



# Deep Learning for Ancillary Pricing

Kartik Yellepeddi (deepair), Prof. Lavanya Marla (UIUC), Naman Shukla (UIUC)

*2019 AGIFORS RM Study Group, Panama City, Panama*

# “Deep Learning is here. Own what happens next.”

01

Research Objectives

02

Ancillary Pricing Use Case

03

Key Learnings and Next Steps

04

Our broader vision



# FROM **AVERAGE** TO **INDIVIDUAL** DECISIONS



# Research Objective

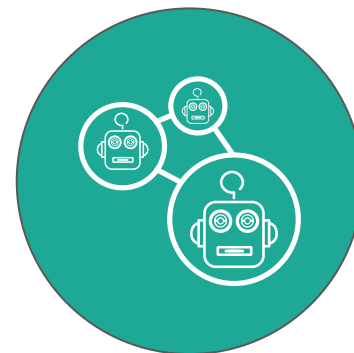
---



Can deep learning  
make better  
predictions at an  
Individual level



Can those  
predictions be  
converted to  
better Individual  
decisions



What happens when  
many AI agents make  
decisions in the same  
environment

# Ancillary Pricing

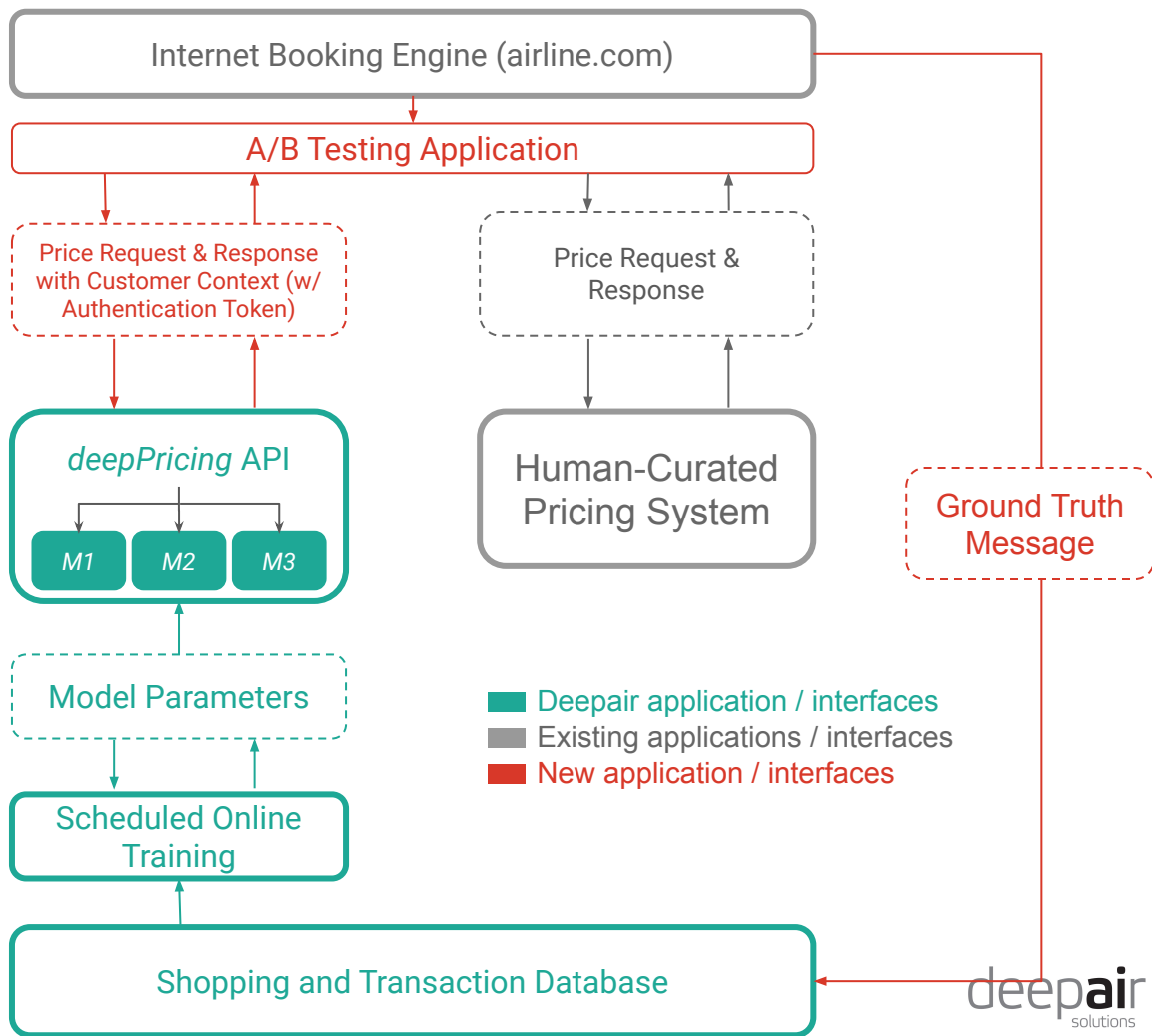
---

Ancillaries complement RM and are perfect to measure the revenue opportunity of moving from “Average to Individual”

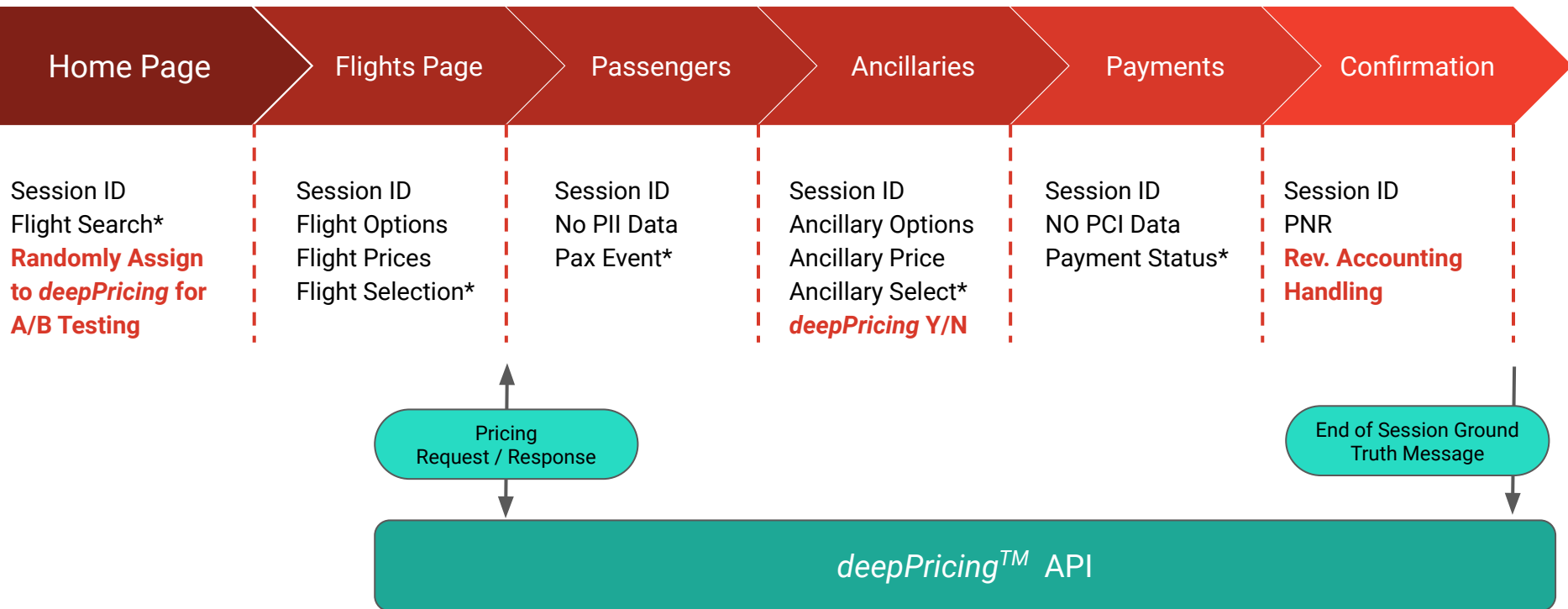
- ✓ Complementary to Right-to-fly
- ✓ Not Optimized at Most Airlines
- ✓ Deeply Personal and Attribute Driven Sale
- ✓ Ancillaries Compete for Wallet Share and Shelf-Space
- ✓ Influenced by Price and Placement (start with price)

