

The image features a dark gray background with decorative geometric shapes. In the top-left corner, there is a cluster of small, light gray and blue triangles. In the bottom-right corner, there is a larger cluster of light gray triangles, some of which are slightly larger and more prominent than the others. The text "ANALYZING BIG DATA" is centered in the middle of the image in a white, bold, sans-serif font.

ANALYZING BIG DATA

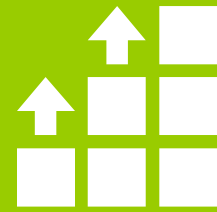
WHAT IS BIG DATA?



Data Complexity: Variety and Velocity

WHAT IS BIG DATA AND WHY NOW?

Volume
Variety
Velocity



Data
explosion



Changing
economics

Hadoop



Cheap, Distributed
Storage & Processing

“

By 2015, organizations that build a modern information management system will outperform their peers financially by 20 percent.

– Gartner, Mark Beyer, “Information Management in the 21st Century”

”

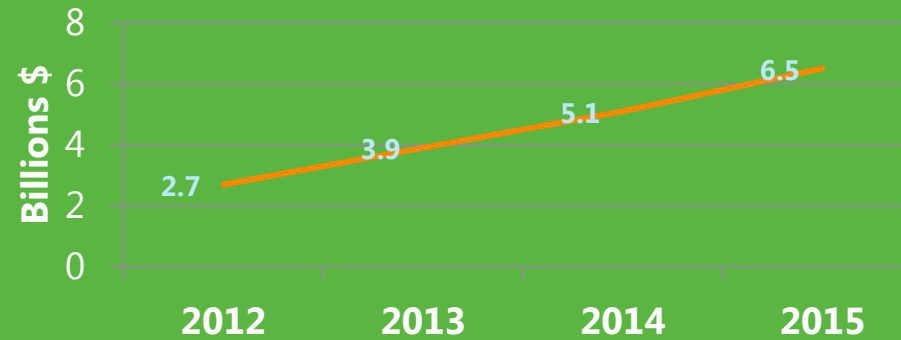
BIG DATA IS A GROWTH OPPORTUNITY FOR CLIENTS

Big Data is a Big Priority for Clients



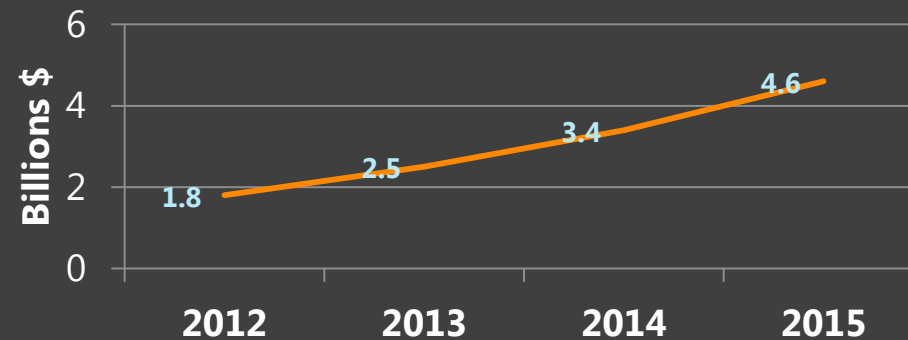
49% of top CEOs and CIOs are currently using Big Data for customer analytics¹

Big Data Services Growth



39% compound annual growth rate²

Big Data Software Growth



34% compound annual growth rate²

1. McKinsey&Company, McKinsey Global Survey Results, Minding Your Digital Business, 2012
2. IDC Market Analysis, Worldwide Big Data Technology and Services 2012–2015 Forecast , 2012

A NEW SET OF QUESTIONS

What's the market sentiment for my brand or products



How do I optimize my target group based on demographics and health patterns?

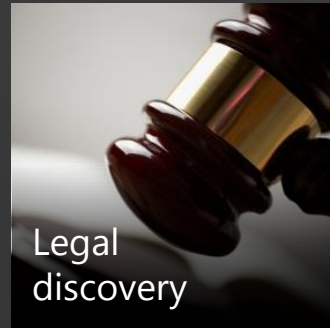
How do I better predict future outcomes?



