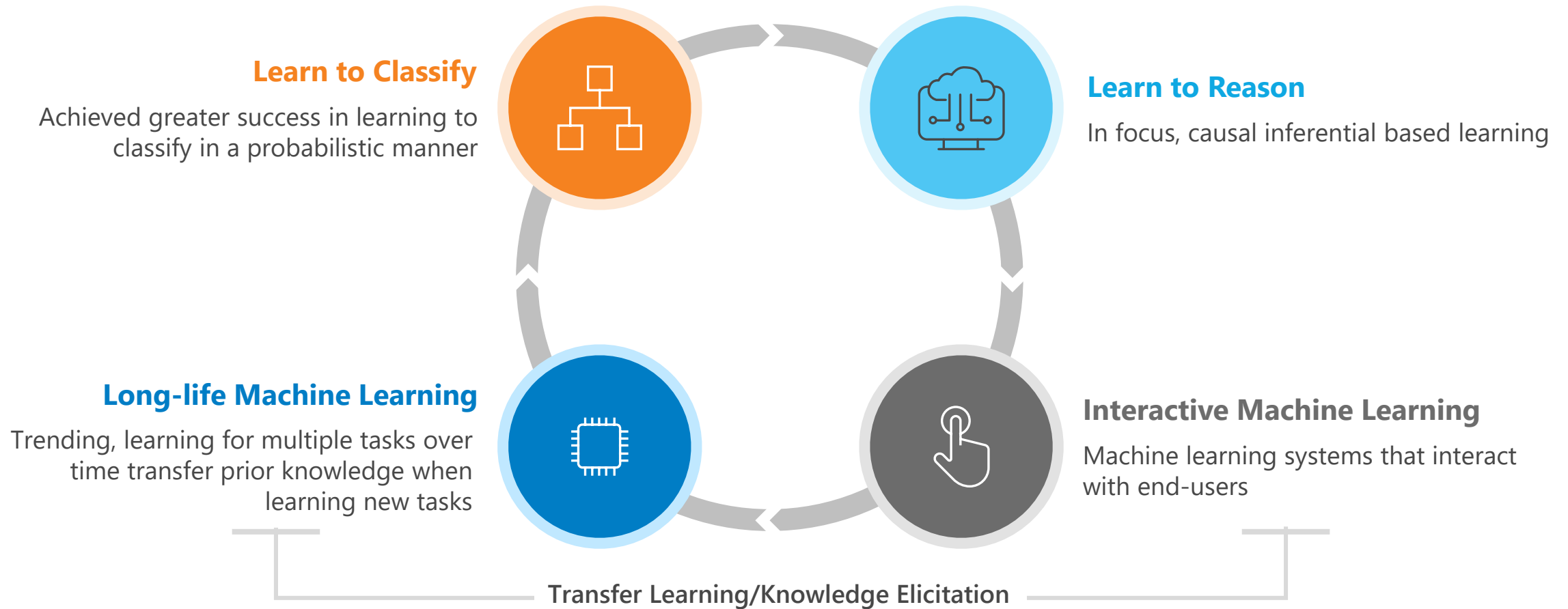


# Knowledge Elicitation and its applications

By Zabi Ulla, Sr. Director Advanced Analytics



# Industry Demand in Machine Learning Space



# Outline



Introduction to Knowledge Elicitation in a machine learning framework



How we are leveraging these concepts in our products and solutions



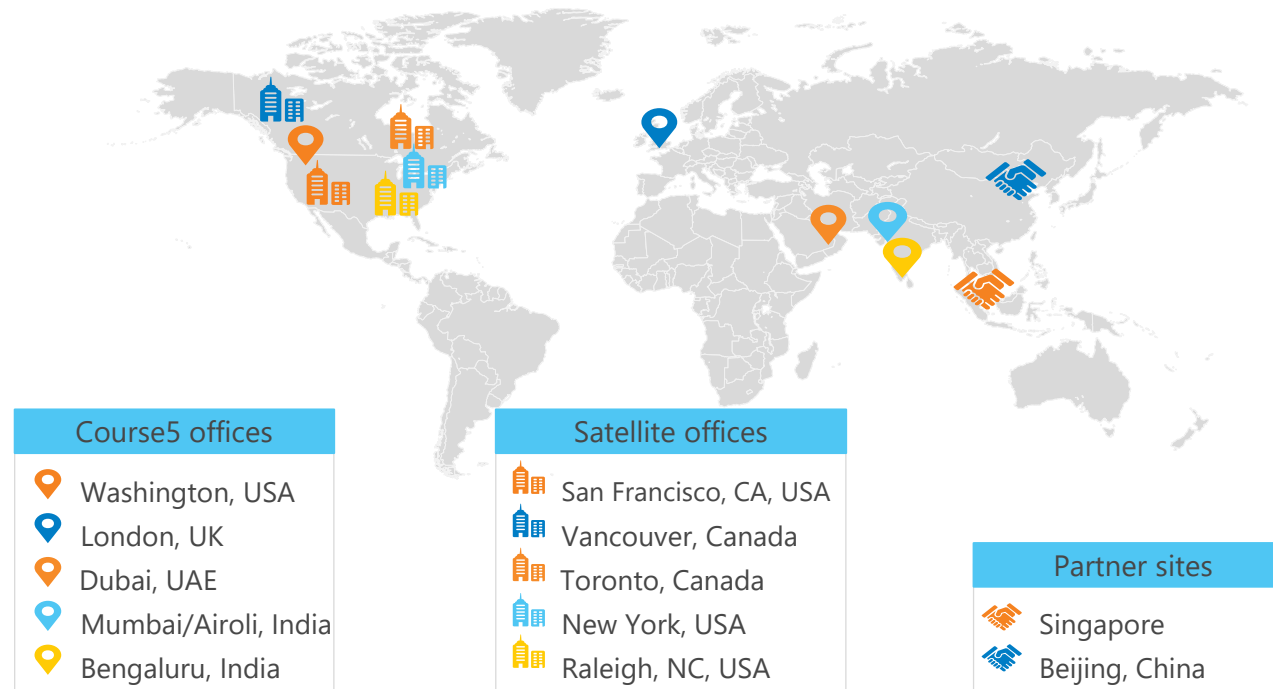
Course5 application of Knowledge Elicitation

