

AI in a SMART Airport

Steve Lee

CIO & Group SVP(Technology)

Changi Airport Group (Singapore) Pte. Ltd.

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Changi Airport : Our Terminals

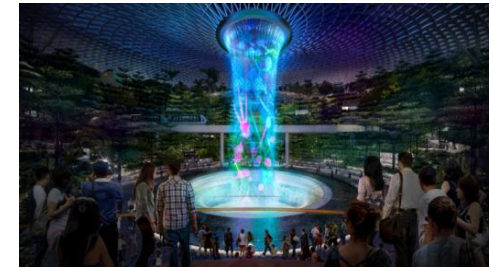


- Four main passenger terminals at Changi (82 Million passenger capacity)
 - Terminal 1: Opened in 1981, refurbished in 1995, recently completed upgrading
 - Terminal 2: Opened in 1990, upgraded in 2006
 - Terminal 3: Opened in 2008
 - Terminal 4: Opens on Oct 31 2017 (16 MPPA)



Opens 2018: Mixed-use complex, incorporating aviation, retail & leisure.

- Singapore's largest indoor garden and vortex
- Fusion of nature and retail space



Many partners, many missions, **ONE**Changi



200+ organisations

10,000 frontline staff

40,000 staff

1,800 CAG staff

Creating a service ecosystem

To provide **160,000** customers ONEChangi experience every day

Changi is a Complex System of Systems



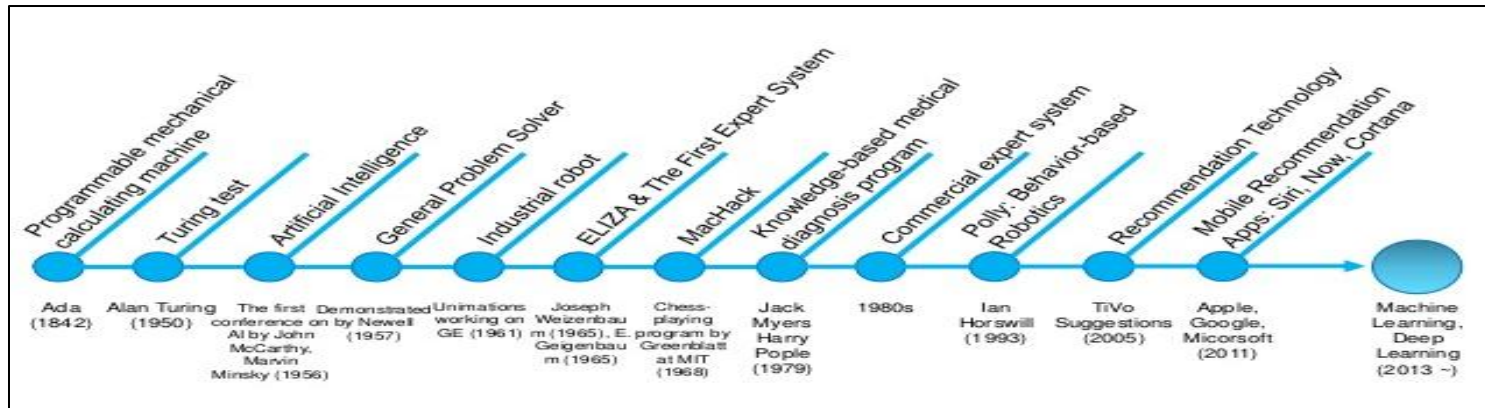
- Activities include Airport Operations, Commercial, Air Hub & Emergency Services.
- Four Terminals with floor area > **1,000,000 m²** with > **70,000 m²** of retail space.
- **58.7 million** passengers annually.
- **A flight every 90 seconds.** 107 scheduled airlines, 380 city links to 90 countries.

How to handle the Diversity, **Complexity** & Scale?

Can AI really solve all our problems? Ask Siri or Alexa.



AI has been around. So what is different now?



I have been following "AI" since last wave circa 1982...