# High Level Set-up

Built an ML experimentation platform from the ground up



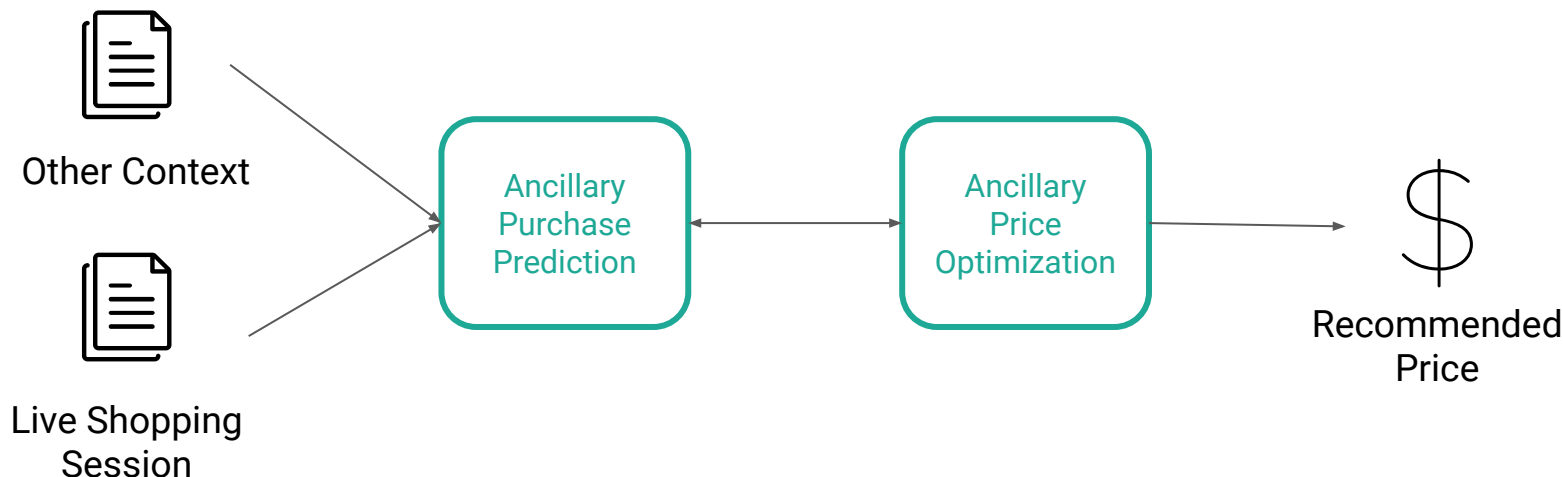
# Functional View on airline.com



\* represents the event that triggers transition to next step in the booking process

