

Improving ED Flow & Safety with Predictive, Real-Time Technology

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About Alcidion

- ASX listed health informatics company based in Adelaide
- Proven track record in innovation
- Customers currently large public hospitals
 - Western Health (Vic)
 - Northern Territory
 - Royal Melbourne
 - Tasmania
 - Midcentral DHB, NZ
- Engagement with Calvary Care







Challenges of Healthcare IT

- Health IT systems slow down clinicians
 - Not integrated into workflows
 - Lack of focus on clinical productivity
 - Lack of focus on safety
- Lack of interoperability
 - IT vendors 'hide data' to prevent competition
 - Innovation is stifled
- Most data entered in EMRs is unstructured
 - Prevents clinical decision support & automation
 - Prevents machine learning







AMA College of Physician survey on attitudes to EMRs (2014)

Paved with Good & Bad Intentions

- By early 2000's strong evidence of clinical benefits of Clinical Decision Support (CDS) integrated into EMRs
- The EMRAM & HiTech Act
- Data hiding as a strategy
- Benefits of CPOE meds
- "CDSs ... seem to reach a plateau with respect to their effectiveness when implemented in real-world settings" Am J Public Health. 2014. 104:e12–e22



HIMSS EMRAM								
Stage	Cumulative Capabilities							
7	Medical record fully electronic; HCO able to contribute CCD as byproduct of EMR ; Data warehousing in use							
6	Physician documentation (structured templates), fu CDSS (variance & compliance) , full R-PACS							
5	Closed loop medication administration							
4	CPOE, CDSS (clinical protocols)							
3	Clinical documentation (flow sheets) , CDSS (error checking) , PACS available outside Radiology							
2	Clinical Data Repository, Controlled Medical Vocabulary, Clinical Data Support System							
1	Ancillaries– Lab, Rad, Pharmacy							



The Results



Leading causes of death in the US





Open standards, real-time platform for healthcare decision support and analytics

- Logistics Blazingly fast, platform for innovation
- Clinical risk Make the right thing to do the easiest thing to do
- Revenue Track patient complexity
- In the cloud, on premise



Miya ED Clinical Risk Management















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Adaptive User Interface





Standard configuration for department





Stream specific elements







Miya ED Benefits

- Patient treatment commences an hour earlier, leading to faster recovery
- Redundant test orders are reduced 5% savings in ED pathology costs
- Critical (abnormal) test results are properly witnessed for follow-on action • Without Miya ED around 40% of test ordered are not read
- - 18% of these will be critical
- Miya ED saves ED Heads/Senior Consultants several hours a day
- Care team satisfaction with Miya ED is very high rated at better than 80% compared to other clinical systems at around 30%



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Real-Time Metrics







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Mailing to be Admitted



Peterla



Predictive Analytics



Deep Learning Performance

- Good at perception type problems
- Image recognition is a 'solved problem'
- State of the art performance
 - Mammographic mass classification
 - Diabetic retinopathy
 - Lung nodule detection on CT
 - Breast cancer metastases detection in lymph nodes
 - Skin lesion classification









Building Statistical Models





Training Deep Models





Prediction for the ED?

- We can now use more contextual data for prediction
- Per patient prediction of admission & destination
 - Age, triage, presenting problem
 - Previous visits, labs, medications
 - Problem list (or ICD codes)
- ED Occupancy
- Resource utilisation
- Detection of patients ready for discharge







Admission Prediction AUC=0.84

Occupancy Prediction

12 hour occupancy prediction model



• Multiple models (e.g. 1, 3, 6, 12, 24 hour models)









Better, faster decisions

Thank You