- Transfer Learning in Bayesian Network Modeling
- Course5 Integrated Marketing Measurement & Optimization

# About Course5 Intelligence

We drive **Digital Transformation** for businesses through **Analytics, Insights, and Artificial Intelligence**.

**Multi-faceted talent** with 1200 personnel across global locations enables our clients to make the most effective strategic and tactical moves related to **customers, markets, and competition**.



## OUR IMPACT

- \$10bn+ impact for client organizations to date
- Real time understanding of customers & markets and enablement of digital business models
- 360 analysis, providing deep insights and decision-making support
- End-to-end solutions – delivering top line and bottom line business impact

# Knowledge Elicitation



A concept which leverage human expertise into a machine learning process



To account of **aspects** which are not feasible to measure and quantify into a mathematical models



Opens up new opportunities by interpreting, otherwise black box machine learning process



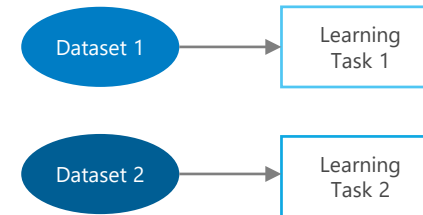
# Transfer learning

It is desirable to use data and knowledge from similar cases; a technique known as “transfer learning”

- C++ > Java
- Maths/Physics > Computer Science/Economics

## Current ML

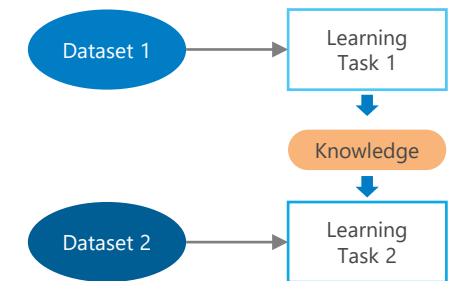
Isolated, single task learning:



## VS

## Transfer Learning

Learning of new tasks relies on the previous learned tasks



## What

which **part of the knowledge** can be transferred from the source model to the new model

## When

To improve new task performance/results and **not degrade them**.

## How

Identifying ways of actually transferring the knowledge across domains/tasks

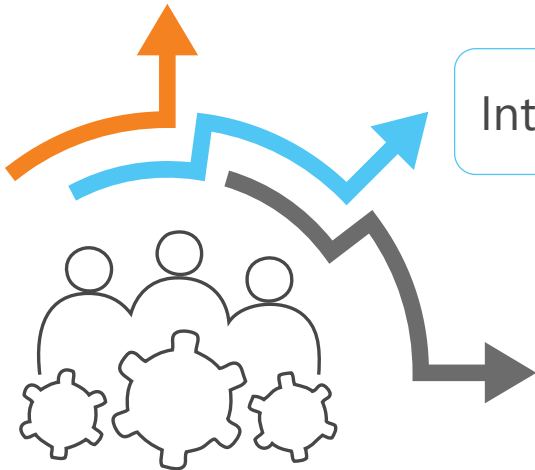
# How we are adopting knowledge elicitation and transfer learning

We are building one type of model (source domain) and transferring knowledge gained into another type of model (target tasks) to

Account for unobserved variables (nodes) which are not feasible to be observed (instance transfer learning)

