

# AEye<sup>™</sup> Big Data Challenges and Opportunities

Anthony Vipin Das, FRCS

L V Prasad Eye Institute



## **Financial Disclosures**

None







# Building a DIGITAL Institution







## Vision Centre Complex

#### 202 Centres - 8 yrs













5 Million Consults and counting...



















# Can Research be in REAL TIME? Bridging the GAP in Human Understanding



# Asking the Question

**Disease Progression** 

**Image Analysis** 

**Surgical Outcomes Prediction** 

**OPD** Forecasting

**Population Health** 





# H-K-M-B-T



# **Tools for Analysis**





# **Continuous Data Pipelines**







#### **AEye Pipeline**

## eyeSmart Rural App (N=501,771)



#### **AEye Pipeline**

So that all may see







#### **AEye Pipeline**



#### **ORIGINAL** INVESTIGATION

#### The Incidence of Lacrimal Drainage Disorders Across a Tertiary Eye Care Network: Customization of an Indigenously Developed Electronic Medical Record System—eyeSmart

Anthony Vipin Das, F.R.C.S., M.D.\*, Suryasnata Rath, F.R.C.S., M.D.†, Milind N. Naik, M.D.†, and Mohammad Javed Ali, F.R.C.S., Ph.D.†

\*Department of eyeSmart EMR & AEye, and †Govindram Seksaria Institute of Dacryology, L.V.Prasad Eye Institute, Hyderabad, India

## Lacrimal Disorders

 TABLE 1.
 Incidence of lacrimal drainage disorders by diagnosis

Diagnosis	No. of patients (%)
1. Primary acquired nasolacrimal duct obstruction	10,364 (51.56)
2. Congenital nasolacrimal duct obstruction	5394 (26.83)
3. Acute dacryocystitis	1074 (5.34)
4. Punctal stenosis	603 (3.00)
5. Failed external DCR elsewhere	460 (2.29)
6. Secondary acquired lacrimal duct obstruction	298 (1.48)
7. Functional epiphora	271 (1.35)
8. Failed endoscopic DCR elsewhere	235 (1.17)
9. Lacrimal abscess	194 (0.97)
10. Bicanalicular obstruction	182 (0.91)
11. Canaliculitis	163 (0.81)
12. Canalicular laceration	109 (0.54)
13. Punctal agenesis	101 (0.50)
14. Isolated lower canalicular obstruction	75 (0.37)
15. Acquired lacrimal fistula	72 (0.36)
16. Congenital lacrimal fistula	65 (0.32)
17. Common canalicular obstruction	60 (0.30)
18. Canalicular stenosis	54 (0.27)
19. Isolated upper canalicular obstruction	50 (0.25)
20. Incomplete punctal canalization	43 (0.21)
21. Peripunctal lesions	37 (0.18)
22. Atonic lacrimal sac	36 (0.18)
23. Failed dacryocystectomy elsewhere	22 (0.11)
24. Punctal ectropion	20 (0.10)
25. Failed laser DCR elsewhere	19 (0.09)
26. Rare punctal disorders	19 (0.09)
27. Canalicular wall dysgenesis	18 (0.09)
28. Lacrimal sac tumors	17 (0.08)
29. Lacrimal sac diverticula	17 (0.08)
30. Lacrimal sac dacryoliths	16 (0.08)
31. Centurion syndrome	10 (0.05)
32. Hemolacria	4 (0.02)
DCR, Dacryocystorhinostomy.	



#### Key Insights

#### 32 Diseases

No Laterality

2/3<sup>rd</sup> Females

20,102 patients



# 1.45 Million

Rao Donthineni, Pragnya & Kammari, Priyanka & Shanbhag, Swapna & Singh, Vivek & Vipin Das, Anthony & Basu, Sayan. (2019). Incidence, demographics, types and risk factors of dry eye disease in India: Electronic medical records driven big data analytics report I. The Ocular Surface. 10.1016/j.jtos.2019.02.007.

## Dry Eye Disease (N=21,290)





#### Professional Work (OR 1.5), Affluence (OR 1.6-3.2)







#### **Bimodal** Distribution (Gender)









## The AEye Pipeline

Pavan Verkicharla & Anthony Vipin Das, Community Eye Health. March 2019

### Refractive Error Prediction (N=333,000)





### Patient Forecasting (N= 3,384,157)



0.5

0.4



#### Increasing patient inflow

over 8 years





Frequency



#### SARIMA Modeling Results



Green = Actual Value ; Red = Predicted Value

Prediction Error <20% for 52 weeks - 63% to 96%





# GIGO

## **Current Scenario**



(1) Researchers lack consensus about the operational definition of Big Data in healthcare;

(2) Big Data in healthcare comes from the internal sources within the hospitals or clinics as well external sources including government, laboratories, pharma companies, data aggregators, medical journals etc.;

(3) Natural language processing (NLP) is most widely used Big Data analytical

technique for healthcare and most of the processing tools used for analytics are based on Hadoop;

(4) Big Data analytics finds its application for clinical decision support; optimization of clinical operations and reduction of cost of care

(5) Major challenge in adoption of Big Data analytics is non-availability of evidence of its practical benefits in healthcare





# 1 Billion











Data is not information,

Information is not knowledge,

Knowledge is not understanding,

Understanding is not wisdom.

Wisdom is not Impact!



## So that all may see