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DEEP LEARNING ANALYSIS TECHNIQUES WITH DIFFERENT PHASES FOR CRITICAL KIDNEY DISEASE PREDICTIVE MODEL

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Abstract - One of the most serious medical problems in the world, critical kidney disease (CKD) with a high incidence of death per capita. Cases usually fail to detect the complaint since there are no outward signs of ongoing kidney failure in its first stages. Pathology data vacuity, machine-literacy usage in healthcare for bracket, and vaticination of complaint have all been more widespread. The SVM method, light GBM, and logistic retrogression are used to compare the results. cardiovascular disease (CVD) is the main cause of morbidity and death for dialysis users, hence managing this group should be a top priority. There are currently a number of therapies available to slow down the gradual loss of renal function and/or stop the onset of CVD. Low-protein diets, anaemia and calcium-phosphate problem treatment, blood pressure and proteinuria management, and quitting smoking are a few of them. Although prospective, controlled, randomised clinical studies are required to prove the clinical utility of other therapies, such as the administration of lipid-lowering medications, anti-inflammatory pharmaceuticals, and anti-oxidant agents, they are emerging as especially promising therapeutic approaches. Although early and frequent nephrology specialist treatment has been linked to lower morbidity and mortality, intervention in the conservative phase of CKD is anticipated to be more beneficial if carried out as early as possible in the course of the illness. Cases with HIV have an increased risk of developing CKD in a serious condition. Early diagnosis of CKD enables patients to get immediate treatment and prevents the problem from worsening. The employment of machine-literate methods for bracketing and vaticinating complaints in healthcare has become increasingly widespread due to the vacuity of pathology data. In this article, deep learning algorithms are used to provide the CKD bracket. The CKD stages are also computed for individuals who have been diagnosed with CKD and are based on the glomerular filtration rate. 97% of the complexity in differentiating CKD patients from HIV cases may be attributed to the DNN model.

Key Words – Chronic kidney disease; CKD stage recognition, Deep learning (DL), Vector Support Machine, KNN

1. INTRODUCTION

A frequent urologic malignancy of the renal cortex is renal cell carcinoma. The effectiveness of computer-based cell-based melanoma therapy may be significantly influenced by accurate measurement and the appropriate bracket of exrescence.

practice because to both the private (e.g., faulty delineation) and ideal (e.g., a large number of photos) variables involved in the process. Because of this, there is a significant need for computer-supported automated segmentation techniques for order exrescences (in CT images).

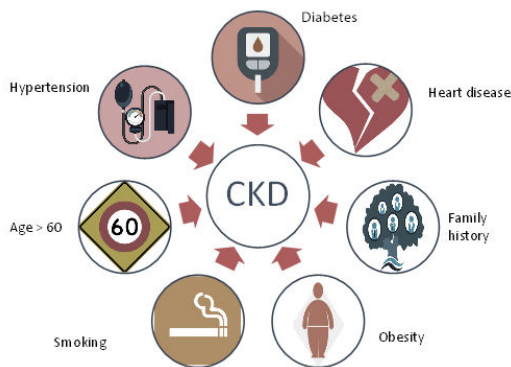


Fig. 1. The factors that lead to chronic kidney disease

In this sense, the correct segmentation of the order exrescence is a critical need for the quantification and bracketing. Conventional mortal- based handmade delineation for order exrescence segmentation is not suitable in clinical

Yet, mechanically separating the order exrescences in CT scans is a very laborious process to do with the findings of clinical and experimental studies. It is difficult to accurately predict the location of distinct order exrescences in medical imaging since the exrescences might potentially arise in veritably different areas across different instances. This makes it difficult to make accurate predictions.

In general, distinct exrescences amongst different instances display veritably variable form appearance and volumetric size according to the various phases of development. Because of the little disparity in CT scans, the textural information of exrescences and the girding apkins that surround them are almost identical to one another.