Introduce right interventions in the model to measure “true” impact of variables

To allow counterfactual inference which otherwise not feasible

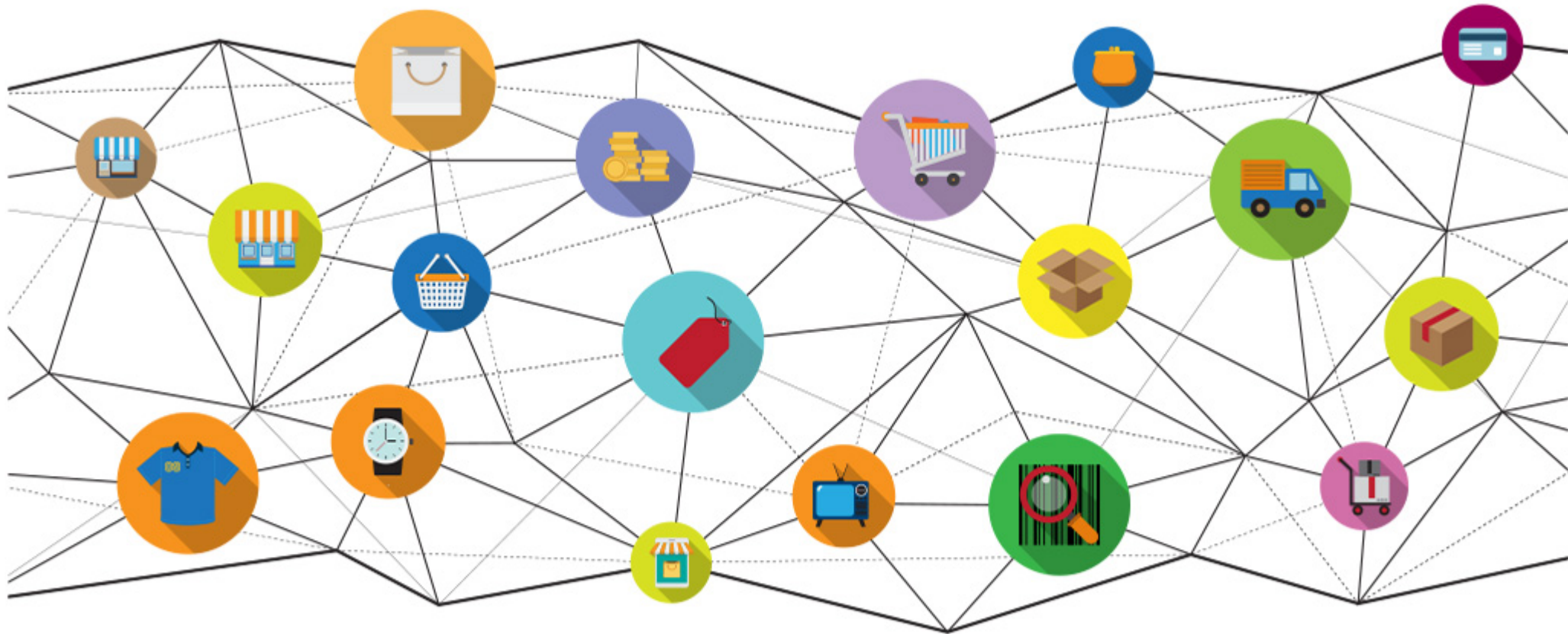




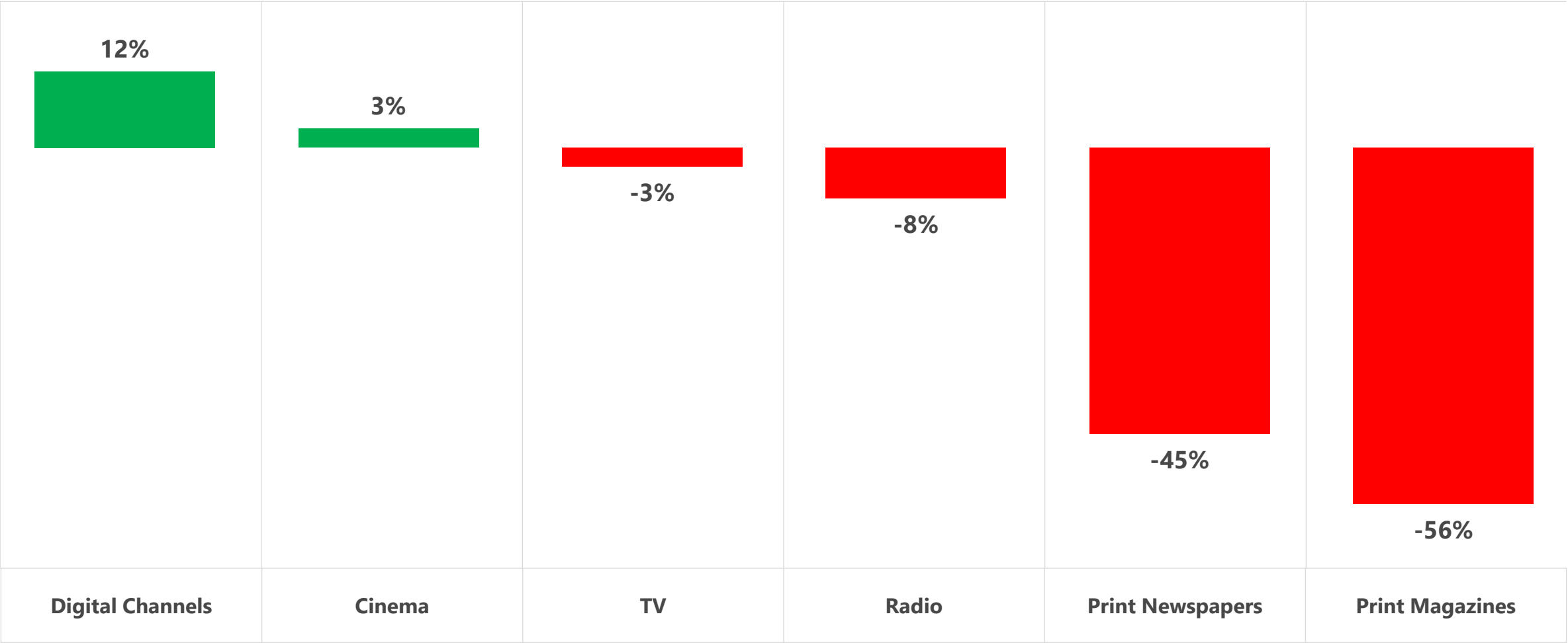
# Applications of marketing measurement



# Consumer Purchase Journeys in Digital World

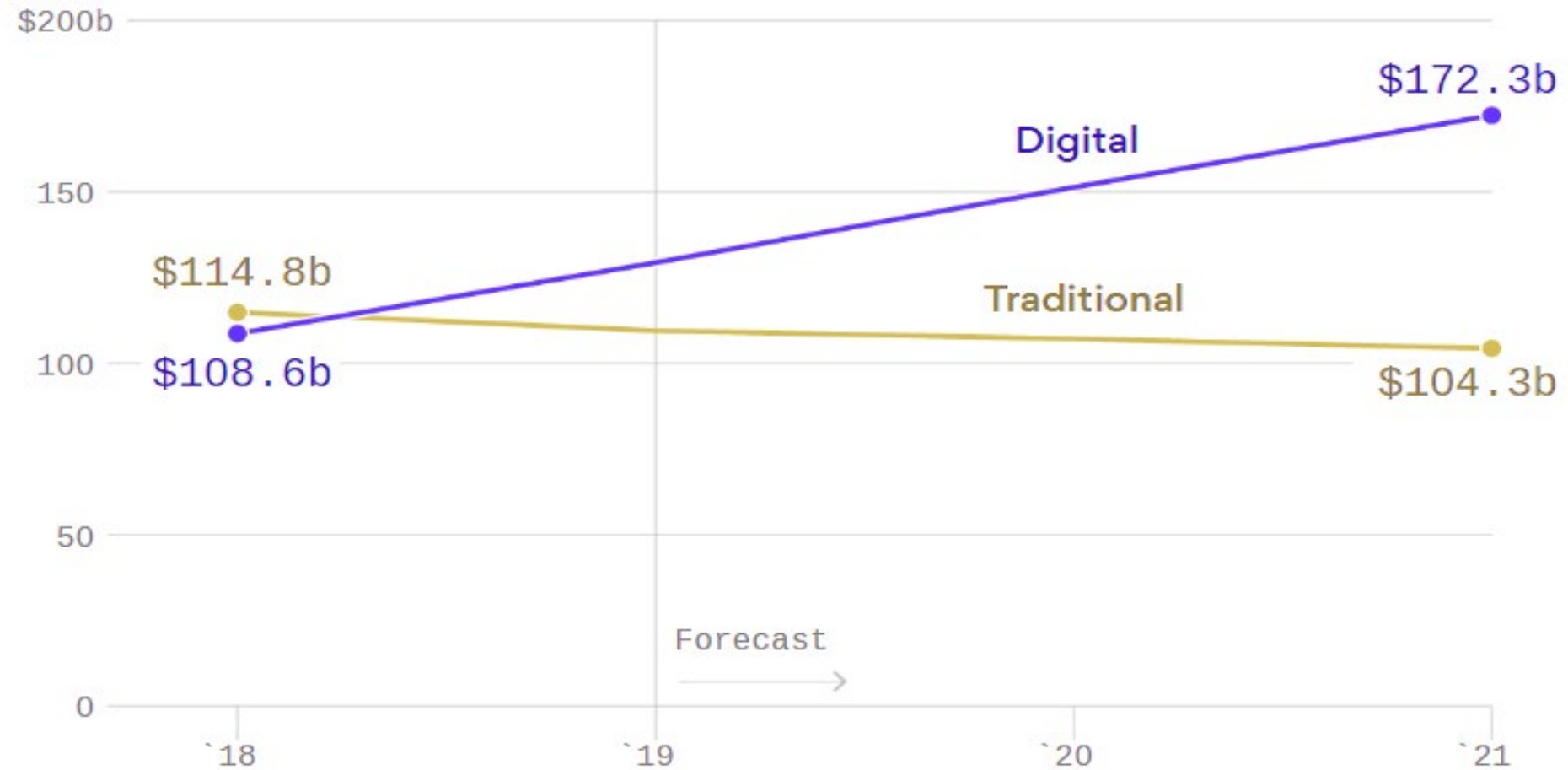


# Change in Traditional Media Consumption 2011-2018



Published on Marketing Charts.com in June 2018

# U.S. Digital vs. Traditional annual AD Spending



Adapted from eMarket; Note: Forecast as of February 2019; Chart: Andrew Witherspoon/Axiosv

# Current Marketing Measurement Models

Top-down

## Marketing Mix

Useful for the strategic planning by brand, geo etc., budget allocation and optimal channel mix



