similar transactions, and specific out-of-pattern transactions.

27. The method according to claim 21, wherein the fraud model comprises transactions related to a specific geographical location.

28. The method according to claim 27, wherein the geographical location comprises at least one of a billing address, a place of shipment, a zip code, and an address on an account.

29. A system for early detection of fraudsters, the system comprising:
   a device memory for storing a fraud model;
   an early detection engine configured to receive a batch of transactions before the system receives a fraud report for transactions in the batch, each transaction including caller audio data and caller identity data;
   a selection engine configured to extract a subset of transactions from the batch of transactions based on the fraud model;
   a voice processing engine configured to process the caller audio data and caller identity data of the subset of transactions extracted from the batch to:
   identify a group of transactions including audio data characteristic of a unique individual who has called multiple times using more than one caller identities, and
   generate a voiceprint from the audio data of the identified group of transactions; and
   an enrollment engine configured to enroll the generated voiceprint into the fraudster database.

30. The system according to claim 29, wherein the caller audio data includes a voiceprint.

31. The system according to claim 29, wherein the batch of caller audio data is recorded by a third party on behalf of an enterprise.

32. The system according to claim 29, wherein the fraud model includes transactions that exceed a predetermined value.

33. The system according to claim 29, wherein the fraud model includes comparing an expected channel associated with a phone number and a detected channel.

34. The system according to claim 33, wherein the voice processing engine is further configured to analyze audio signal characteristics to determine the detected channel.

35. The system according to claim 29, wherein the fraud model includes calls from frequently used Automatic Number Identifications.

36. The system according to claim 29, wherein the fraud model includes determining if a call is received from a VOIP, landline, or mobile service provider.

37. The system according to claim 29, wherein the fraud model includes calls received from the same phone number.

38. A non-transitory computer readable storage media having a program embodied thereon, the program being executable by a processor to perform a method for enrolling voiceprints into a database of suspected fraudsters, the method comprising:
   analyzing caller audio data in a batch of transactions related to calls to identify a plurality of groups of multiple calls, each group characteristic of a unique individual;
determining from caller identity data for each of the identified groups of multiple calls if the transactions related to the group of multiple calls include multiple caller identities; and
enrolling a voiceprint of the unique individual into the database of suspected fraudsters.

39. The system according to claim 38, wherein the analyzing caller audio data is performed before receiving a fraud report for the batch of transactions.

40. The system according to claim 38, further comprising extracting a subset of transactions from the batch of transactions by analyzing audio signal characteristics of the call audio data for channel detection.