1.1 KIDNEY TUMOR

A kidney tumour or kidney mass refers to an abnormal growth that occurs in the kidney. The feathers are composed

consisting of a pair of organs shaped like beans, each of which is around the size of a fist. They are protected by the lower caricature pen, which is linked to the top reverse wall of the stomach and holds them in place. Both orders are located just next to the backbone, one to its left and the other to its right. An adrenal gland is a very tiny organ that may be found on the top of each kidney. chubby and a little skinny, stringy tissue referred to as Gerota's fascia surround each kidney and adrenal gland on their respective sides, creating a girdle around the organs. The kidney's primary function is to cleanse the blood by filtering out waste items, excess water, and swabs as it is brought in from the renal highways. These waste chemicals are expelled in the form of urine. Urine is collected in the middle of each kidney in a region that is referred to as the renal pelvis. Urine also exits the special cells via long, thin tubes that are referred to as ureters. Urine is kept in the bladder until it can be expelled by the ureter, which leads to the bladder.

The feathers are also used for the following: Despite the significance of our particular cells, we can only fulfil one request at a time. In the United States, there are a significant number of individuals who are leading normal, healthy lives despite having just one kidney. A medical process known as dialysis is necessary for the survival of certain individuals since their kidneys do not have any specific cells that are capable of functioning. The most frequent kind of dialysis involves the use of a specific machine that filters blood in a manner that is analogous to that of a healthy kidney. Cancer of the kidney develops in the kidney's particular cells. The kidneys in our bodies are bean-shaped organs, and each one is around the size of a human fist. They are situated behind your abdominal organs, one pair on each side of your chine, and are arranged in an order. Renal cell melanoma is the most prevalent form of order cancer seen in adults, accounting for around 90 percent of all malignant excrescences. There are also other, less prevalent forms of cancer that might take place. Wilms' excrescence is a kind of order cancer that is more prone to develop in children who are still quite young. It seems that the incidence rate of order cancer is increasing. It's possible that this is due to the increased use of imaging techniques such as computerized tomography (CT) scans, which might be a contributing factor. It is possible that these tests may result in the unintentional identification of further orders of cancer. In many instances, kidney cancer is detected at an early stage, when it is simpler to cure since the excrescences are still tiny and contained to the kidney. This is the situation when kidney cancer is first diagnosed.

1.2 DEEP LEARNING

The prophetic analytics tools that are available today are driven by a variety of distinct models and algorithms, each of which is adaptable to a broad variety of application situations [4]. Getting the most out of a predictive analytics result and utilizing data to make insightful judgements both require that you first determine what predictive modelling approaches are

appropriate for your organization and then implement those approaches. Within the realm of statistics, the term "deep learning" refers to an application of artificial intelligence in which readily accessible information is mined by means of algorithmic processes in order to either aid in or facilitate the analysis of statistical data. Although it does incorporate certain aspects of robotization in general, deep learning still needs human supervision.

Deep learning requires a high level of idea in order to produce a system that functions effectively on data situations that it has not yet seen. The subfield of computer science known as deep learning offers a variety of approaches to data analysis and has only been around for a very short time. Some of these methods such as logistic retrogression and top element analysis are based on well-established statistical approaches, but many of these methods are not. On a diet for CKD, some foods will be restricted or eliminated in order to preserve our kidneys, while other foods will be included into the plan in order to keep you energized and fed. The precise diet that CKD patients follow will be determined by whether they are on dialysis, as well as whether or not they are in an early or late stage of CKD [5, 6].

The paradigm that most statistical methods adhere to is one in which a specific probabilistic model is selected from a group of related models in order to provide an accurate description of the observed data [7]. In addition, the majority of approaches to deep learning are intended to locate models that provide an excellent fit to the data (i.e., they solve specific optimization problems), with the exception that these approaches to deep learning no longer limit themselves to probabilistic models. Because of this, one benefit of Deep learning methods over statistical bones is that the latter hold the underpinnings of probabilistic models, whilst the former do not. In point of fact, while certain methods of deep learning employ probabilistic models, the traditional statistical methods are usually insufficiently flexible for the next era of big data. This is due to the fact that data sources are becoming more complicated and multi-faceted.

