





Leisure and Tourism Data and Decision Analytics for Experience Management



Hoong Chuin LAU Singapore Management University





Analytics for User's Dynamic Experience in Leisure and Tourism

• YouTube Link

- http://youtu.be/v8NM6E96s1g









Partner:





Research:

Analytics:

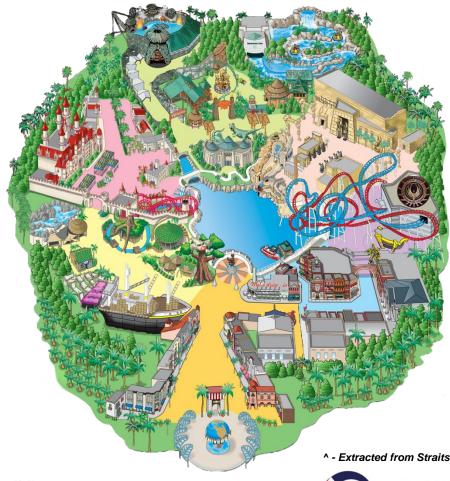
Dynamic Experience Management Revenue Management Experiments, Systems, and Tools







Our Research Playground 1



- Universal Studios Singapore (USS) welcomed its 10 millionth visitor in Apr 2013[^]
- USS drawing an average 9,500 visitors daily*
- USS consists of 7 Zones with more than 20 attractions and shows, along with more than 30 F&B outlets and Merchandise shops

*- Based on Genting Singapore's Annual Report 2012 • - Extracted from Straits Times – Universal Studios Singapore Hits 10 Million Visitor Mark dated 18 Apr 2013







Motivation

- Visitor's Perspective:
 - Dynamic Guidance (experience maximization)





- Operator's Perspective:
 - Crowd Coordination (better facility utilization)

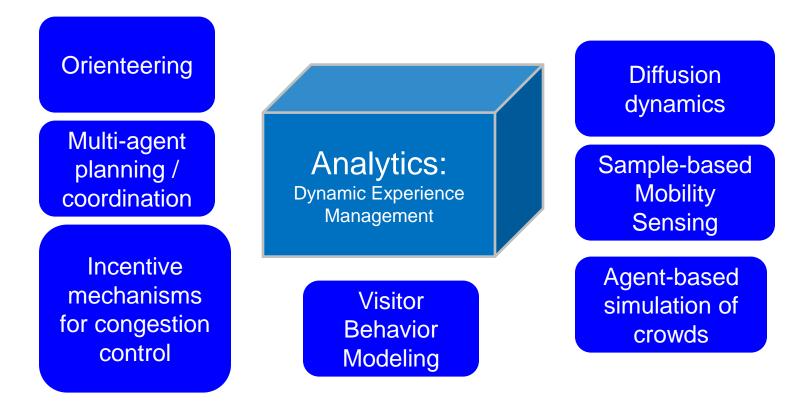






Research Overview

Data: Attraction wait times, visitor trajectories from ground surveys **Theme:** Experience management via real time tracking, route guidance and incentive-based coordination in a dynamic network environment