# A Two-Stage Dynamic Pricing Model

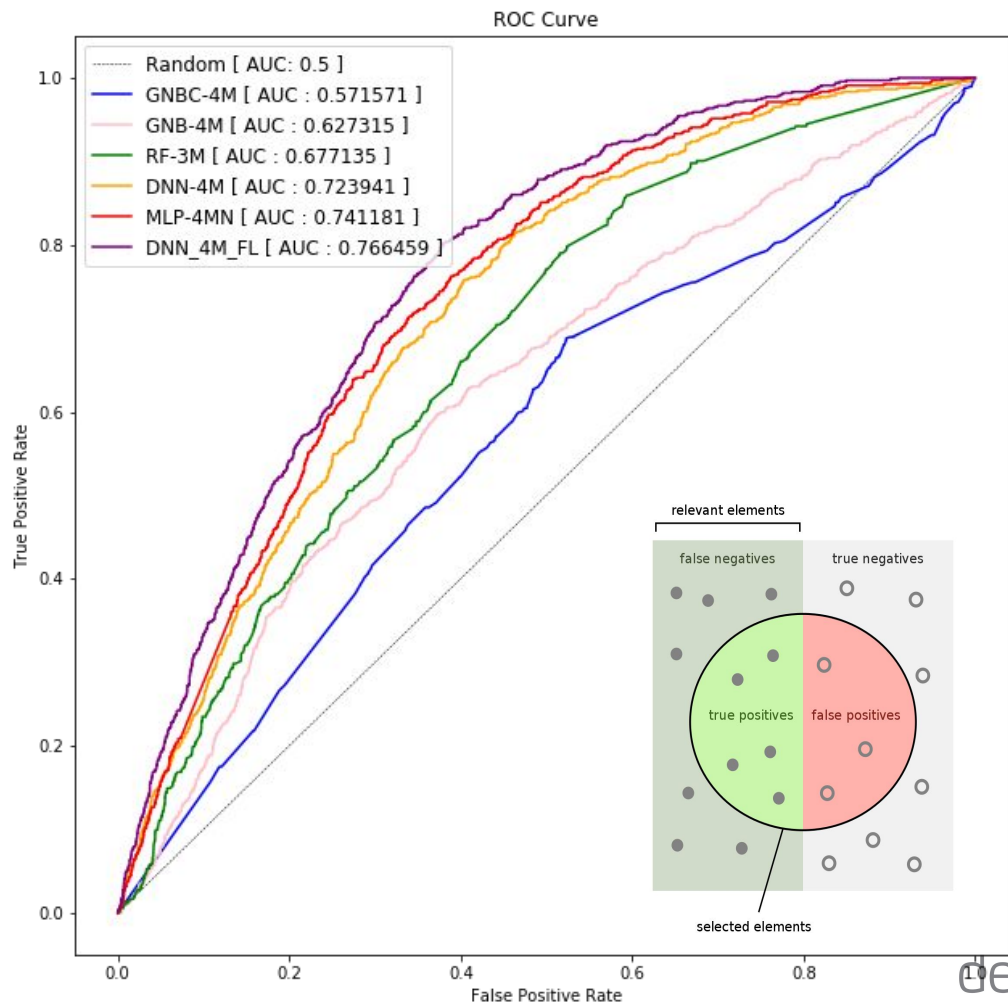
---





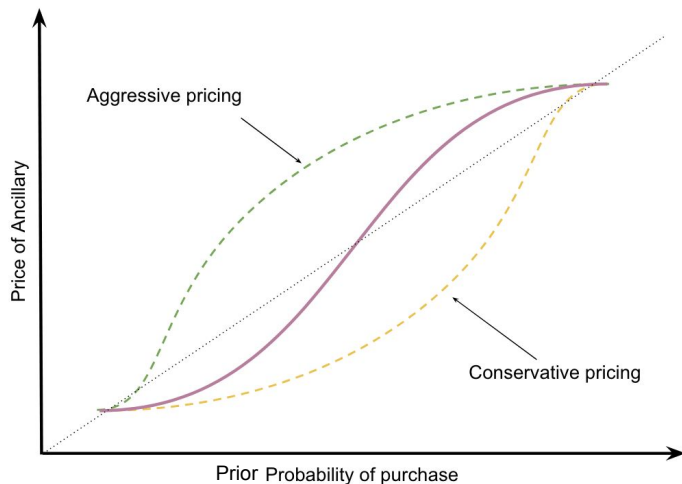
# Ancillary Purchase Prediction

Deep Neural Network  