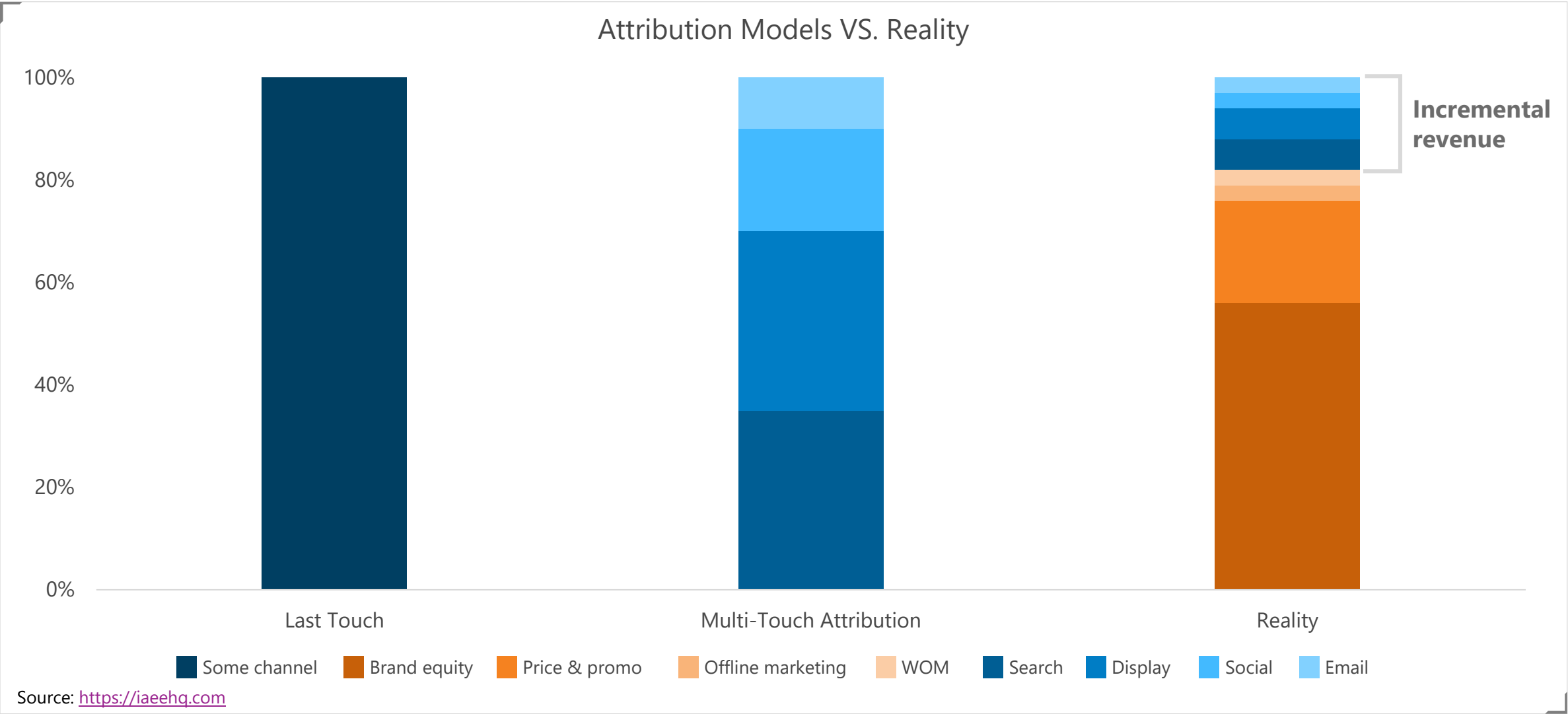
Tactical media buying, campaign execution in right channel and optimization

Bottom-up

## Digital Attribution



# Incorrect digital media measurement



# In Summary

## Key Challenges



Complex consumer journeys



Industry is biased towards digital channels



Disparate data sources



Models are separate and used in silos



Incorrect attribution and declining ROI



Enables dynamic campaign effectiveness and align with overall marketing strategy

Course5 Integrated Marketing Measurement Solution is a **continuously evolving marketing solution**, that brings a top-down and bottom-up meta intelligence to marketing insights.

Aggregate static data model and individual user level configurable models

Doesn't provide granular insights at customer groups

MMM

Historical aggregate spend patterns and correlate with sales over time

Doesn't account offline channels hence biased

Digital Attribution

Individual customer journeys (digital path) lead to conversion mostly real time data

