

Disease Identification System: System Implementation and Attack

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Abstract: *Interactions with outside world are not only natural but mandatory, in DIS user can communicate to doctors through the system for retrieving information about diseases by giving symptoms as input. The system is designed specifically for users who are patients and diagnostic clinics. DIS provides a collaboration and communication platform for first doctors, extension diagnosticians to share information on human body related diseases. The system uses patient data and AI based database for medicines to respective diseases as tools for enhancement of diagnosis of diseases. Through interactions on the Internet between extension doctors and specialists, problems can be communicated immediately and assessed. Specialists authorized to the system can perform diagnosis and identification and provide the best remedy recommendations to the users. Also, an algorithm independent approach for mounting poisoning attack across a dataset of DIS system is being implemented which in turn may lead to false diagnostics.*

Keywords: Poisoning Attack, machine learning, NLP, DIS, AI

1. Introduction

Much of our work on language, speech, translation, and visual processing relies on machine learning and AI. In all of those tasks and many others, we gather large volumes of direct and indirect evidence of relationship of interest, and we apply learning algorithm to generalize from that evidence to new cases of interest. Machine learning algorithms are being used in critical applications where they drive decisions with large personal, organizational, or societal impact. These applications include healthcare, network intrusion detection systems (IDSs), spam and fraud detection, phishing detection, political decision making, and financial engineering.

The branch of computer science concerned with making computers behave like humans means interaction between computers and human languages. The process of a computer extracting meaningful information from natural language input and producing natural language output. There are many sites which provide the facility of AI in computer. And ALICE is highly recommended chat Robot. Currently the most famous and intuitive chat service is provided by them. We intend to make such AI application that it is useful for patients. User can communicate to the system for retrieving information about diseases by giving symptoms as input. Also medical related information will be provided by our bot.

Two main categories of security attacks on machine learning have been considered in the literature: exploratory and causative. Exploratory attacks exploit existing vulnerabilities without altering the training process. On the other hand, causative attacks alter the training process, typically by modifying the training dataset. Poisoning attacks are a class of causative attacks in which carefully-crafted malicious instances are added to the training dataset, leaving the rest of the dataset intact.

New results indicate that it may be easier than we thought to provide data to a learning program that causes it to learn the wrong things. The key idea is that, given machine

intelligence, the trick to defeating it is to feed it the wrong data. Security experts call the idea of breaking a system by feeding it the wrong data a poison attack. The approach also provides the countermeasure for measuring the defending power of the application dataset against the poisoning attack.

In the healthcare system, hindrance of a diagnosis may have life threatening consequences and could cause distrust. On the other hand, not only may a false diagnosis prompt users to distrust the machine learning algorithm and even abandon the entire system but also such a false positive classification may cause patient distress.

It could be possible to direct the induced errors so as to product particular types of error. For example, a spammer could send some poisoned data so as to evade detection in the future. The biggest practical difficult in using such methods is that, in most cases, the attacker doesn't control the labeling of the data points - i.e. spam or not spam - used in the training. A custom solution would have to be designed to compromise the labeling algorithm.

These attacks may be applied to both fixed and evolving datasets. They can be applied even when only statistics of the training dataset are available or, in some cases, even without access to the training dataset. Finally, we present countermeasures against the proposed generic attacks that are based on tracking and detecting deviations in various accuracy metrics, and benchmark their effectiveness.

2. Previous Work

Depending upon the research based on this system we have following topics available, which state methodologies for implementing system and attacking and securing the attack on machine learning algorithm used for creating medical dataset. These applications are based on Artificial Intelligence.

2.1 Language Translator

This is one of the most important applications of Natural Language Processing. Translation of a sentence from one language to another, retaining the meaning, is a difficult task. A lot of research has been done on this now in different parts of the world.

2.2 Existor: System named Evie

There are no limits. Artificial Intelligence is communication. Natural language is universal. Evie was created by Existor. Evie is an Electronic Virtual Interactive Entity. This bot is capable of communicating with human beings in very normal way and also on general topics. Also it had now learned to understand smiley's and meaning of other symbols. In early September 2011, Existor's system Cleverbot, recently took the Turing Test at the Techniche computer festival in Guwahati - and was judged to be 59% human! The humans were only judged to be 63% human, so Cleverbot has arguably passed the test.