Heuristics in Search

Evans & Sunderland
GPU for simulators

Local Optimisation e.g. Genetic Algorithms, Neural Nets

Autolisp in Autocad

Image Recognition
for military applications

Expert Systems
Simulation & Ops Analysis

Turbo Prolog : Chess, Language Interface,
Knowledge Representation

OpenGL and Vector based "language" for
graphics display across narrow band networks

Programming Text Adventure
Games, Experiments in ELIZA

Smalltalk & other
interesting "grammars"

Access to massive amounts of **usable data through sensors & massive computing power easily available.**

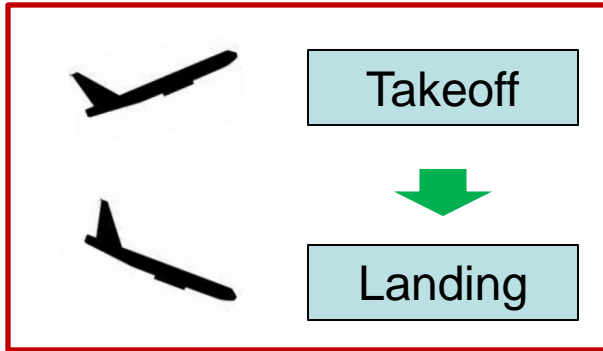
CAG is investing in AI to improve/transform

Challenge in past few years was access to usable data

- Airport Service and Retail
 - **Ops Anticipation – Example of Arrival Journey**
 - Retail / Customer Genome
 - Customer Engagement Channels
 - Chatbots
 - Language Translation to help service staff and our digital channels
- Security & Safety
 - Image Recognition
 - Analytics applications

Must always answer “How can we improve the lives of our customers and employees?”

Journey of Arriving Passengers



Has the plane taken off?

How many people are on board?

Important to predict landing time due to downstream impact.

How early can we know its predicted landing time?

Any possible congestion on arrival e.g. stand conflict?



At Gate

When will it reach the stand? Enough trolleys?
Any potential immigration queue issue?



Immigration

Any possible congestion on arrival?
Right number of resources deployed?



Baggage Claim

Bags unloaded on time? Distance from stands?
Any possible congestion at claim hall?



Taxi Queue

Predicted taxi demand? Taxi queue buildup?
Taxi supply situation – need to activate?

Every day 1,000 flights take off and land in Changi airport from all over the world carrying 160,000 passengers