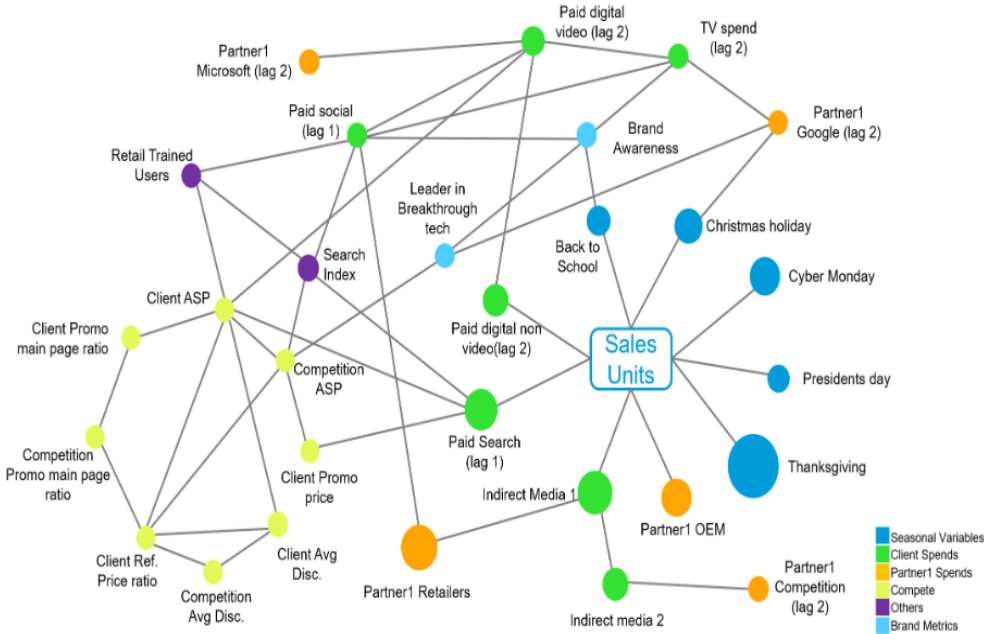




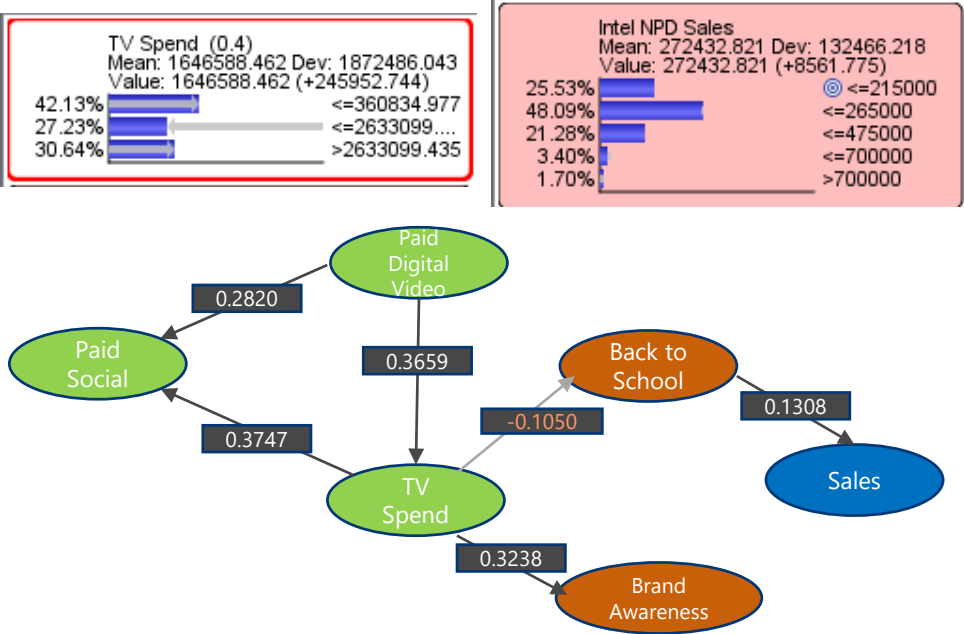
# How we are integrating best of both?