beats traditional models

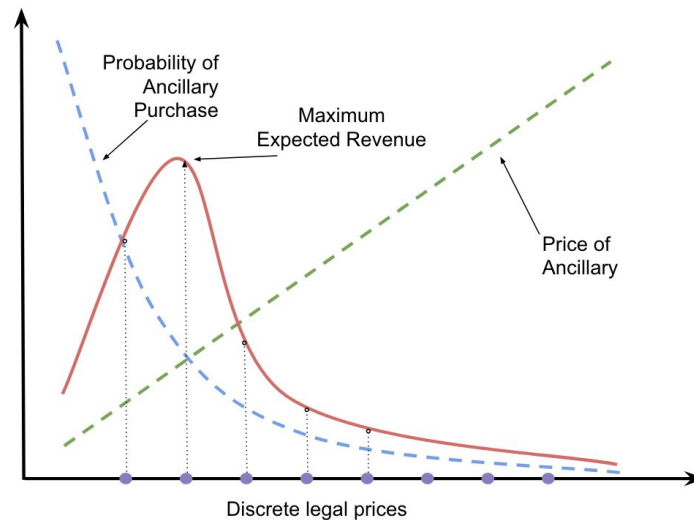


# Ancillary Price Optimization

## Logistic Pricing

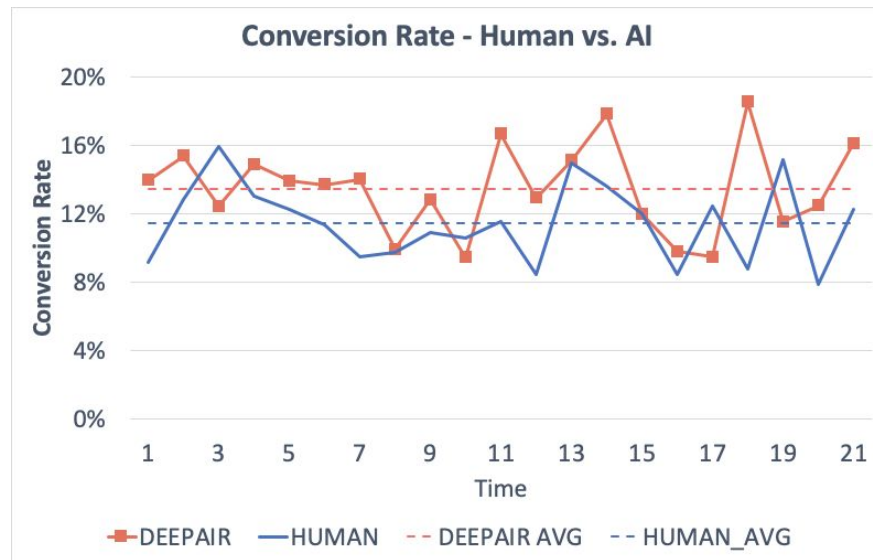
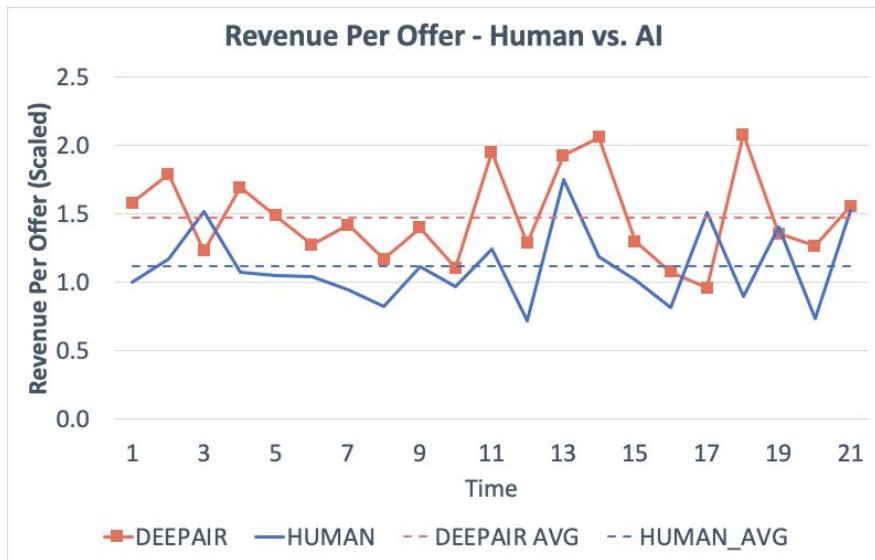