2.3 Dr. Romulon System

This system is based on the ALICE artificial intelligence chat platform. Unlike the Eliza system, this bot has a larger set of responses. But the Doctor does have a very poor memory. Also this system is unable to learn from experiences and also from past conversations. **2.4**

Eliza : System (Psychotherapist) Eliza is one of the oldest chat programs around. She is the classic chat bot, born 46 years ago in 1966. Eliza is a conversation simulation. But more so, she emulates a psychotherapist in the way she asks and answers your statements. Also this system is unable to learn from experiences and also from past conversations.

2.5 A Basic Program for Clinical Problem-Solving

Doctors use a combination of a patient's case history and current symptoms to reach a health diagnosis when a patient is ill. In order to recognize the combination of symptoms and history that points to a particular disease, the doctor's brain accesses memory of previous patients, as well as information that has been learned from books or other doctors. A neural network has the ability to mimic this type of decision-making process, and use a knowledge base of information, and a training set of practice cases, to learn to diagnose diseases.

2.6 Poisoning Attack

Systematic attack schemes for mounting poisoning attacks against machine learning algorithms used for medical datasets, and suggested countermeasures against them. A key feature of the proposed attack schemes is that they can be applied to a wide range of machine learning algorithms, even when the machine learning algorithm is unknown. Author evaluated the effectiveness of the attacks against six machine learning algorithms and five datasets (Thyroid Disease, Breast Cancer, Acute Inflammations, Echocardiogram, and Molecular Biology (Splice-junction Gene Sequences)), and ranked the algorithms based on their ability to withstand the

attacks. They then presented countermeasures against these attacks and evaluated their effectiveness. Finally, we identified the machine learning algorithms that are easiest to defend [6].

3. Proposed Work

In healthcare, attackers may have varying motivations for poisoning training datasets, ranging from generally degrading the accuracy of the algorithm to biasing the results in a specific, targeted manner. As an example of targeted attacks, let us consider the Thyroid Disease dataset, in which data instances are associated with two classes: normal and hypothyroid. Targeted attacks might compromise the effectiveness of the machine learning algorithm either to prevent a hypothyroid diagnosis or to falsely lead to a hypothyroid diagnosis.

The application aims at providing medical assistance. The application will take as input text i.e. the symptoms and process it using the database. There are various parts of the project. The first part will be completing the basic requirements of the system. It will take input and search in the database for the corresponding output and gives the result. For this search we will use various search-engines like special search engine, matrix engine which stores the basic conversation [3].

The second part of the application will involve helping the system to learn. It means there maybe some symptoms which the system doesn't know and if the patient mentions it then the system stores it in the learning database. The system also uses engines like previous conversation engine, auto-talk engine

The third part of the application will involve applying poisoning attack on the dataset of system and checking whether that dataset can be affected or not. Also the countermeasure algorithm will be applied to see till what extend the system dataset can defend the applied poisoning attack. The proposed attack schemes can be used for targeted or non-targeted attacks. However, we describe the procedure in the context of targeted attacks [6].

4. Project Preparation

System to be developed is a system that is capable of carrying on conversation with humans in natural language that is English. The main aim of the project is to create a system related to AI in medicines. Basically this application will be providing facility to the patients to have a conversation with a system and ask him the diseases by providing symptoms to the bot. Many chat bots are available but they don't provide such facility for hospital applications. So we had thought of developing this application [1][2].

It is important to investigate whether machine learning algorithms used for healthcare applications are vulnerable to security and privacy threats. The robustness of machine learning algorithms to noise in the training data has also been investigated to evaluate its effects on the decision making

process. In healthcare applications, poisoning attacks are highly relevant because although manipulation of existing data in the training dataset may be difficult or impossible for attackers, addition of new data might be relatively easy. For instance, hindrance of a hypothyroid diagnosis may have life-threatening consequences due to delayed treatment. This may reduce trust in the machine learning algorithm.

