



Tan Teck Jack

TeleMedC Group, Singapore

## An AI-based diabetic retinopathy screening system and its use in real-world clinical settings in Australia and Singapore

The World Health Organization announced in 2018 that 422 million people worldwide suffer from diabetes mellitus. The projected impact of vision impairment and blindness caused by Diabetic Retinopathy (DR) will result in significant public health and economic consequences. DR is preventable and treatable if detected early through an annual eye screening. However, screening rates are low globally due to a paucity of trained eye-health professionals in developing countries and in rural or remote areas of developed countries. Based on the research from CSIRO Australian e-Health Research Centre, TeleMedC group commercialized an AI-based Diabetic Retinopathy screening system-DR grader, an automated DR grading and preliminary referral decision support tool for patients with diabetes. The cloud-based tele-ophthalmology system has the functionalities of: (1) Deep learning based image quality assessment tool; (2) Deep learning based DR “disease/no disease” grading for color retinal images; (3) DR lesion localization and DR level indication; (4) Preliminary report of patient referral/no referral decision; and (5) DR disease audit by eye experts and developing patient referral pathway. DR grader has been deployed in a GP Super clinic at Midland, Western Australia from December 2016 onwards.

### Keywords

Cortisol, Alzheimer’s disease, Diabetes, Folic Acid

### Back Ground

Diabetic retinopathy (DR) is a microvascular complication of diabetes mellitus, leading to progressive damage to the retina. DR has been deemed as a global public health problem, and diabetes-related visual impairment requires early detection to be prevented. From epidemiology studies, approximately one in three persons with diabetes has signs of DR, and a third of these might suffer from vision-threatening retinopathy, defined as a severe non-proliferative diabetic retinopathy or diabetic maculae edema (DME). Recently, DR incidence has been estimated to range from 2.2 to 12.7% and remains a leading cause of vision loss in many developed countries, particularly among the adult working-age population. Management and identification of patients at risk for DR condition by risk stratification are crucial. To prevent development and progression of DR condition, regular follow-ups are

required and DR screening programs have long been recommended for individuals with diabetes. The international Council of Ophthalmology (ICO) has initiated a comprehensive guide for physicians and ophthalmologists incorporating evidence-based principles with real-world experience with recommendations for diagnosis, definition, screening and referral criteria, follow-up, and management options. However, since the implementation and maintenance of comprehensive and effective programs require substantial resources, DR screening is not widely practiced world-wide as a national program, and many patients with diabetes are unaware of their risk of DR and other related complications. AI Application in Healthcare Over the past decades, artificial intelligence (AI) has exploded into the scientific consciousness with breakthroughs that are sparking increasing interests among computer science and medical communities. AI is a branch of computer science in which machines mimic the cognitive function of human mind, making decisions that were regarded to require human cognition. Since the mid-twentieth century, researchers have identified the need for support systems to process the increasing amounts of clinical data required to make clinical decisions. Machine learning (ML) algorithms empower computers that suggest diagnosis or clinical management without direct human intervention, by extracting clinically relevant information from medical data. With recent progress in digitized data acquisition and computing infrastructures, driven by the successful applications of AI and by harnessing the computation power of graphics processing units (GPU), deep learning (DL) has emerged as an attractive ML method within AI. DL was inspired by the operation of the human brain in processing data and involves model architectures called convolutional neural networks (CNN). CNNs operate on the raw data to learn useful features, identifying significant patterns from training datasets. DL has shown substantial potential in medical imaging. These technologies have achieved robust diagnostic performance in detecting major medical conditions from several medical specialties including radiology, dermatology, ophthalmology, pathology, and pediatric diseases. DL-based tool impact is manifold: for clinicians, as image interpretation assistance; for health systems, as workflow improvement; and for patients, in terms of more reliable diagnosis. Although DL-based technologies in medicine are advancing rapidly and multiple retrospective studies have demonstrated that AI can perform on par with clinical experts, most of these tools have not been evaluated in controlled clinical prospective studies. The real-world implementation into patient-care settings is hindered by several practical issues, such as algorithm transparency, records sharing, privacy, and data standardization. In addition, DL algorithms are extremely 'data hungry' and large resources of medical data are required to conduct large-scale studies. However, given the rapid and impressive progress of AI technologies, the medical community is cautiously optimistic that AI might enable higher-capacity and lower-cost care in reality. In the field of ophthalmology, AI using ML and DL has been broadly studied. Over the past few years, several applications in fundus photography, optical coherence tomography (OCT), and visual fields have shown clinically acceptable performance. Specifically, robust classification performance in detection of age-related macular degeneration (AMD), glaucoma, and retinopathy of pre-maturity has been achieved from retinal fundus photographs. Promising diagnostic accuracies on par with clinical experts have also been obtained for DR detection. DL in ocular imaging has been shown to be effective when applied to OCT for AMD automated detection, macular telangiectasia automated detection, visual

acuity out-comes, and DME-automated segmentation and de-tection. DL techniques have also been successfully applied for the detection and quantification of different types of macular fluid from OCT images. In addition, performance comparable to human experts in referral triage decisions and classification of 10 OCT pathologies with DL strategies have been attained. DL has also been applied to non-traditional tasks on retinal images such as to estimate refractive error, quantity of anterior chamber inflammatory cells, forecast future Humphrey visual fields, and predict cardiovascular risk factors.

**Results:** Results of this implementation were published in a JAMA Network Open article (September 2018) evaluated a total of 291 patients. The system correctly identified all 12 patients with true disease (sensitivity 100%) and misclassified 23 patients as having disease (specificity 92%). The DR grader has been undergoing testing in Singapore since early 2018 at the Department of Ophthalmology, National University Hospital and in 30 GP clinics with similar or better preliminary results pending publication.

**Conclusion:** The AI-based DR screening system provides quick DR patient referral decision support in the primary care setting. It benefits patients from poorly-resourced and underserved remote areas for its low cost, time savings and high patient acceptability. The system was well received by primary care providers.