## Discrete Exhaustive Search



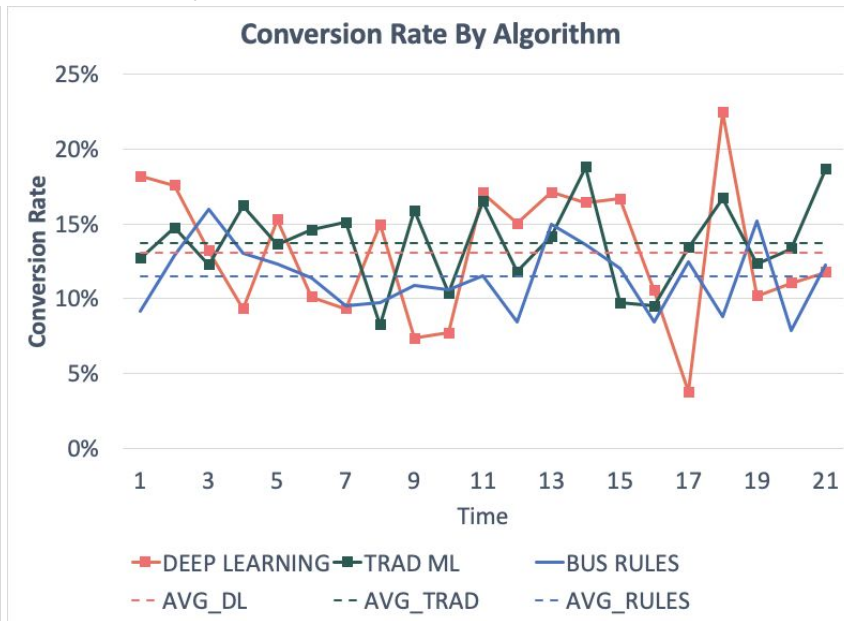
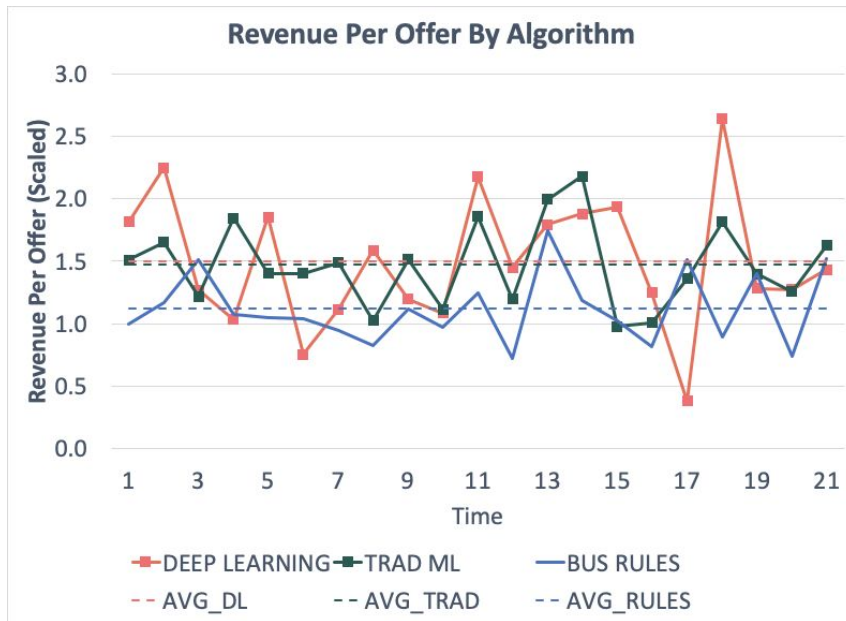
# Human vs. AI Performance