NEW OPPORTUNITIES



IT infrastructure optimization



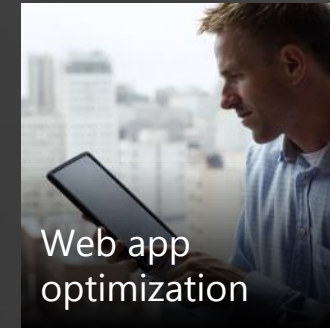
Legal discovery



Social network analysis



Target group optimization



Web app optimization



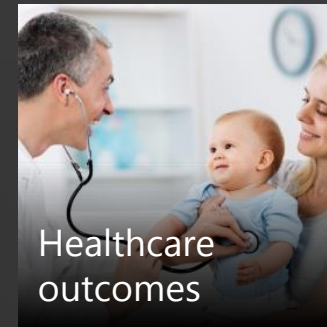
Churn analysis



Natural resource exploration



Weather forecasting



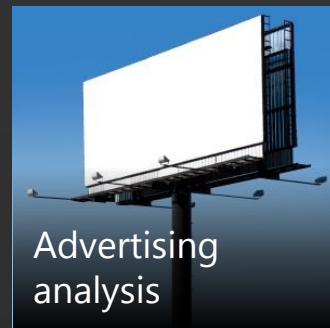
Healthcare outcomes



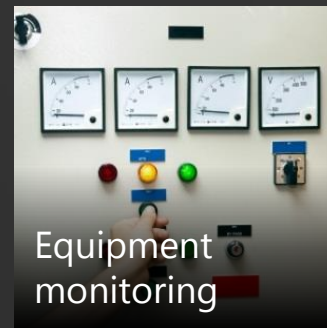
Fraud detection



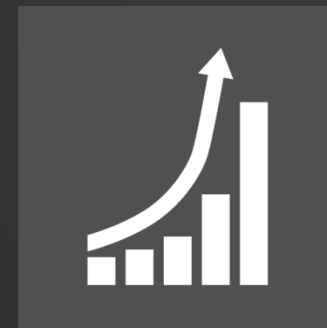
Life sciences research



Advertising analysis

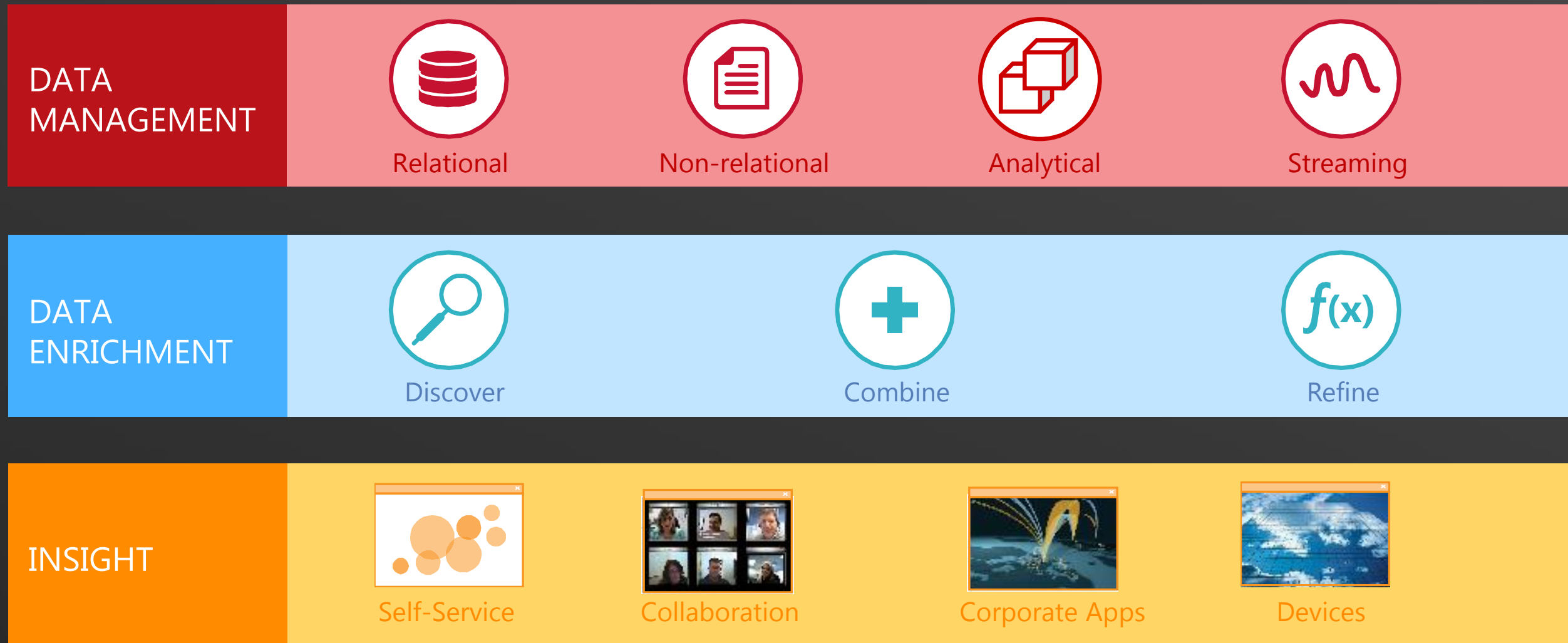


Equipment monitoring

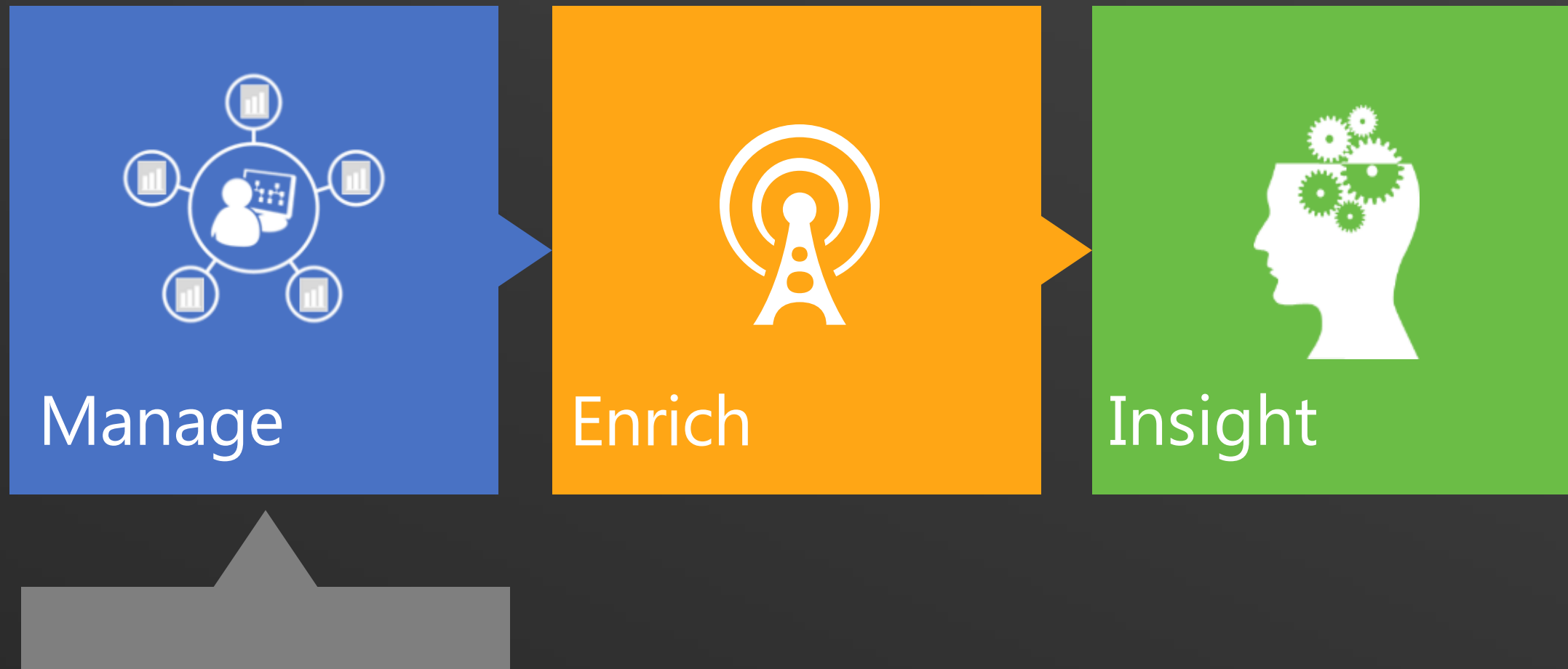


Smart meter monitoring

BIG DATA REQUIRES AN END-TO-END APPROACH



THE BIG DATA LIFECYCLE