It is possible that defining probabilistic models that relate variables from disparate data sources in a way that is presumptive and can be subjected to statistical analysis may be exceedingly difficult, if not completely impossible. It's possible that deep learning might be used to provide a wider class of more flexible and adaptable data analysis methods that are better suited to contemporary sources of information.

It is imperative for statistical agencies to explore the possible use of Deep learning ways to determine whether their unborn requirements might be better met with similar methods than with traditional bones. This question can be answered by determining whether or not deep learning can be used to learn from data. Meanwhile [11], a deep learning model-based autonomous or self-diagnosing approach is suggested to

quickly and accurately identify all three illnesses. [12] Proposed a protocol-based framework to secure the leakage of data in the smart farming system using data security techniques. [13] The experimental results showed that the singular value decomposition entropy is more stable and deep CNN models always performed better for this entropy.

2. LITERATURE SURVEY

Jinming Duan and Ghalib Bello proposed in the field of cardiac magnetic resonance (CMR) image segmentation, techniques that make use of deep learning have reached the state of the art. The object identification of anatomical shape priors has entered a lower attention state, whereas these approaches have still focused on learning image intensity features for the purpose of segmentation. Using atlas propagation, we combine a multi-task deep literacy approach to develop a shape-meliorated bi-ventricular segmentation channel for short-axis CMR volumetric images. This channel is used to analyse the anatomy of the heart.

The channel begins by using a fully convolutional network (FCN) that simultaneously learns the segmentation and corner localization tasks in order to proceed. Here, the use of FCNs in 2.5D representation aims to combine the computational advantages of 2D FCNs networks with the capacity to deal with the intensity of 3D space, all without compromising the severity of the phase. Additionally, a refinement step is implemented in order to explicitly put previous knowledge into shape and improve the segmentation quality. The step that was mentioned earlier is the one that is most effective for eradicating image remnants.

Andrzej Skalski and Jacek Jakubowski designed a new order segmentation system for Reckoned Tomography case data with order cancer as part of the research. A Hybrid Level Set system with elliptical shape constraints underpins the segmentation process. An entirely automated method of order region bracketing is presented, and it is based on the results of segmentation. RUS Boost and a decision tree-like structure are used as the basis for the identification of the order, excrescence, and vascular tree. With this method, it is possible to solve the primary issues that are associated with the imbalance of region bracket classes and the quantity of voxels that need to be categorized. The bracket is established on 64-element point vectors that were computed for the order region. These vectors correspond to information about the 3D edge, region, exposure, and spatial neighborhood. On a data set consisting of clinically ordered cancer CT scans, the proposed methodology was estimated. The effectiveness of segmentation, as measured by Bones, was equal to 0.85 plus or minus 0.04. The overall degree of complexity of the suggested bracket model is equal to 92.1. The utility of the proposed result is validated by the presented results. We believe that this is the first result that allows the member (peak) order region to

be divided into divisible chambers, specifically the order, excrescence, and vascular tree.

Jong Jin Oh and Jung Keun Lee developed a threat-positioning system that is more precise by investigating the prognostic impact of excrescence growth within adipose apkins girding the order and/or renal tone. We conducted a retrospective review of the medical records of 211 cases with a pathologic opinion of T3aN0M0RCC among renal cell melanoma (RCC) cases from February 1988 to December 2015 according to the number of T3 a pathologies extrarenal fat irruption (EFI) and/or renal venous irruption. This review was based on the number of cases that had a pathologic opinion of T3aN0M0RCC (RVI). During a mean follow-up duration of 38.8 months, the cases that had both pathologies (EFI RVI) had a lower rush free survival (RFS) rate than those who had only a single pathology ($p = 0.001$) The presence of both factors was found to be an independent predictor of RFS (HR = 1.964, $p = 0.032$) through the use of multivariable Cox regression analysis; however, the cancer-specific survival rate was not different among cases with EFI and / or RVI. Cases with pathologic T3aN0M0 RCC that presented with both EFI and RVI were at an increased risk of experiencing rush after nephrectomy. According to the existence of EFI and/or RVI disorders, pathologic T3a RCCs might be subdivided into those with a favourable complaint and those with an unfriendly complaint.