**Summary:** AI driven pricing resulted in 25% incremental Revenue Per Offer (p-value 0.000297) and 17% higher Conversion compared to Human, just with a discounting strategy.



# Algorithm Level Performance

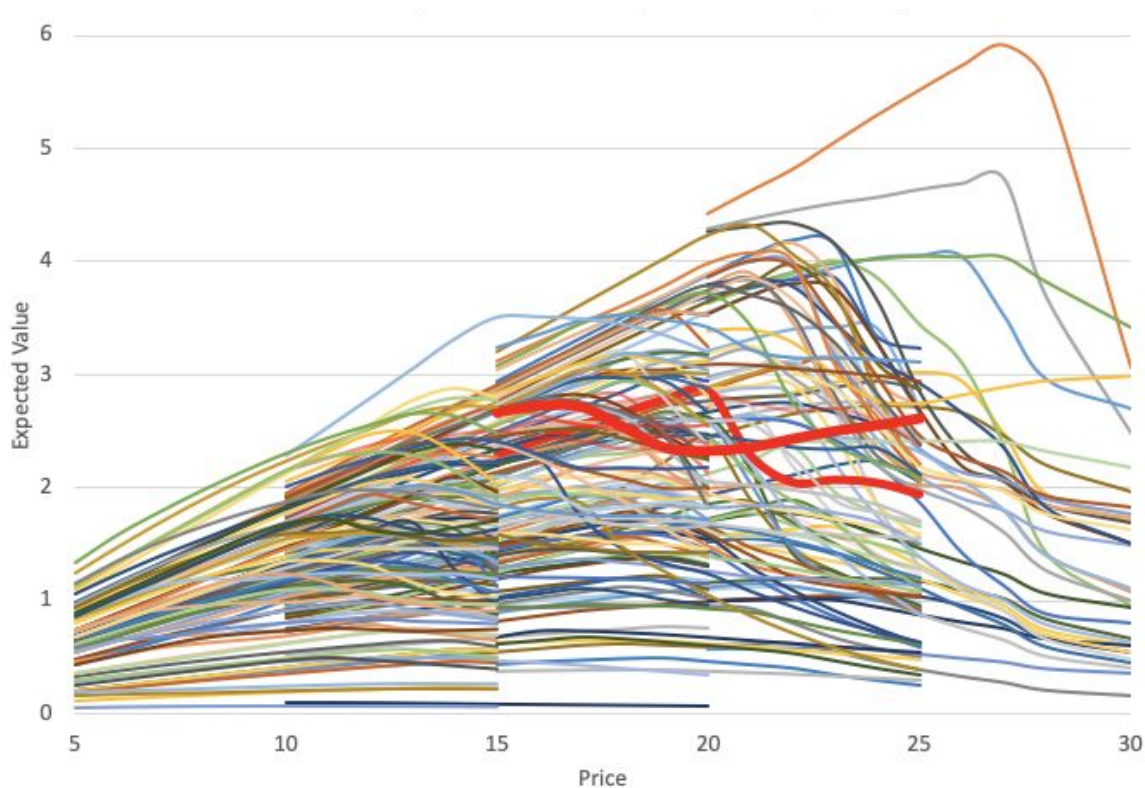
**Summary:** Although deep learning is 35% more accurate in predicting the probability of purchase compared to Traditional ML, this isn't resulting in proportionally better online performance.