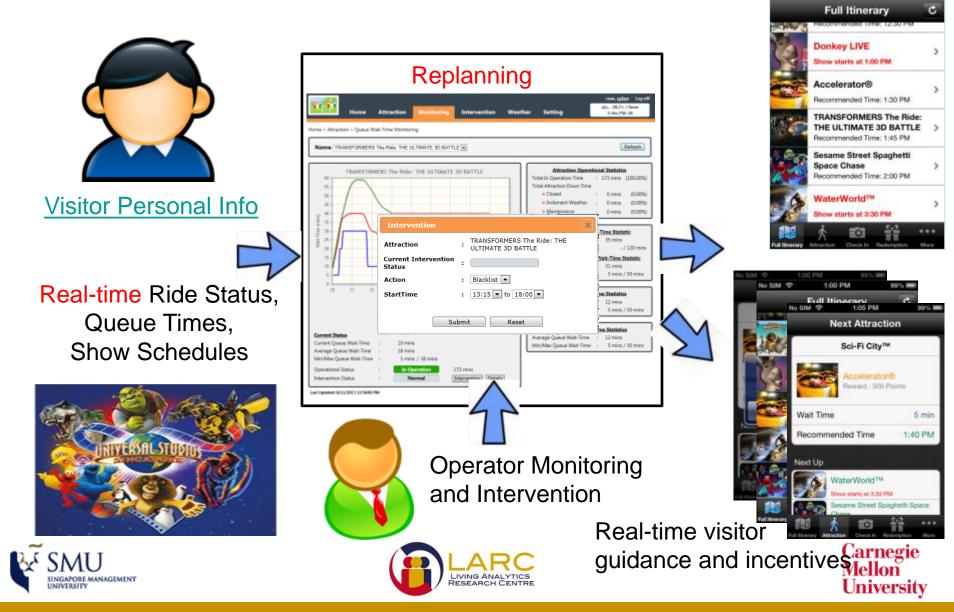








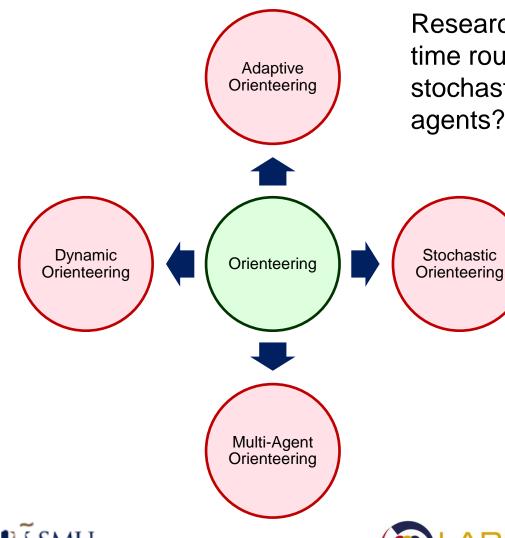
Real-time Guidance and Intervention



No SIM 3

12:54 PM

Orienteering



Research Question: How to provide realtime route guidance in a dynamic and stochastic environment involving multiple agents?

> H. C. Lau, W. Yeoh, P. Varakanham. Dynamic Stochastic Orienteering Problems for Risk-Aware Applications. Conf. on Uncertainty in AI (UAI), August 2012.

> P. Varakantham and K. Akshat. Chanceconstrained Stochastic Orienteering Problem. Algorithmic Decision Theory (ADT), November 2013.

> Chen Cen, S. F. Cheng, and H. C. Lau. Multi-Agent Orienteering Problem. Web Intelligence and Agent Systems Journal. Conditionally Accepted.



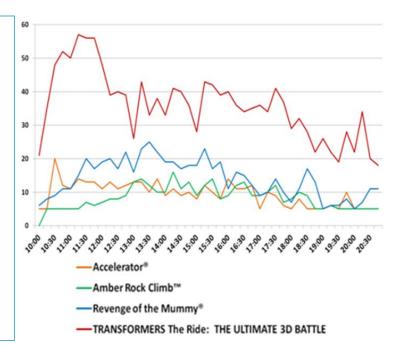




Incentives for Congestion Smoothing

Research Question: Given

- Agents and their Utility Functions
- Resources
- Budget (incentives)
- Desired Resource Congestion Thresholds
- Find **Nash Equilibrium** (allocation of incentives to agents at different time points) that meets desired congestion thresholds



G. Gordon, P. Varakantham, W Yeoh, H. C. Lau, A. Aravamudhan and S. F. Cheng. "Lagrangian Relaxation for Large-Scale Multi-Agent Planning". International Conference on Autonomous Agents and Multiagent Systems (AAMAS), May 2012.

P. Varakantham, N. Fu, W. Yeoh, S. Cheng and H. C. Lau. Budgeted Personalized Incentive Approaches for Smoothing Out Congestion in Resource Networks. Algorithmic Decision Theory (ADT), November 2013.

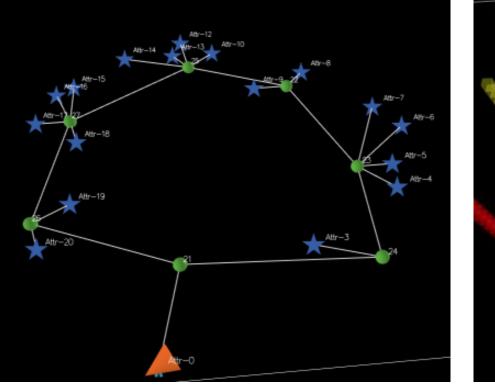


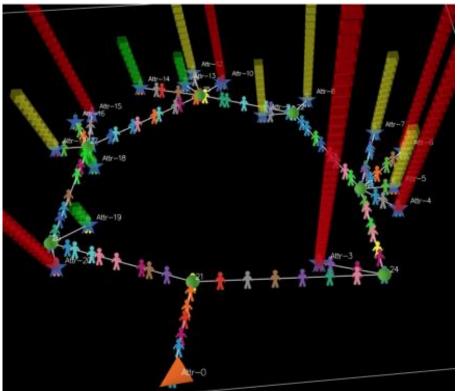




Agent-based Simulation

- Features:
 - Populated with 10,000 visitor agents with preferences.