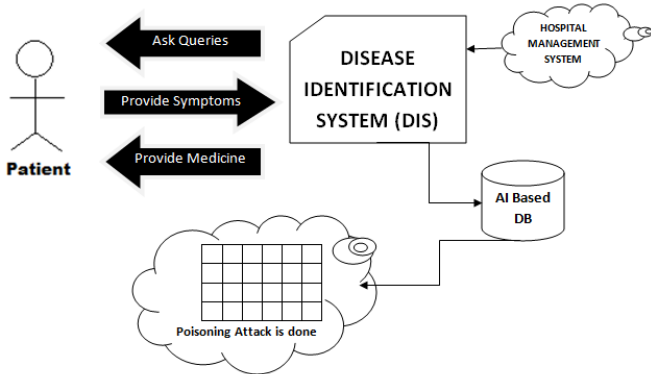


Figure 1: Disease Identification System with poisoning attack

On the other hand, a false positive classification may cause unnecessary concern. If poisoning attacks are detected, the user or owner of the dataset may take appropriate action, such as disregarding the results of machine learning or attempting to cleanse the dataset of the malicious data. From an attacker’s perspective, it is therefore desirable to mount poisoning attacks such that they are difficult to detect. On the other hand, if such attacks are successful, the resulting erroneous conclusions may lead to serious adverse impact on people, institutions, and healthcare services.^[12]

To evaluate the proposed scheme, we will experiment with the system dataset and show that the proposed attack is successful or not. Moreover, we show that one may obtain a surrogate dataset to mount the attacks, eliminating the need for access to the training dataset.

5. Attack Scheme and DIS System Scheme

Our objective is to propose a generic and algorithm-independent attack scheme. In other words, the proposed attacks can be applied to a wide range of machine learning algorithms and medical datasets.^[12] In fact, the attacker does not even need to know the type of machine learning algorithm used to apply the proposed attack scheme. Furthermore, highly algorithm-specific attacks may be thwarted by simply changing the machine learning algorithm used. However, knowledge of the machine learning algorithm being used increases the efficacy of the attacks, as discussed later. In this attack model, we assume that the attackers have knowledge of the training dataset and use this knowledge to construct malicious data. In practice, this knowledge can be obtained either because the dataset is publicly available, or because the attackers have employed various means, such as eavesdropping on network traffic or compromising a system where the dataset is stored, in case security measures, such as the ones presented in, are compromised. However, the

success of the proposed attacks is only dependent on the knowledge of the statistics of the training dataset [6].

In scenarios where gaining access to the training datasets is difficult, we present an alternative approach in which attackers construct a proxy training dataset drawn from the same distribution as the original dataset. This is possible since our proposed attacks are based on the statistics of the training dataset (and not the exact values of attributes within the dataset). By presenting artificial test instances as inputs to the targeted machine learning application and observing its responses, one can construct a “proxy” dataset that can be used to mount the attack [6]. Moreover, in many cases, launching poisoning attacks may be much easier than launching general causative attacks in which modifications to current instances are required.

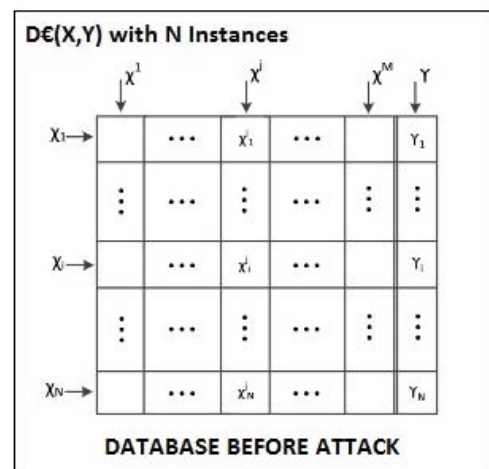


Figure 2: Database before poisoning attack

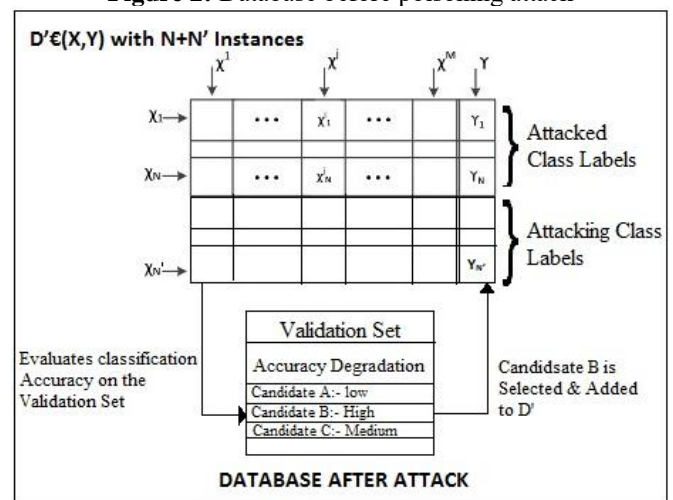


Figure 3: Database after poisoning attack

Algorithm 1:- Algorithm-independent attacks.