Han Sang Lee and, Helen Hong explained common knowledge that the undertaking is difficult because of the wide diversity of sizes, shapes, and positions involved. This research presents a robotic technique on behalf of identifying and segmenting SRM in discrepancy-enhanced CT images by making use of texture point brackets and environmental point brackets. Initially, the order of regions of interest (ROIs) is established by thresholding intensity and location. Second, the thresholding of intensity and location may dislodge mass campaigners completely. The third step involves reducing the number of false positives using a patch with a grounded texture and an environment point bracket. In the end, mass segmentation is carried out by making use of the findings of the discovery process as a seed. This process involves region growth, active silhouettes, and outlier junking based on size and shape criteria. During testing, our system was able to identify SRM with a particularity of 99.63 and a PPV of 64.2, both of which were independent, and it was able to segment it with a perceptivity of 89.91, a particularity of 98.96, and a DSC of 88.94, all of which were independent.

3. EXISTING SYSTEM

Included characteristics from the super-pixel position are used in the suggested super-pixel bracket-grounded technique, which results in a large increase in the finding of kidney complaints. The evaluation of increased intraocular pressure, often known as IOP, is the method that is first utilised to diagnose renal complaint.

A problem with human kidneys may impair human body's capacity to filter waste products, remove excess water from human blood, and assist keep your blood pressure under control. They are located on each side of our chine, just over our middle section. When our kidneys are compromised, our bodies can make up for lost fluid and waste products on their own. Because of their remarkable flexibility, our kidneys are able to make up for some of the difficulties that might arise as a result of renal disease. Hence, if our kidney impairment progresses slowly over time, then our symptoms will also develop slowly over time. Actually, we may not feel symptoms until our complaint is presented here. There's a chance that we have high blood pressure. Nausea as well as vomiting a loss of appetite, a taste of metal in your tongue, and other symptoms fatigue, weakness, allowing for difficulty sleep difficulties, contraction and twitching of the muscles Swelling in our bases and ankles [8]. A persistent itching that refuses to stop the lungs, if fluid builds up in. The individual may have shortness of breath and chest discomfort. If fluid accumulates around the heart, the individual may experience chest pain.

4. CRITICAL KIDNEY DISEASE PREDICTIVE MODEL

The model that is being offered is an order compliant vaticination utilizing Lite GBM with various deep learning algorithms then enforced. The advancement of technology, which also includes machine literacy, has a significant influence on people's health by facilitating the correct assessment of a broader range of lifestyle habits, which in turn leads to more effective treatment. The proper ventilation of a complaint is the most challenging and time-consuming process. Data mining is a key component of the solution to this issue, since it helps to predict how the problem will be resolved. Analyzes habitual circumstances using machine literacy methods based on a dataset of habitual conditions taken according to device literacy data repository at the University of California, Irvine [9, 10].

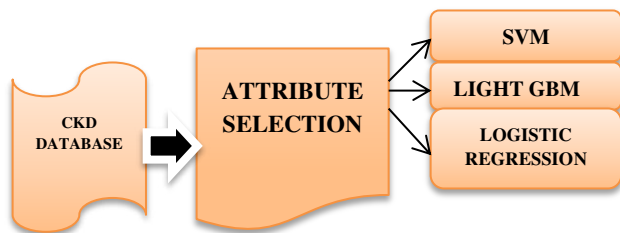


Fig. 2. Architecture of Critical Kidney Disease Model

We employ datasets of heart complaints, order complaints, cancer complaints, and diabetes complaints in order to develop

reliable prediction models for chronic illnesses such as heart disease, order, cancer, and diabetes utilizing data mining techniques. The most relevant characteristics are named from the dataset in order to improve sensitivity and cut down on the amount of time spent training.