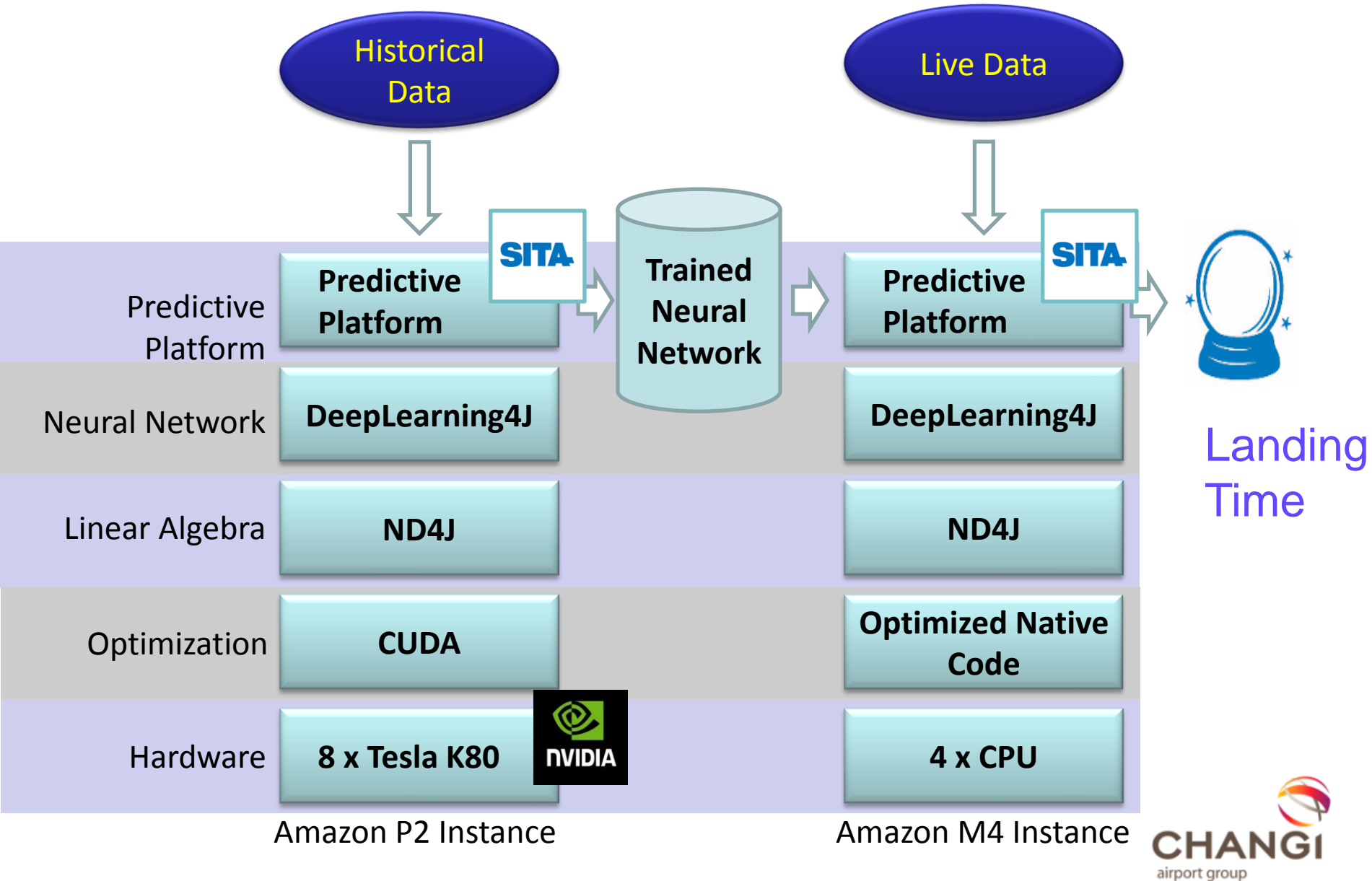


Predicting Landing Time : AI problem

We adopted a “Hybrid” approach

- Input data:
 - ADS-B location information
 - Flight data from open sources
 - Air Traffic Information e.g. runway in use direction, standard terminal arrival routes (STARS) chosen
 - Weather information e.g. bad weather or wind direction
 - Flight information e.g. aircraft type, pax
 - Actual landing time
- Output data:
 - Predicted landing times upon takeoff
 - Predicted landing times (2 hours away from actual landing time)

We partnered with SITA to set this up.



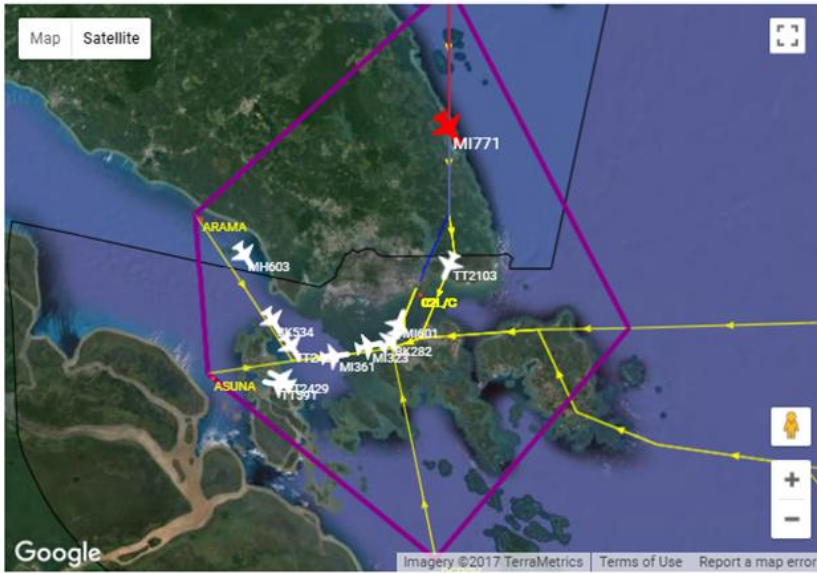
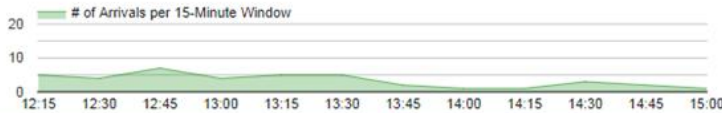
Some Technical Points