# Build Holistic MMO using Bayesian Network Modelling

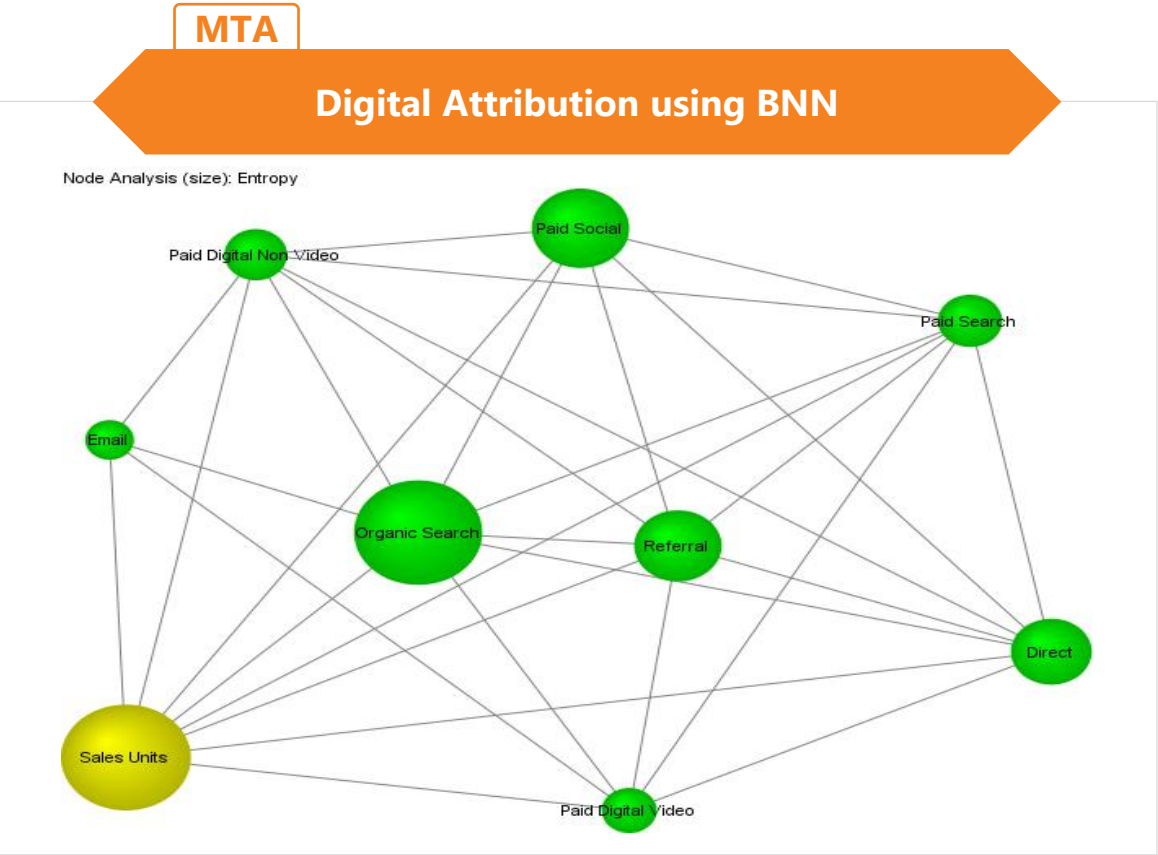
## Structural Learning using BNN



## Knowledge discovery from BNN



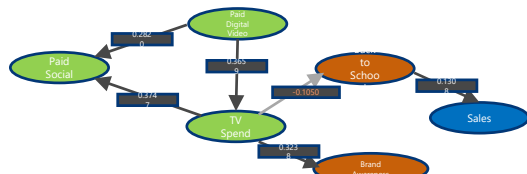
# Build Digital Attribution Network model



Channel	Contribution
Organic Search	59.9%
Referral	16.8%
Email	7.1%
Direct	5.8%
Paid Social	4.4%
Paid Digital Video	4.2%
Paid Digital Non Video	1.1%
Paid Search	0.6%

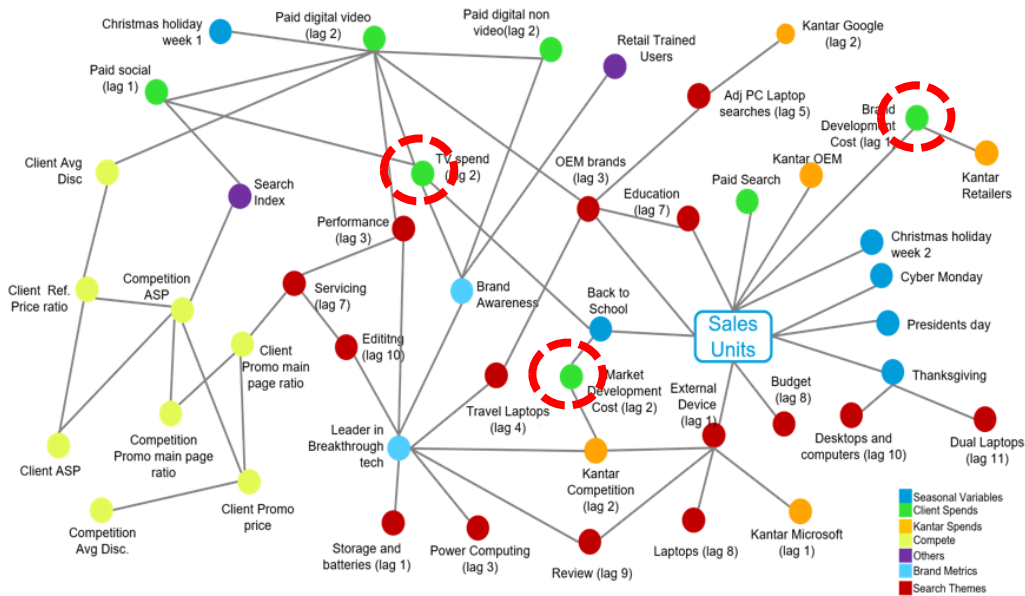
# Build Attribution network model (including offline channel nodes) with the helps of a prior probabilities specified in network

## Knowledge discovery from BNN



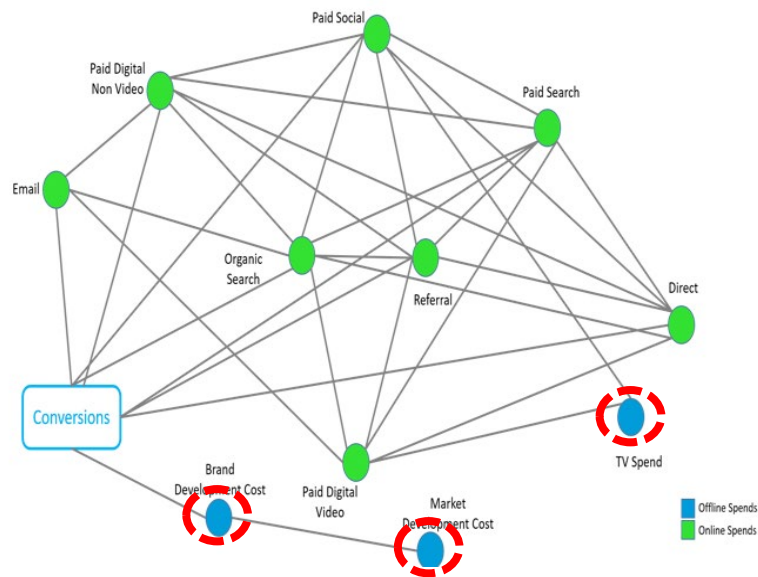
MMO

## Structural Learning using BNN



IMM

The Network Represents how sales is interrelated with online as well as offline channels