MANAGE ANY DATA, ANY SIZE, ANYWHERE

e.g. STRUCTURED & UNSTRUCTURED DATA

Unified Monitoring, Management & Security



Relational



Non-Relational

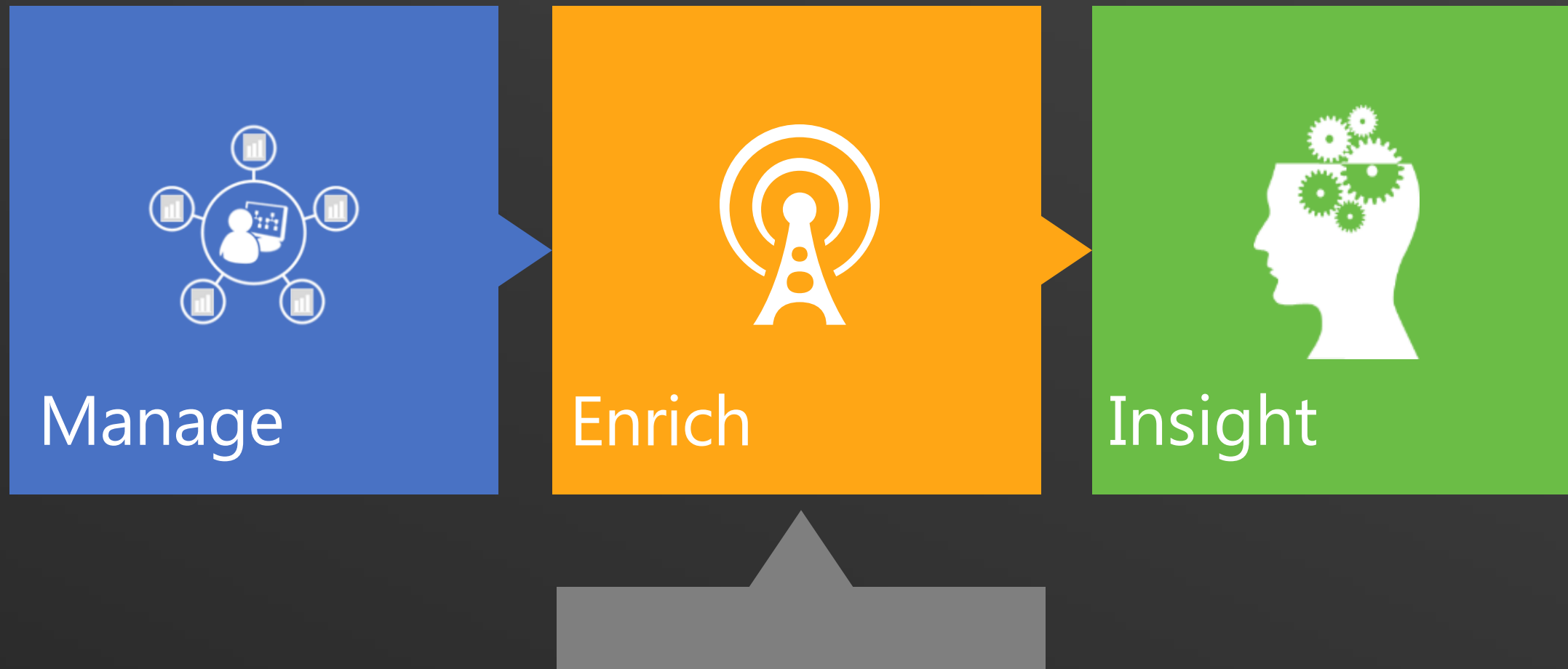


Streaming

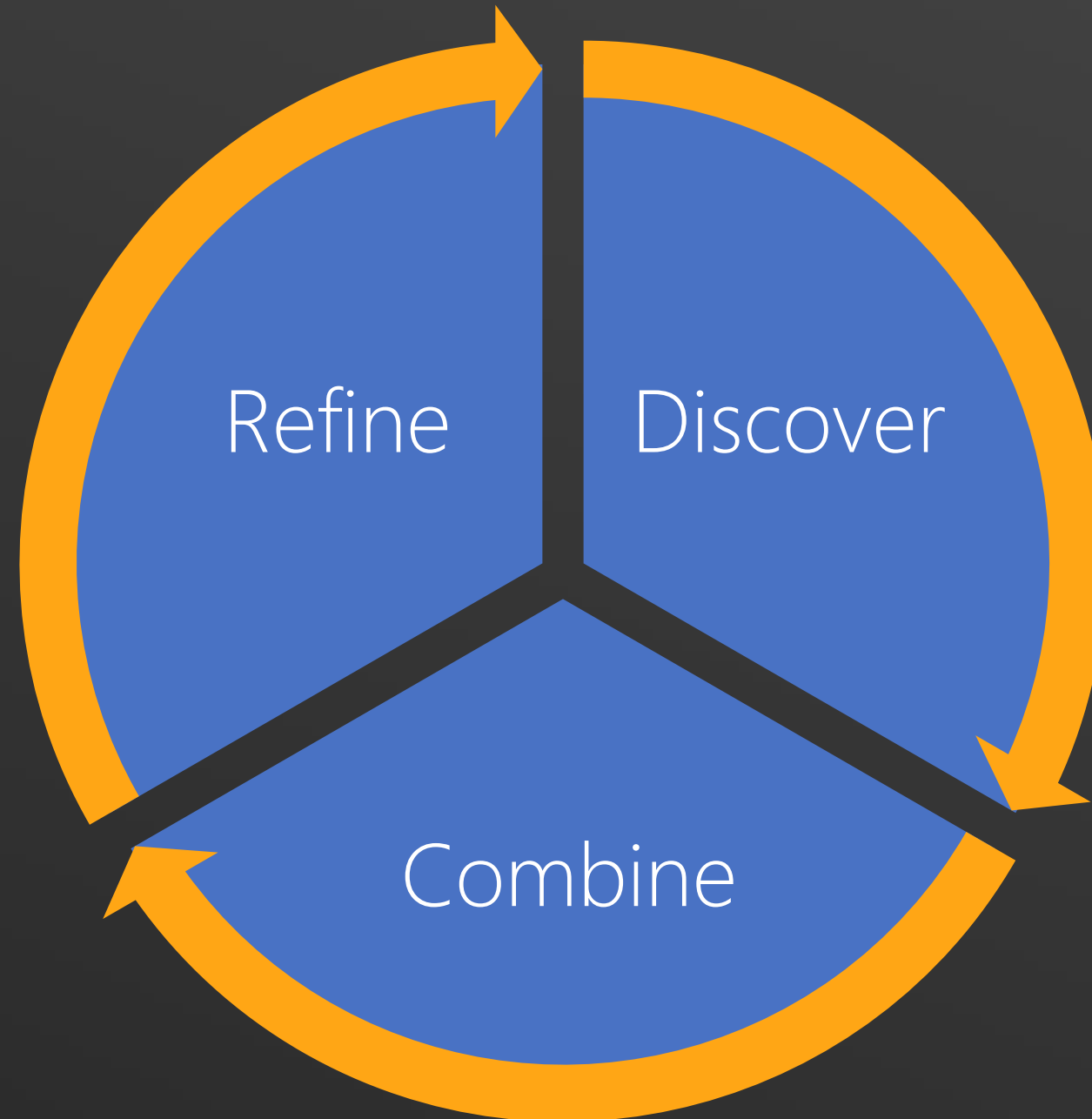


Data Movement

THE BIG DATA LIFECYCLE



ENRICH BY CONNECTING TO THE WORLDS DATA



DISCOVER DATA

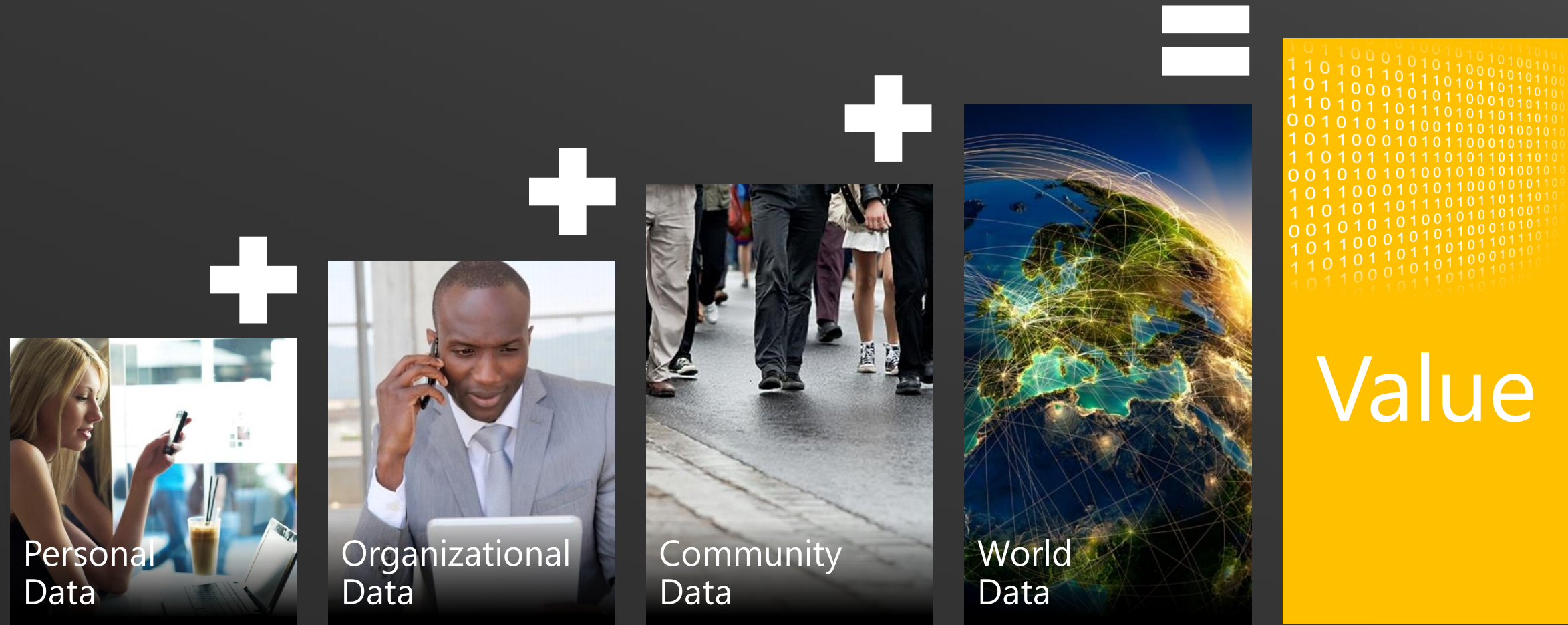
FROM



TO



POWER OF COMBINING THE WORLDS DATA

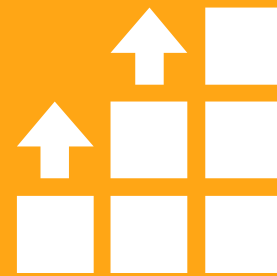


REFINE DATA

Credible,
Consistent Data



Data Mining

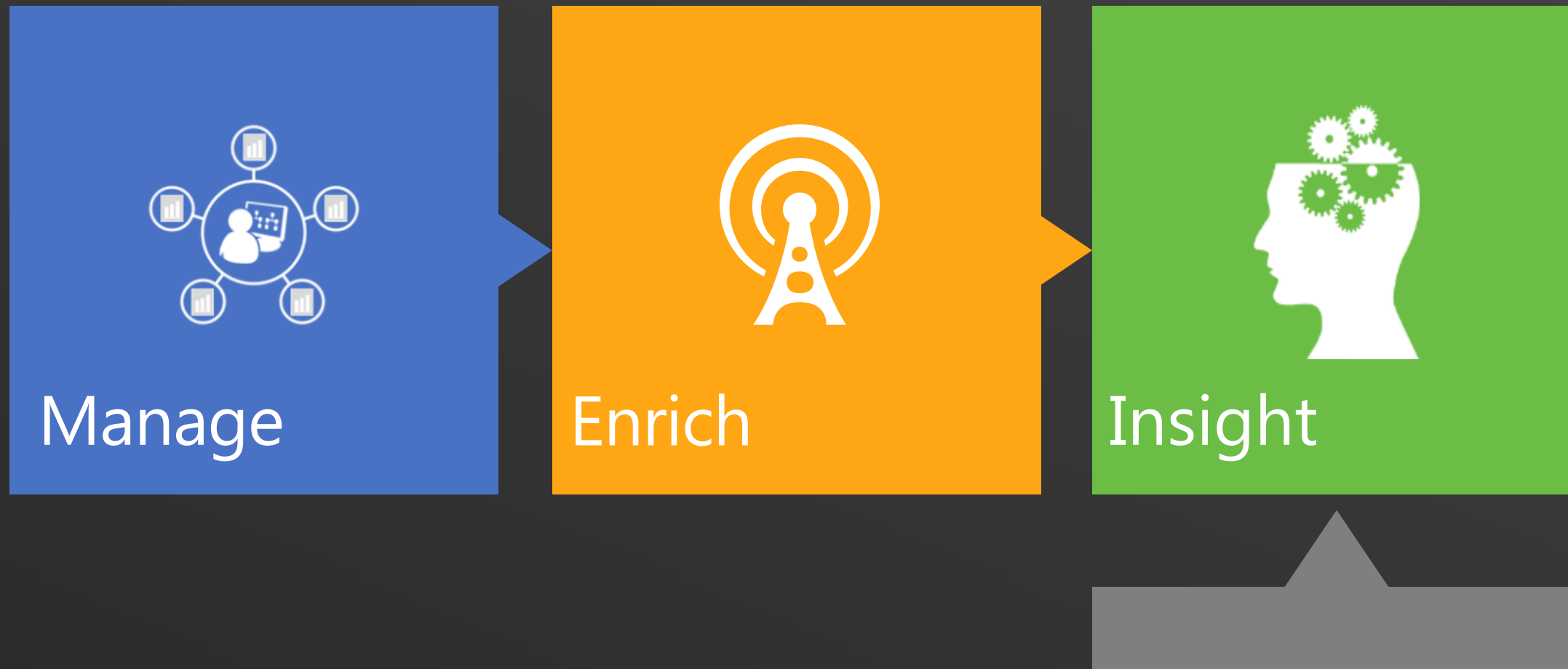


Advanced
Analytics

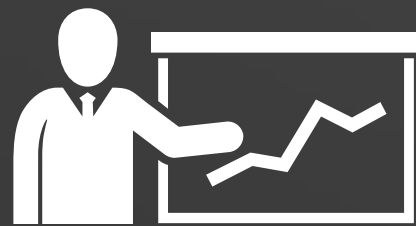


Enterprise Information Management & Full Analytic Spectrum

THE BIG DATA LIFECYCLE