- Predictive platform is based on machine learning, hence it is split into two modules; a learning model (on the left) and the real-time predictive module (on the right).
- The neural network implementation that we use is provided by DeepLearning4J. DeepLearning4J is supported by ND4J, which is a high performance linear algebra library.
- ND4J is very portable and flexible. It allows us to reduce the training time of the neural network.
- Need to be able to have quick and easy access to GPU computing power. We can re-train our neural network very frequently, and in this way continuously improve the prediction accuracy.

Based on current training improvement, a usable model can be ready in a few months' time

User Interface is important for AI development

SITA Predictive Platform - Arrivals SIN @ 12:15:03(Asia/Singapore) -- Active Runway: 02L/C



12:00	12:30	13:00
10,000m	10,000m	10,000m

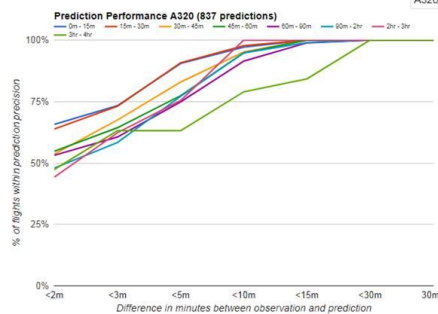
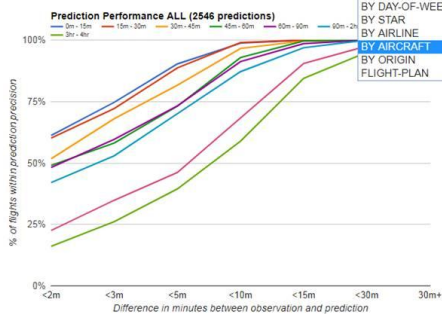
FLT#	AIRCRAFT		ORIG	OFF	STATUS	DATA		APPROACH		ON	
	TAIL	TYPE				POS	F/P	STAR	TIME	PRED	ACT
MH643	9MMXU	B738	KCH	10:56:29+	landed			KART01A	11:32:27	12:02:13	12:03:28
3K556	9VJSB	A320	SGN	09:32:04+	landed			ELALO1A	11:26:02	12:04:03	12:10:28
6E51	VTIEM	A320	MAA	05:51:22	landed			ARAMA1A	11:45:54	12:04:51	12:12:15
811	VNA690	A320	SGN	09:29:53+	landed			ELALO1A	11:27:22	12:05:45	12:15:46
MI323	9VSLI	A320	KUL	11:19:51	onapproach			ARAMA1A	11:51:36	12:10:51	
3K282	9VJSL	A320		10:06:37+	onapproach			ARAMA1A	11:49:58	12:11:19	
MI601	9VLSL	A320	PNH	09:24:00	onapproach			ELALO1A	11:33:48	12:12:20	
MI361	9VSLI	A320	LGK	10:51:00	onapproach			ARAMA1A	11:54:27	12:13:51	
TT2429	9VTAE	A320	PEN	11:04:00	onapproach			ARAMA1A	11:57:17	12:18:11	
TT591	9V0FI	B788	JED	22:31:00+	onapproach			ARAMA1A	12:04:11	12:22:50	
TT2103	9VTAN	A320		08:50:00	onapproach			BIKTA1A	11:56:11	12:23:31	
TT2455	9VTRM	A320	KUL	11:32:00	onapproach			ARAMA1A	12:07:14	12:27:05	
3K534	9VJSH	A320	HKT	10:07:43+	onapproach			ARAMA1A	12:08:35	12:29:47	
MH603	9MMSF	B738	KUL	11:51:04+	onapproach			ARAMA1A	12:11:32	12:30:43	
MI771	9VSBF	A319	USM	09:47:00	onapproach			BIKTA1A	12:03:34	12:32:16	
CZ353	B6288	A320	CAN	09:05:03	departed			VELPI1A	12:22:28	12:40:38	