An armature is an unseen model that depicts the structure, gesture, and other perspectives of a system. In this case, the system has an armature that describes its structure. A formal description and representation of a system that is arranged in a manner that supports the mind about the structures and activities of the system is what is meant when we say that something is an arm.

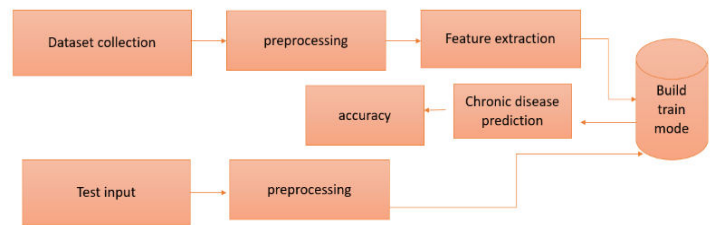


Fig. 3. Flow Process for Critical Kidney Disease Model

Light GBM (Light Gradient Boosting Machine): A grade-boosting frame based on decision treeson Light GBM that is used to improve the efficiency of the model and to decrease the amount of memory operation required. Two new ways are used by light GBM. These are Grade- grounded one Side Slice and Exclusive Point Speeding (EFB), both of which fulfil the imitations of histogram grounded algorithm that is primarily used in all GBDT (Grade Boosting Decision Tree) fabrics. Both of these new ways are used by light GBM. The two methods of GOSS and EFB, which have been discussed, are responsible for the formation of the properties of the Light GBM Algorithm. They operate in tandem to ensure that the model functions correctly and to offer it an advantage over alternative GBDT fabrics. Grade-grounded one side testing fashion for light GBM. The numerous data cases are placed in the computation of information gain in a variety of different orders.

Cases with bigger amounts of information will contribute more to the overall information gain. In order to preserve the nuanced nature of information gain estimate, GOSS only randomly discards instances that have tiny slants and preserves the cases that have big slants in its database. When the value of the information gain spans a wide range, this technique may lead to a more accurate gain estimate than somewhat arbitrary slicing, with the same desired slice rate. This is particularly true when the value of the information gain varies widely.

5. IMPLEMENTATION AND RESULTS

Cases with bigger amounts of information will contribute more to the overall information gain. In order to preserve the delicate

nature of the i5, GOSS maintains the cases that have huge slants and only carelessly discards the cases that have minor slants. The website's table of contents or a mathematical data matrix where each row refers to a particular data piece and each column represents a certain variable in question [11].

Preprocessing of Data: Data filtering is the process of locating and (or deleting) missing entries from a record set, table, or database. In addition to changing, amending, or removing the soiled or coarse data, data filtering also refers to relating the incomplete, erroneous, inaccurate, or inapplicable portion of the data. Collected datasets shown in Table 1. The act of collecting data orders in the database is considered data collection.

Table 3.1 Attributes of CKD Patients Dataset

Attribute number	Attributes	Attribute values	Attribute codes
1	Age	Years	Age
2	Blood pressure	mm/Hg	bp
3	Specific gravity	1.005, 1.010, 1.015, 1.020, 1.025	sg
4	Albumin	0, 1, 2, 3, 4, 5	al
5	Sugar	0, 1, 2, 3, 4, 5	su
6	Red blood cells	Normal, abnormal	rbc
7	Pus cell	Normal, abnormal	pc
8	Pus cell clumps	Present, not present	pcc
9	Bacteria	Present, not present	ba
10	Blood glucose random	mg/dl	bgr
11	Blood urea	mg/dl	bu
12	Serum creatinine	mg/dl	sc
13	Sodium	mEq/L	sod
14	Potassium	mEq/L	pot
15	Hemoglobin	g	hemo
16	Packed cell volume	-	pcv
17	White blood cell count	cells/cumm	wbcc
18	Red blood cell count	millions/cmm	rbcc
19	Hypertension	No, yes	htn
20	Diabetes mellitus	No, yes	dm
21	Coronary artery disease	No, yes	cad
22	Appetite	Good, poor	appet
23	Pedal edema	Yes, no	pe
24	Anemia	Yes, no	ane
25	Class	CKD, NOTCKD	-