Carnegie Mellon

University

S.-F. Cheng, L. Lin, J. Du, H. C. Lau, and P. Varakantham. An Agent-Based Simulation Approach to Experience Management in Theme Parks. Winter Simulation Conference (WSC), December 2013.

IVING ANA



Our Research Playground 2



Motivation

- Visitor's Perspective:
 - Dynamic Guidance (experience maximization)
- Operator's Perspective:
 - Bundle design (revenue management)



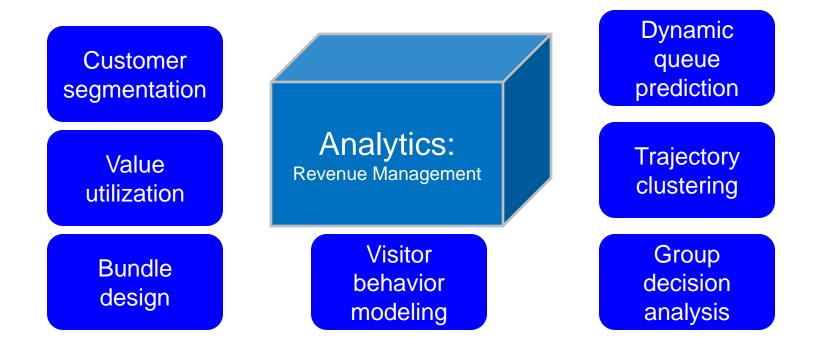






Research Overview

Data: Playpass redemption records **Theme:** revenue management through better understanding of consumer behaviour and market segmentation







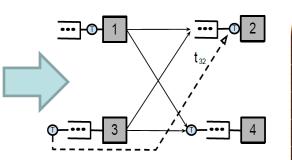


Dynamic Queue Prediction Scope of Work Key Idea

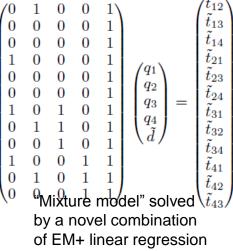
- Using Playpass data entry times to *infer "queuing delays" at different attractions.*
- Depart from classical queuing theory approach and provide a data-driven and online approach for estimating the timevarying queuing delays experienced at different attractions in a theme park
- Modeled via a statistical technique of Expectation Maximization (EM) applied to a "network" problem

Outcome

Experiments on simulated network data shows that our approach can predict queuing delays with estimation errors of less than 10% for queues with medium to long delays.



Problem modeled as heterogeneous network of "queues"



lellon

Universitv

Variable	Mean Delay	30%-70%	50%-50%	70%-30%
q_1	4.00	3.67	3.88	4.24
	8.00	6.56	8.15	9.14
$q_2 \\ q_3$	12.00	10.67	12.19	13.27
$q_4 \\ q_5$	16.00	14.56	15.92	17.69
	20.00	17.61	20.07	22.02
q_6	24.00	21.79	24.07	26.16
$egin{array}{c} d_1 \ d_2 \end{array}$	15.00	15.59	14.86	16.01
	35.00	34.04	34.96	34.64

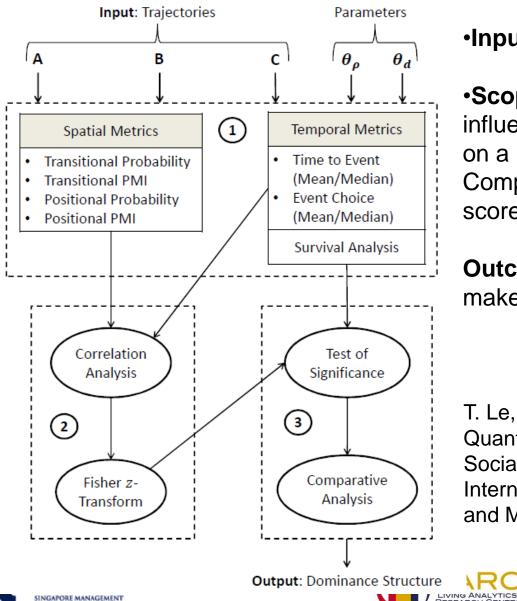
TABLE VI: Effect of change in mixture composition (ratio of $d_1 : d2$ observations) on queuing delay estimates

A. Aravamudhan, A. Misra and H. C. Lau. "Network-Theoretic" Queuing Delay Estimation in Theme Park Attractions. IEEE Conf. on Automation Science & Engineering, 2013.