SITA Predictive Platform - Performance Reports



1 DAY 1 WEEK 1 MONTH 2 MONTHS 3 MONTHS

- BY AIRCRAF
- ALL
- BY TIME-OF-DAY
- BY DAY-OF-WEEK
- BY STAR
- BY AIRLINE
- BY AIRCRAFT**
- BY ORIGIN
- FLIGHT-PLAN



Live Demo



Improving passengers and employees' lives through landing prediction:

- Improved information sharing with meeters and greeters
- Improved resource allocation for all touchpoints especially aircraft turnaround
- Arrival trolleys fulfillment
- Minimise stand conflicts
- **Reducing immigration queue**
- Improving taxi queue demand forecast

The better estimate is necessary for improvement but not sufficient. Other measures are needed e.g. concept of operations changes.

Another Possible AI Project



Takeoff

Has the plane taken off?
How many people are on board?



Landing

How early can we know its predicted landing time?
Any possible congestion on arrival e.g. stand conflict?



At Gate

When will it reach the stand? Enough trolleys?
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Immigration

Any possible congestion on arrival?
Right number of resources deployed?

Predict possible immigration queue buildup



Baggage Claim

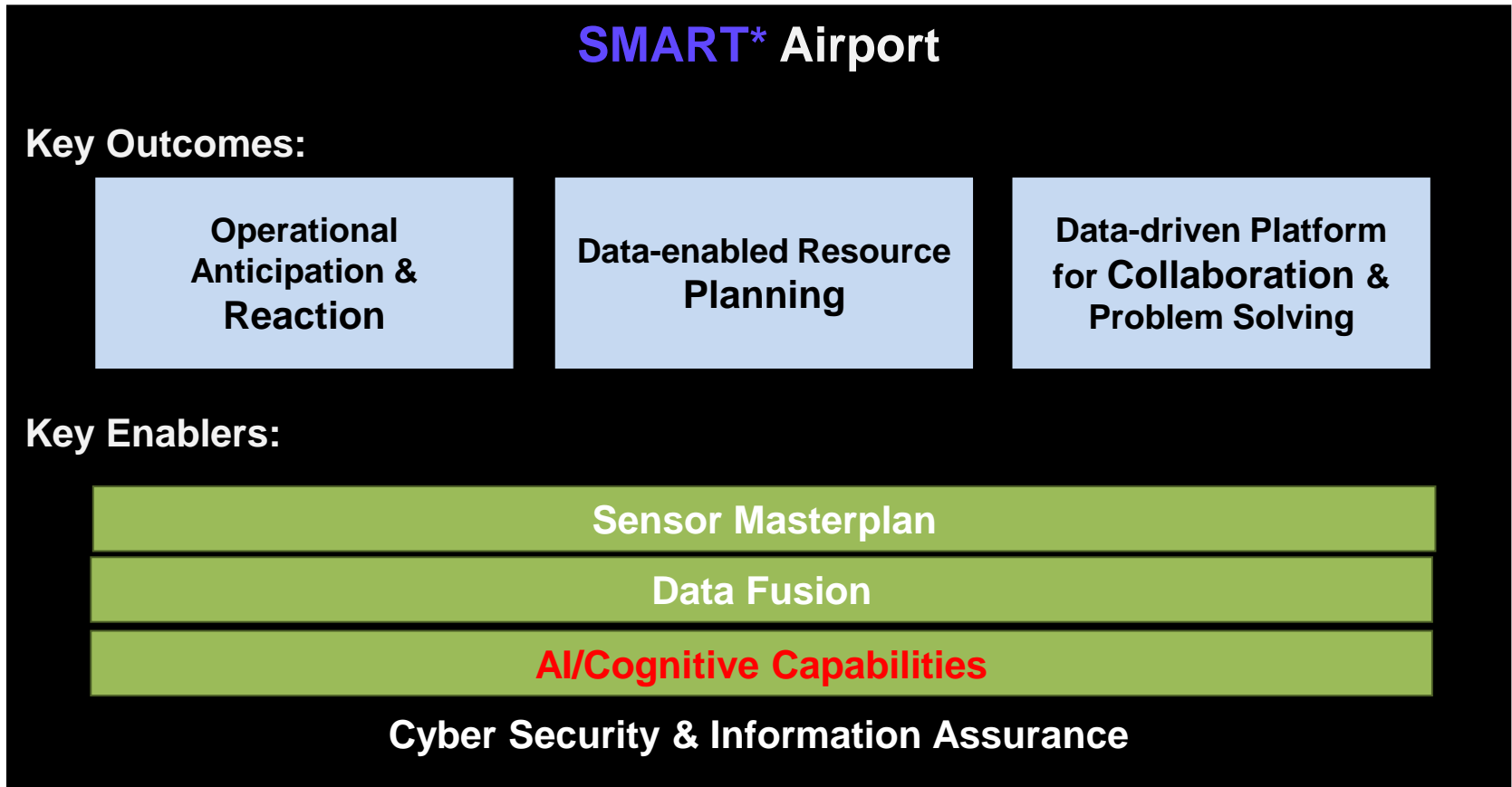
Bags unloaded on time? Distance from stands?
Any possible congestion at claim hall?