# Relative importance of offline and online channels in IMM

Channel	Digital Attribution (standard)	True Contribution (IMM)	Inflation in standard models
Organic Search	59.9%	28.24%	↓ -31.66%
Referral	16.8%	7.60%	↓ -9.20%
Email	7.1%	3.40%	↓ -3.70%
Direct	5.8%	2.67%	↓ -2.67%
Paid Social	4.4%	2.04%	↓ -2.36%
Paid Digital Video	4.2%	1.35%	↓ -2.85%
Paid Digital Non Video	1.1%	0.47%	↓ -0.63%
Paid Search	0.6%	0.16%	↓ -0.44%

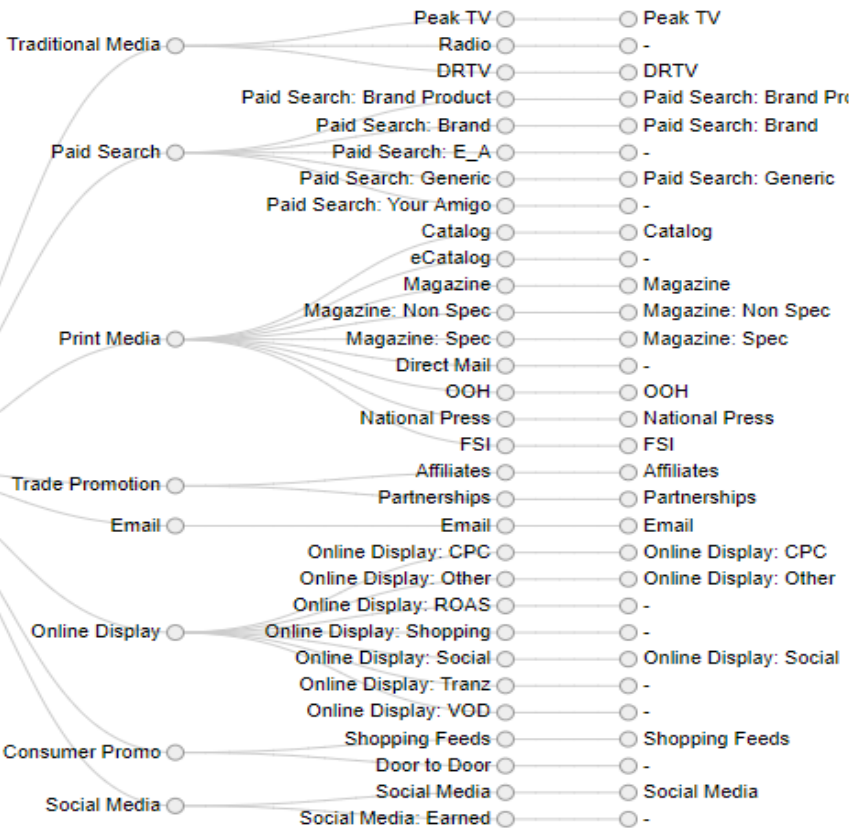


# Key Benefits across industry applications

## Better ROI



## Purchase Path



# In Summary

## Key Challenges



- ☐ Complex consumer journeys
- ☐ Industry is biased towards digital channels
- ☐ Disparate data sources
- ☐ Models are separate and built in silos
- ☐ Incorrect attribution and declining ROI


## Key Benefits of IMM



- ☐ Unified measurement provides holistic insights
- ☐ Able to handle disparate data very well
- ☐ Learnings from one model is transferred into another model
- ☐ Accurate attribution accounting for mediating variables
- ☐ The model insights are actionable in real time campaign management

# Integrated Marketing Measurement platform demo


[https://course5intelligence.shinyapps.io/integrated\\_marketing\\_mix/](https://course5intelligence.shinyapps.io/integrated_marketing_mix/)



## Integrated Market Measurement Prototype

[Introduction](#)[Trend Analysis](#)[Marketing Response Analysis](#)[Market Mix Model](#)[Digital Spend Optimization](#)[Integrated Marketing Mix](#)

[Our Vision](#)[Methodology](#)



Course5 Integrated Marketing Measurement Solution is a continuously evolving marketing solution, that brings a top-down and bottom-up meta intelligence to marketing insights through a blend of technology intervention, frameworks, critical thinking, and advanced data science techniques.

**“ Our IMM solution leverages Marketing Mix Modeling (MMM) and Multi-Touch Attribution (MTA) to bring together aggregate full-business models, and user-level models for a customer view”**

### Why consider a merger?

#### Multitouch Attribution Modeling

Data-driven attribution requires a wealth of granular, user-level data, which can limit offline channel visibility

+

#### Market Mix Modeling


Traditional MMM offers high-level analysis on a quarterly or yearly basis, which can limit more granular, or on-the-fly optimization

=


#### Integrating MTA & MMM

When the two measurement practices are combined, however, they improve the outputs from each. Data-driven attribution informs MMM models. MMM data feeds attribution analysis.


### How does implementing IMM help?




Actionable Real-Time Data



Omnichannel Delivery



Unified Marketing



Adapt to fluctuations

Course5  
Transformative Intelligence

© Course5 Intelligence

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“  
The surest kind of knowledge is what you construct  
yourself.

*~ Judea Pearl, The Book of Why: The New Science of Cause and Effect*

”





THANK YOU