Input: Dataset $D \in (x, Y)$ with N instances, validation dataset V , number of iterations I .

Output: Maliciously manipulated dataset $D' \in (x, Y)$ with $N + N'$ instances, where N' is the number of added malicious instances.

Begin

- Assign $D' \leftarrow D$

- for $k = 1$ to N' do

- //Select k^{th} malicious instance

- for $i = 1$ to I do
- Use Algorithm 2 to generate malicious instance candidate i
- Add the candidate to D' to create a temporary training set $D_T \in (x, Y)$ with $N + k$ instances
- Build the model using D_T and record its classification accuracy on the validation set V as A_i
- endfor
- Select instance i^* such that $A_{i^*} = \min(A_i), 1 \leq i \leq I$
- Add instance i^* to D'
- end for

End
Return: $D' \in (x, Y)$.

Algorithm 2: Deriving a malicious instance candidate.
Input: $x^j, 1 \leq j \leq M$ and $Y^j, 1 \leq j \leq N$, g bins (a specified constant).
Output: $x_{N+1}; Y_{N+1}$ (malicious instance candidate).
 Denote: $n_{k,j}$ and $n'_{k,j}$ as the number of entries in $x^j(k)$ corresponding to the attacked class and the attacking class respectively.

Begin

- for $j = 1$ to M do
- for $k = 1$ to g do
- Calculate $n_{k,j}$ and $n'_{k,j}$
- Assign $W_k \leftarrow n_{k,j} / n'_{k,j}$
- endfor
- Compute attribute probabilities ($P_k = W_k / \sum_{1 < l < g} W_l$), $k = 1$ to g
- Weighted function S selects attribute value α_j pseudo randomly based on attribute probabilities
- endfor

End
 Return: Malicious instance candidate is $x_{N+1} = \{\alpha_j, 1 \leq j \leq M\}$, $Y_{N+1} =$ Attacking class.

The proposed attack scheme is described in Algorithm 1. Let the original dataset be denoted as $D \in (x, Y)$ with N instances, where x and Y represent an instance's attributes and class label, respectively. Algorithm 1 adds N' malicious instances to the original dataset to create a manipulated dataset $D' \in (x, Y)$ with $N + N'$ instances. To add a malicious instance, I pseudorandom candidates are generated (using Algorithm 2). Intuitively, Algorithm 2 generates candidates whose attribute values match the statistics of the attacked class, but whose labels are set to the attacking class (recall that the objective is to bias the model away from the attacked class and towards the attacking class). For each candidate, the algorithm builds a model on the same machine learning algorithm and evaluates its classification accuracy on the validation set. The candidate that results in the highest degradation in classification accuracy is selected and added to the dataset. I is a constant set by the attacker and trades off efficacy of the malicious instances vs. the computational effort expended to generate them.