Taxi Queue

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Taxi supply situation – need to activate?

AI is a key enabler in a SMART* AIrport

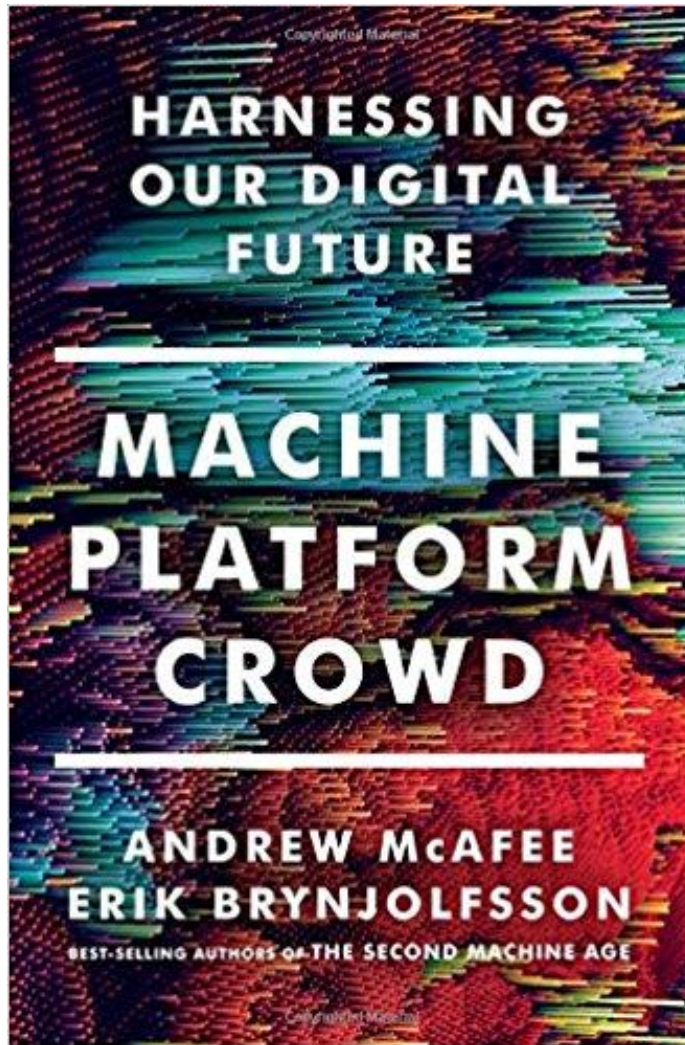


***SMART** : **S**ervice, **S**afety and **S**ecurity **M**anagement through **A**nalytics and **R**esource **T**ransformation



CHANGI
airport singapore

SMART Airport : Smart People augmented by AI



Thanks!

Recommended to me by Steven Miller, Vice Provost (Research) and immediate past dean School of Information Systems, Singapore Management University.