SMART HEALTH PREDICTION FOR AVOIDING FUTURE HEALTH RISK BY USING MACHINE LEARNING

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Abstract:

Artificial intelligence (AI) aims to mimic human cognitive functions. It is bringing a paradigm shift to healthcare, powered by increasing availability of healthcare data and rapid progress of analytics techniques. We survey the current status of AI applications in healthcare and discuss its future. AI can be applied to various types of healthcare data (structured and unstructured). Popular AI techniques include machine learning methods for structured data, such as the classical support vector machine and neural network, and the modern deep learning, as well as natural language processing for unstructured data. Major disease areas that use AI tools include cancer, neurology and cardiology. We then review in more details the AI applications, in the major areas of early detection and diagnosis, as well as outcome prediction and prognosis evaluation.

Keyword – Ai, Machine learning, Health care.

I. INTRODUCTION:

Numerous social insurance associations (doctor's facilities, medicinal focuses) in China are occupied in serving individuals with best-exertion social insurance benefit. These days, individuals give careful consideration on their physical conditions. They need higher quality and more customized medicinal services benefit. In any case, with the restriction of number of talented specialists and doctors, most medicinal services associations can't address the issue of open. Step by step instructions to give higher quality social insurance to more individuals with restricted labor turns into a key issue. The social insurance condition is for the most part seen as being 'data rich' yet 'information poor'. Doctor's facility data frameworks normally produce enormous measure of information which appears as numbers, content, outlines and pictures. There is a parcel of concealed data in this information immaculate. Information mining what's more, prescient investigation expects to uncover examples and principles by applying propelled information examination systems on a substantial arrangement of information for expressive and prescient purposes. Information mining is appropriate for preparing vast datasets from healing facility data framework and discovering relations among

information highlights. It takes just a couple of scientists to examine information from doctor's facility data frameworks, and give enormous medicinal learning which can be used to help clinical basic leadership. Likewise, we could utilize information mining to give a self-benefit human services framework, which can serve bunches of individuals in the meantime. Oneself administration human services framework is of incredible hugeness to take care of the issue of unevenness between constrained therapeutic assets and requests.

II. LITERATURE SURVEY:

1) Author Name- Neill DB, 2013 (Using artificial intelligence to improve hospital inpatient care. IEEE Intell System 2013;

Description-Electronic health records have become more available due to the guidelines of the Health

Information Technology for Economic and Clinical Health (HITECH) Act, which offers incentives to healthcare providers to adopt EHR to advance clinical processes and improve outcomes. Meanwhile, health insurance providers and non profits such as the Health Care Cost Institute have committed to providing health insurance claims data with the goal of reducing costs while improving the quality and availability of coverage. Such sources provide detailed, time-stamped, and Highly multivariate data for a large patient population, enabling the use of AI techniques to connect care practices and outcomes.

Another new approach that might improve patient care focuses on statistical machine learning methods for

Detecting anomalous patterns in massive quantities of healthcare data. We recently developed a variety of machine learning methods based on *fast subset scanning* to detect patterns in massive datasets, efficiently identifying

Subsets of data records and attributes that are collectively anomalous or that maximize some measure of interest, such as a likelihood ratio statistic. In the patient care setting, our primary focus is to detect anomalous patterns of care that influence patient outcomes. 2) Author Name- Kolker E, Özdemir V, Kolker E (How Healthcare can refocus on its Super-Customers (Patients, n=1) and Customers (Doctors and Nurses) by Leveraging Lessons from Amazon, Uber, and Watson.

Description-The super-customers of Big Data and innovations, such as the ability to track one's physical activities and personal genomics, have been, by and large, patients and the healthy future patients. Contrary to that, the originally intended customers of Big Data doctors, nurses, and allied healthcare

Providers have lagged behind relative to the super customers. With patients as super-customers and sitting in the

driver seat as users, there are valuable lessons to be learned from innovative industries.

The lessons learned from successful industries and their businesses, such as Amazon, Apple, Boeing, IBM, Facebook, Google, Mercedes, Microsoft, Samsung, Starbucks, Toyota, and Uber, are instructive in that they share a common theme: consistent, year-after-year accommodation of the needs, pain points, and wants of their customers worldwide. Of all the industries, we think healthcare ought to be the most concerned with providing high-quality service to its super-customers as health is a common denominator in all countries and societies

3) Author Name- Dilsizian SE, Siegel EL.2014 (Artificial intelligence in medicine and cardiac imaging: harnessing big data and advanced computing to provide personalized medical diagnosis and treatment. Curr Cardiol Rep201

new and important scientific information cannot be processed and stored by a single human brain. Physicians learn thousands of different diseases in medical school and are expected to remember and apply a substantial subset of these in daily practice. But it is impossible for an individual physician to keep current on the broad spectrum of new data and discoveries and to reliably recall and utilize that information at all relevant time points. This is part of a major challenge in medical imaging, where real-time errors are estimated to average between 3 % and 5 % and constitute nearly 75 % of medical malpractice claims . Graber et al. estimated that approximately 75 % of diagnostic errors were related to "cognitive factors" Diagnostic errors outnumber other medical errors by 2- to 4-fold and represent nearly 40 % of total ambulatory malpractice claims .