Clustering of Data: Clustering is the challenge of organizing a group of things (referred to as a cluster) such that the objects in the cluster are more similar (in some way) to each other than to the objects in other groups. Clustering is another name for cluster analysis.

K Mean: The clustering system begins with the set number of initial clusters, which is k. With each replication, the records are put into the cluster that has the centroid or centre that is geographically closest to them. The distance from each record to the centre of the cluster is determined after each replication, and this distance is recorded.

Classification Data: Categorization is a procedure that takes place inside a repository or database and includes the use of a variety of colorful styles and criteria for the product dataset. SVM bracket refers to the process of arranging data into orders for the purpose of making the most efficient and effective use of it and predicting the outcome. These are often accomplished with the use of a database or some kind of business intelligence software, both of which provide users the power to survey, distinguish, and divide data.

5.1 Stage Identification

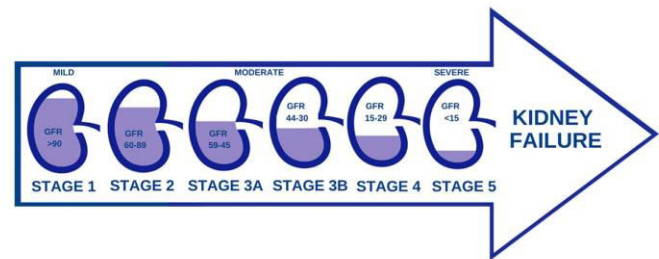


Fig. 4. Stage Identification in Kidney Failure

5.2 Results

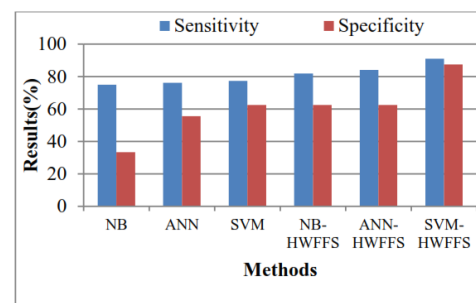


Fig. 5 (a) Comparison of Sensitivity and Specificity

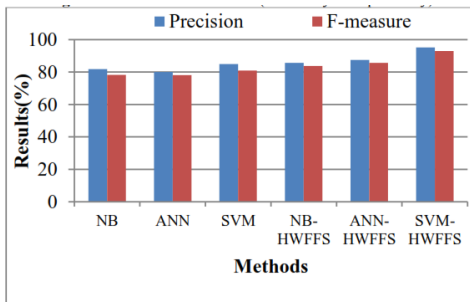


Fig. 5 (b) Comparison of Precision and F-measure

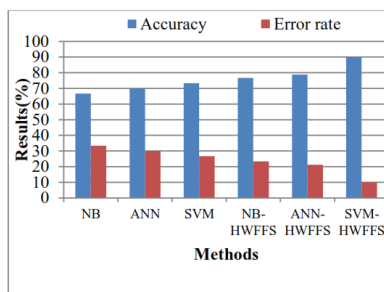


Fig. 5 (c) Comparison of Accuracy and Error Rate

Fig. 5. (a),(b),(c) Comparison of NB, ANN, SVM, NB-HWFFS, ANN-HWFFS and SVM-HWFFS Methods.