INSIGHTS ON ANY DATA, ALL USERS, WHEREVER THEY ARE

Three white icons of database cylinders of increasing size.

Relational

A white icon of a piggy bank, representing non-relational data.

Non-Relational

A white icon of a server tower with a funnel of binary code (0s and 1s) above it, representing streaming data.

Streaming

INSIGHTS FOR ALL USERS THROUGH FAMILIAR TOOLS

PB

TB

GB



Data Scientists

BI Professionals

Business Analysts

Advanced Analytics
from Microsoft and 3rd
parties

Self Service Analysis
with PowerPivot &
Power View

Interactivity &
exploration with
Hadoop data in Excel

DEMO: FROM DATA TO INSIGHTS!



Simplicity



Analysis with familiar tools



Collaboration on insights

MONTE CARLO SIMULATION

Without a High Performance Computing Cluster

| | A | B | C | D | E | F | G | H | I | J | K | L | M |
|----|--|----------|------------------------------------|-----------|---|----------|----------|----------|----------|----------|----------|----------|----------|
| 1 | PRICING AN ASIAN OPTION USING MONTE | | | | 100 Iterations of Monte Carlo Pricing Runs | | | | | | | | |
| 2 | Up | 1.4 | | 9.5644481 | 9.536125 | 9.521659 | 9.539996 | 9.514193 | 9.526057 | 9.56552 | 9.546593 | 9.517877 | 9.517816 |
| 3 | Down | 0.8 | | 9.5404735 | 9.531961 | 9.540362 | 9.540075 | 9.55256 | 9.535553 | 9.532727 | 9.508754 | 9.542949 | 9.508293 |