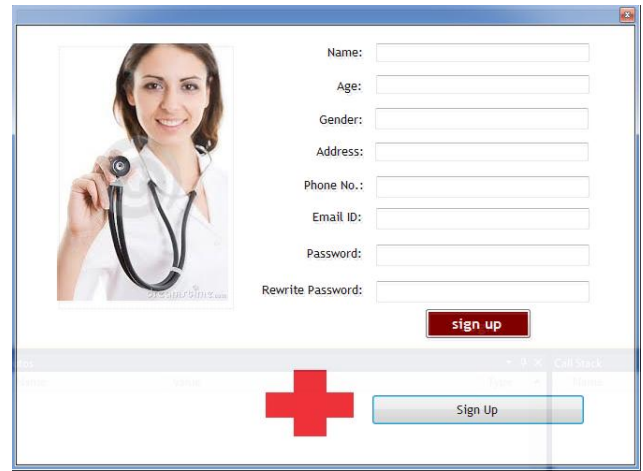


Figure 4: login page for user as well as doctor

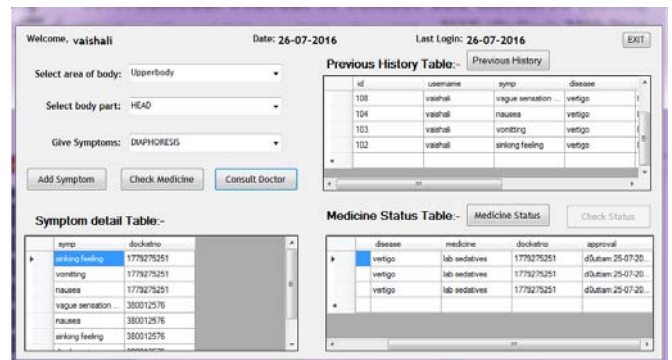


Figure 5: DIS with patient details

Figure 4 and figure 5 indicates the login page for user which is patient. Once patient logs in the DIS system his previous history can be viewed. Also patient can provide the new symptoms if any and get the name of disease and medicine recommended by registered and authorized doctor.

Doctor can log in through his account and view the patients name as well as symptoms he provided. Based on that information doctor can recommend some medicine to patient which is present in the database. If doctor comes across the new combination of symptoms he may also recommend the new medicine to the patient. Database will use the algorithm for self learning and update itself with symptom combinations as well as medicines.

id	username	symp	disease	medicine	docterno	approval
102	vaishali	sinking feeling	vertigo	lab sedatives	1779275251	dOuttam:25-07-2016
103	vaishali	vomitting	vertigo	lab sedatives	1779275251	dOuttam:25-07-2016
104	vaishali	nausea	vertigo	lab sedatives	1779275251	dOuttam:25-07-2016
105	prasanna	sinking feeling	vertigo	sedatives	680509314	dOuttam:26-07-2016
106	prasanna	vomitting	vertigo	sedatives	680509314	dOuttam:26-07-2016
107	prasanna	nausea	vertigo	sedatives	680509314	dOuttam:26-07-2016
108	vaishali	vague sensation of loss of confidence	vertigo	labyrinthine sedatives	380012576	xxxx
109	vaishali	nausea	vertigo	labyrinthine sedatives	380012576	xxxx
110	vaishali	sinking feeling	vertigo	labyrinthine sedatives	380012576	xxxx
111	vaishali	diphoresis	vertigo	labyrinthine sedatives	380012576	xxxx
112	prakash	depressed mood	psychogenic headache	abcd	1641494554	dDapama:26-07-2016
113	prakash	anxiety headache	psychogenic headache	abcd	1641494554	dDapama:26-07-2016
114	prakash	vomitting	psychogenic headache	abcd	1641494554	dDapama:26-07-2016

Figure 6: Database

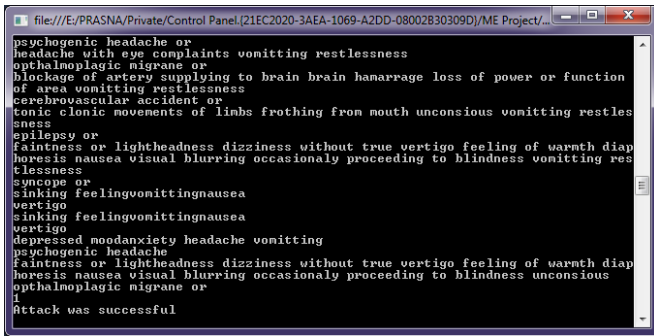


Figure 7: Implementation of Attack

Snapshot 7 shown above depicts the implementation of poisoning attack on database. Due to this poisoning attack self learning database will get poisoned with fake symptom combination and medicine and disease as well. This is a slow poisoning attack which if done several times may impact the system harshly resulting in false diagnostics of patients. Hence this is a threat to system as well as patient's life too.

We would like to conclude this paper with the system medical field related system useful for patients and also user friendly for them. Patients can diagnose themselves with the use of this system. systematic attack schemes for mounting poisoning attacks against DIS system used for medical datasets is also shown. A key feature of the proposed attack schemes is that they can be applied to a wide range of machine learning algorithms, even when the machine learning algorithm is unknown.

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