Fig.4. depicts the support vector machine and the radial basis function kernel. The effectiveness of chronic renal disease treatments for a variety of distinct algorithms SVM, KNN, and Logistic Regression are three methods of analysis that each have their own unique sensitivity to accuracy. SVM has a sensitivity of 58%, KNN has an accuracy of 87%, and Logistic Regression has a sensitivity of 99.1%.

eGFR Calculation

Nowadays, kidney failure is more common among the approximately 37 million persons in India who have CKD. Monitoring patients with and at risk for CKD becomes crucially essential for lowering morbidity and mortality since research has shown that therapy at early stages is often successful in avoiding or postponing poor consequences.

The Kidney Disease Improving Global Outcomes (KDIGO) guideline defines chronic kidney disease (CKD) as the presence of a glomerular filtration rate (GFR) of 60 mL/min/1.73 m² for >3 months and/or evidence of kidney damage (e.g., structural abnormalities, histologic abnormalities, albuminuria, urinary sediment abnormalities, renal tubular disorders, and/or history of kidney transplantation).

$$eGFR \text{ (mL/min)} = [(140 - \text{age}) \times \text{Wt} / (0.814 \times \text{S.Cr in } \mu\text{mol/L})] \times (0.85 \text{ if female})$$

Testing for GFR, albuminuria, and urine sediment should thus be part of monitoring. GFR was formerly calculated using 24-hour creatinine clearance, however the National Institutes of Health (NIH) and the National Kidney Foundation now advise using estimated GFR (eGFR). The eGFR test based on blood creatinine and/or cystatin C is less complicated and often more accurate. With a glomerular filtration rate, one may determine where one is in the progression of chronic kidney disease.

6. CONCLUSION

In most nations, there are no obstacles to referral to nephrology specialists, indicating that the issue is mostly cultural, i.e., there is a general lack of knowledge of the potential advantages of early and routine therapy by a nephrologist. When considered together, there is now compelling evidence that early and consistent nephrology specialist treatment in the pre-dialysis stage of chronic kidney disease (CKD) is linked to lower rates of morbidity, short-term death, increased long-term survival on dialysis, and lower costs. Yet, it's also crucial to take into account how a broad switch to early referral will affect available resources. As the prevalence of CKD is high and rising, evaluation of the cost effectiveness of screening programmes for groups at high risk of developing progressive CKD (e.g. elderly patients with diabetes or hypertension) and the development of new models of health care delivery, integrating care from nephrologists and other physicians and other healthcare professionals, will inevitably be needed in the future. Yet, the most pressing difficulty facing nephrologists today is undoubtedly emphasizing the critical value of early CKD discovery and the prompt referral of patients to routine nephrology specialist treatment. The fact that a delayed referral to a nephrologist results in "loss of opportunity for the patient, loss of money for society" must be disregarded any longer. HIV-positive patient's chronic kidney disease categorization may be of great assistance to both patients and clinicians in making correct and timely clinical choices. For the purpose of classifying CKD in HIV-positive patients, we examined how well DNN performed in comparison to the most recent and cutting-edge deep learning methods. According to the findings of our study, using Light GBM to classify CKD is quite useful. We also gave an example of how the eGFR calculation may be used to determine the stage of an illness. In this work, the effectiveness of several deep learning algorithms for classifying CKD in HIV-positive patients was assessed and compared.

7. FUTURE WORK

In the future, diagnostics based variety of imaging approaches could be supported by features that are generated by DNN. The suggested model has the drawback of having only been tested on limited data sets. In the future, considerable amounts of more complex and representative CKD data will be gathered to gauge the severity of the illness in order to enhance the model's performance. The specialists in pathology will gather the

clinical data. In the future, a sizable clinical data set based on acid-base parameters, hyperparathyroidism, inorganic phosphorus concentration, and nocturnal urination will be used to assess the performance of the proposed model. In order to assess the prediction accuracy, additional characteristics will also be used to get a wider view on the useful parameters connected to CKD illness.

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