Effective, evidence-based medical practice requires that physicians be familiar with the most recent guidelines and appropriate use criteria. Because of the exponentially growing amount of information in peer-review journals, textbooks, periodicals, consensus panels, and other sources, it is impossible for health care practitioners to keep up with more than a small fraction of relevant literature. Adherence to guidelines and evidence-based medicine may be made even more complex by the variability in "standards of practice" across different communities and states, a variability that complicates the concept of a "gold standard" for diagnosis and treatment of certain illnesses. Advanced computer systems, such as IBM's "Watson" technology, could assist by providing the most upto-

date evidence-based information to inform proper patient care decisions. This information could combine data from a specific patient's history with data from large numbers of other patients with similar disease manifestations. 4) Author Name- Patel VL, Shortliffe EH, Stefanelli M, et al.2009 (The coming of age of artificial intelligence in medicine. Artif Intell Med 2009;

Description- The general AI research community was fascinated by the applications being developed in the

medical world, noting that significant new AI methods were emerging as AIM researchers struggled with

challenging biomedical problems. In fact, by 1978, the leading journal in the field (Artificial Intelligence, Elsevier, Amsterdam) had devoted a special

issue solely to AIM research papers. Over the next decade, the community continued to grow, and with

the formation of the American Association for Artificial Intelligence in 1980, a special subgroup on medical applications (AAAI-M) was created. It was against this background that Ted Shortliffe was asked to address the June 1991 conference of the organization that had become known as Artificial Intelligence in Medicine Europe (AIME), held in Maastricht.

The Netherlands. By that time the field was in the midst of "AI winter", although the introduction

of personal computers and high-performance workstations was enabling new types of AIM research

and new models for technology dissemination. In that talk, he attempted to look back on the progress of AI in medicine to date, and to anticipate the major challenges for the decade ahead.

5) Author Name- Karakülah G, Dicle O, Koşaner O, et al (Computer based extraction of phenoptypic features of human congenital anomalies from the digital literature with natural language processing techniques. Stud Health Technology Inform 2014; 205:570–4. [PubMed]Jha S, Topol EJ. Adapting to Artificial Intelligence: radiologists and pathologists as information specialists. JAMA 2016;

Description-Natural language processing is techniques have gained a wide use in obtaining and evaluating the information especially in life sciences due to the excessive increase of information in recent years. Besides, the automatized extraction of clinical information, such as signs, symptoms, medications and/or observations, from scientific documents and free-text formatted medical records using NLP should be considered crucial for the development of the CDSSs. Here we aimed at developing a computational strategy to extract the phenotypic features, which characterize HCAs from the case reports in the literature via text processing and NLP methods. In addition to this, by using the extracted information, we created an initial framework of an information base for a potential CDSS in the diagnosis of HCAs.

Natural Language Toolkit modules and the scripts written in Python were utilized in all text processing steps on the corpus . At the first step, the string. lowercase() function was used for standardizing the characters in the corpus. In the following preprocessing step, the sentence segments for each abstract were determined with Punkt algorithm. From these segmented sentences in each abstract, the white spaces were removed via tokenization in the third pre-processing steps.

III COMPARISON:

III. COMPARISON:							diagnosis	therapeutic			
							and treatment	records.			
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	Leveragin	broader	surroundin				digital	combination	the		
	g Lessons	social ecosystems	g				with	of big data	literature		
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5	intelligen	drug is at an	demands	handling big			technique	industrial			
	ce in	intersection,	on	data and			S	revolution.			
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	and cardiac	increments in	create	uatabase.		IV. CONCLUSION:					
	imaging:	persistent	greater								
	harnessin	volume, a	risk for								
	g data and	sum and	and			Revie	wed the moti	vation of using A	AI in healthca	re, presented the	
	advanced	multifaceted	therapeuti			variou	is healthcare	data that AI ha	is analyzed a	nd surveyed the	
	computin	nature of	c errors.			major	disease type	s that AI has been	n deployed. W	e then discussed	
	g to provide	and logical				In det	alls the two i	major categories	OF AI devices	s: ML and NLP	
	personali	learning, and				techni	iques: RF and	d neural network	x, as well as t	he modern deep	
	zed	the change to				learni	ng technique	. We then survey	yed the three 1	najor categories	
	medical	electronic				of AI	applications.				

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A successful AI system must possess the ML component for handling structured data (images, EP data, genetic data) and the NLP component for mining unstructured texts. The sophisticated algorithms then need to be trained through healthcare data before the system can assist physicians with disease diagnosis and treatment suggestions.

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