Group Decision Analysis



•Input: Trajectories of different group types

•Scope of work: Compare each type's influence on the decision outcomes based on a number of spatio-temporal criteria. Comparisons are aggregated into a final score to make inference.

Outcome: Adult is the dominant decisionmaker in the Family groups

T. Le, S. Liu, H. C. Lau, and R. Krishnan. A Quantitative Analysis of Decision Process in Social Groups Using Human Trajectories. International Conference on Autonomous Agents and Multiagent Systems (AAMAS), 2014.

Carnegie

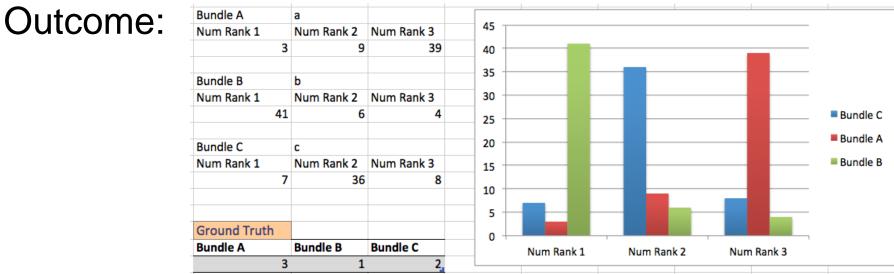
University

Mellonĭ

Bundle Design Optimization

INGAPORE MANAGEMENT

- Problem: Select *K* items from a set of *N* based on customer profile that maximizes Expected Revenue while keeping bundle attractive
- Stochastic Interacting Knapsack Problem: p_i and w_i are functions on the selected bundle



H. Nguyen, P. Varankantham, H. C. Lau and S. F. Cheng. Heuristics for Generalized Knapsack with Application to Theme Park Revenue Management. Metaheuristics International Conf. (MIC), 2013.

IVING ANALYTICS

Iniversity

Tools Developed

Agent-based Simulation Model



Visitor Tracking and Heatmap

LARC Participant Monitor					
	Controllers Get Distinct List of All Participants Get Last Known Positions of All Participants				
	Who will we focus and				
	Get Participant Activity History				
	Get Audit Log				
Data SNI Device ID Push Notification Token Latitude LongStude Time					
1 android@testing.com/d0/eeb9c.b519-44c0-9ee0-87/b51ee3c/4c[1255873]103.82209 [3/11/2014.5.34.14.PM	GPS Track Plot				
2 android@testing.com/d0feeb9c-b519-44c0-9ee0-87fb51ee3o4c[1.255873]103.82209 3/11/2014.5.35.08 PM					
3 android@testing.com/d6feeb9c.b519-44c0.9ee0.87fb51ee3d tc1 255873[103.82209 [3/11/2014 5.36.09 PM 4 android@testing.com/d6feeb9c.b519.44c0.9ee0.87fb51ee3d tc1 255873[103.82209 [3/11/2014 5.37.08 PM					
4 android@testing.com/d8feeb9c.b519-44c0.9ee0.87fb51ee3d/cc 1.255873 103.82209 3/11/2014.5.37.08 PM 5 android@testing.com/d8feeb9c.b519-44c0.9ee0.87fb51ee3d/cc 1.255873 103.82209 3/11/2014.5.38.10 PM					
6 android@testing.com/d0feeb9c-b519-44c0-9ee0-87b51ee3c4c1 255873 103.82209 3/11/2014 5.39.08 PM					
7 android@testing.com/d6feeb9c-b519-44c0-9ee0-87fb51ee3c/cc11.255873 103.82209 3/11/2014 5:40.09 PM	GPS Location Data				
8 android@testing.com/d8teeb9c.b519-44c0.9ee0.87tb51ee3ctc 1 255873 103 82209 3/11/2014 5:41:09 PM					
9 android@testing.com/d6/eeb9c.b519.44c0.9ee0.87b51ee3c4c1255873103.82209 3/11/2014.5.42.09 PM	of Visitor				
10 android@testing.com/d6feeb9c-b519-44c0-9ee0-87fb51ee3c4c[1.255873]103.82209 [3/11/2014.5.43.09 PM					
11 android@testing.com/d0feeb9c.b519-44c0-9ee0-87fb51ee3c1c[1.255873]103.82209 3/11/2014 5.44.09 PM					
12 android@testing.com/d6feeb9c.b519-44c0.9ee0-87fb51ee3c4c 1 255873 103.82209 3/11/2014 5 45 09 PM					
13 android@testing.com/d6feeb9c-b519-44c0-9ee0-87tb51ee3o4c[1.255873]103.82209 3/11/2014 5/46:09 PM					
14 android@testing.com.d0/eeb9c-b519-44c0-9ee0-87tb51ee3d4c 1 255873 103.82209 3/11/2014 5 47.09 PM					
15 android@testing.com d8leeb9c.b519.44c0.9ee0.87lb51ee3c1c 1.255448 103.82227 3/11/2014.5.48.09 PM					
16 android@testing.com d6feeb9c.b519-44c0-9ee0-87lb51ee3d/c 1.254961 103.822132 3/11/2014 5.49.09 PM 17 android@testing.com d6feeb9c.b519-44c0-9ee0-87lb51ee3d/c 1.254961 103.82207 3/11/2014 5.50.09 PM					
[17] Jandrologgeesing.com/dowedac-bb19-44c0-9eeu-87bb1ee3d62[1,254188]103.82207 [3/11/2014 5:50:09 PM]					