Traditional vs. Rules (p-value): 0.000376; Deep Learning vs. Rules (p-value): 0.00360

# Price Sensitivity

Neural Network did not learn the monotonic relationship between price and probability



# Next Steps

---

AI Driven Pricing show  
revenue potential

- ✓ Explicitly “Teach” the Neural Network Price-Demand Relationship
- ✓ Single Stage Optimization To Reduce Error
- ✓ Meta Learning AI that explores and exploits various AI algorithms
- ✓ More ancillaries and more touch points to research multi-agent environments

# The Big Picture

---

Delighting travelers at scale by moving from average to individual  
Decisions





“In an Experience Economy, businesses will find that the next competitive battleground lies in staging experiences.”  
- HBR



# Revenue Optimization Redefined

Imagine 100s of services and revenue sources to choose from, to curate a truly individualized experience

Traveller Experience

## AI Driven Retailing



Pricing



Content Discovery



Contracts

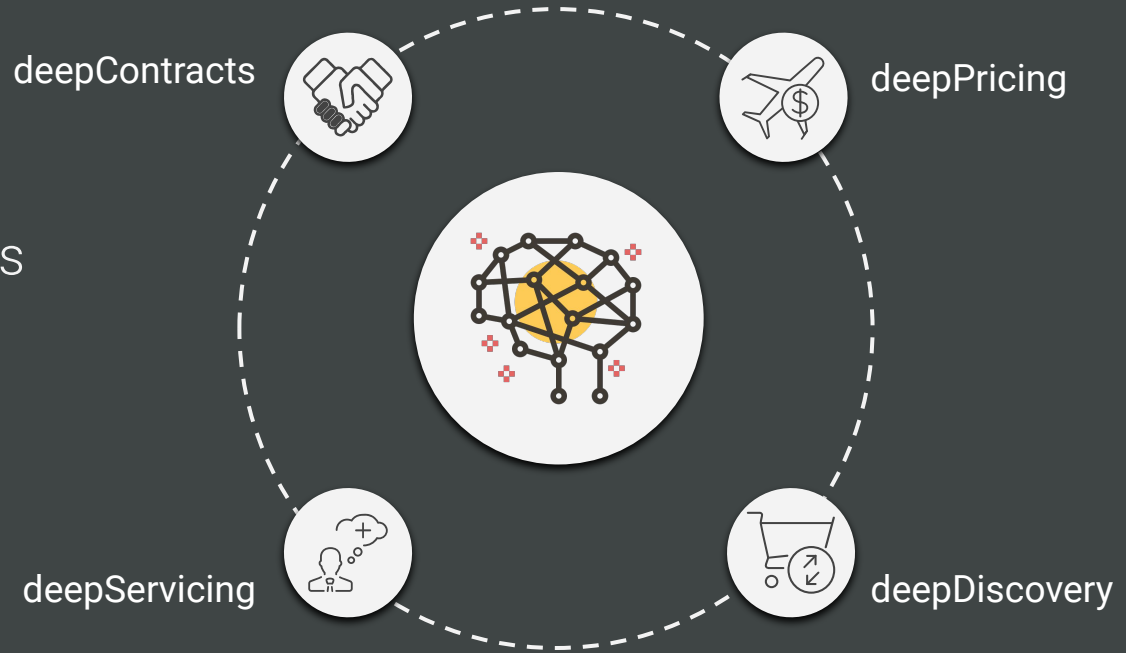


Service



Settlement and Distribution Platform

## Four Deep Learning Agents Working in Harmony



Join us at KDD 2019  
Anchorage, AS  
Aug 4 - Aug 8

deepair  
solutions



# GET SOCIAL WITH US



@deepair\_io



[linkedin.com/company/deepair](https://linkedin.com/company/deepair)