| 4 | Interest | 1.08 | | 9.5220375 | 9.515621 | 9.549361 | 9.516186 | 9.511546 | 9.528434 | 9.527856 | 9.540048 | 9.527244 | 9.548104 |
| 5 | Initialprice | 30 | | 9.5176984 | 9.495455 | 9.538583 | 9.532983 | 9.536823 | 9.544104 | 9.521204 | 9.497344 | 9.533591 | 9.52301 |
| 6 | Periods | 20 | | 9.5387288 | 9.530959 | 9.530416 | 9.524015 | 9.485705 | 9.500098 | 9.509829 | 9.556509 | 9.52649 | 9.537136 |
| 7 | Exercise | 30 | | 9.5380206 | 9.520836 | 9.544378 | 9.54073 | 9.55753 | 9.534405 | 9.533942 | 9.512118 | 9.544364 | 9.536555 |
| 8 | Runs | 1000000 | | 9.5442322 | 9.539823 | 9.514977 | 9.515132 | 9.511578 | 9.53056 | 9.512695 | 9.529904 | 9.527162 | 9.53982 |
| 9 | Run on Cluster | No | | 9.5487908 | 9.535268 | 9.51969 | 9.505409 | 9.546035 | 9.549276 | 9.509327 | 9.506064 | 9.525897 | 9.526522 |
| 10 | Headnode | HEADNODE | | 9.5600378 | 9.544524 | 9.513452 | 9.52237 | 9.533193 | 9.544009 | 9.55402 | 9.542352 | 9.505605 | 9.558371 |
| 11 | | | | 9.5015553 | 9.523678 | 9.548339 | 9.501794 | 9.552113 | 9.546841 | 9.54658 | 9.524661 | 9.514506 | 9.542268 |
| 12 | | | | | | | | | | | | | |
| 13 | | Run | Average of Monte Carlo Runs | 9.53032 | | | | | | | | | |
| 14 | | | Min | 9.48571 | | | | | | | | | |
| 15 | | | Max | 9.56552 | | | | | | | | | |
| 16 | | Clear | Standard Deviation | 0.00167 | | | | | | | | | |
| 17 | | | Standard Error | 0.00017 | | | | | | | | | |
| 18 | | | Execution Time (seconds) | 73.3862 | | | | | | | | | |
| 19 | | | | | | | | | | | | | |