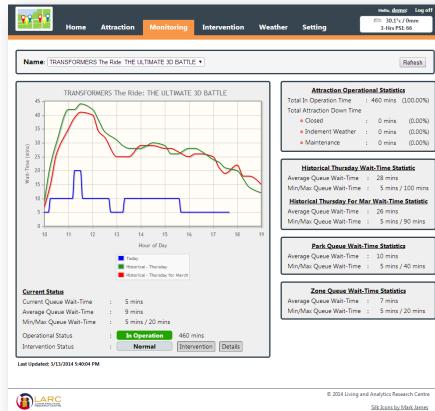




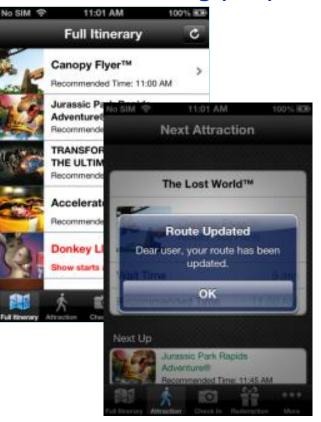


Tools Developed

Attraction Queue Monitoring & Intervention Portal (Web)



Personalized Dynamic Route Planning (iOS)



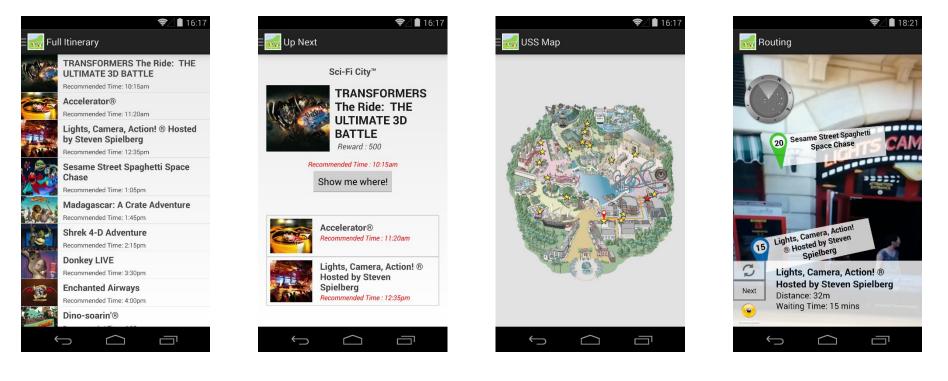






Tools Developed

Theme Park Navigation with Augmented Reality (Android)









Questions and Comments

Hoong Chuin LAU Professor of Information Systems Deputy Director, Living Analytics Research Center Singapore Management University http://www.mysmu.edu/faculty/hclau