MONTE CARLO SIMULATION

With a 6-Node High Performance Computing Cluster

| | A | B | C | D | E | F | G | H | I | J | K | L | M |
|----|--|----------|------------------------------------|---|---------|---------|---------|---------|---------|---------|---------|---------|---------|
| 1 | PRICING AN ASIAN OPTION USING MONTE | | | 100 Iterations of Monte Carlo Pricing Runs | | | | | | | | | |
| 2 | Up | 1.4 | | 9.51221 | 9.53118 | 9.50463 | 9.54795 | 9.5245 | 9.51026 | 9.53652 | 9.52952 | 9.52985 | 9.52331 |
| 3 | Down | 0.8 | | 9.53598 | 9.51639 | 9.54376 | 9.52429 | 9.52849 | 9.52468 | 9.50633 | 9.50655 | 9.54271 | 9.52965 |
| 4 | Interest | 1.08 | | 9.51503 | 9.51099 | 9.53665 | 9.53031 | 9.51726 | 9.53966 | 9.52963 | 9.52828 | 9.53336 | 9.55192 |
| 5 | Initialprice | 30 | | 9.53977 | 9.5113 | 9.54376 | 9.52983 | 9.51897 | 9.53966 | 9.54592 | 9.54448 | 9.49397 | 9.55033 |
| 6 | Periods | 20 | | 9.5281 | 9.52883 | 9.51661 | 9.5079 | 9.50159 | 9.52188 | 9.50775 | 9.55156 | 9.50308 | 9.51876 |
| 7 | Exercise | 30 | | 9.54999 | 9.56914 | 9.53289 | 9.53315 | 9.51726 | 9.53155 | 9.50481 | 9.53431 | 9.49397 | 9.55033 |
| 8 | Runs | 1000000 | | 9.51221 | 9.57521 | 9.52185 | 9.53812 | 9.55011 | 9.50767 | 9.53242 | 9.52976 | 9.53401 | 9.54191 |
| 9 | Run on Cluster | Yes | ← | 9.51503 | 9.53289 | 9.53737 | 9.51661 | 9.53746 | 9.54163 | 9.53751 | 9.55544 | 9.53603 | 9.51495 |
| 10 | Headnode | HEADNODE | | 9.53977 | 9.51241 | 9.53193 | 9.49235 | 9.53931 | 9.53937 | 9.55348 | 9.51894 | 9.52946 | 9.54519 |
| 11 | | | | 9.5281 | 9.56914 | 9.50796 | 9.55238 | 9.51726 | 9.57313 | 9.54133 | 9.53281 | 9.53205 | 9.55887 |
| 12 | | | | | | | | | | | | | |
| 13 | Run | | Average of Monte Carlo Runs | 9.53007 | | | | | | | | | |
| 14 | | | Min | 9.49235 | | | | | | | | | |
| 15 | | | Max | 9.57521 | | | | | | | | | |
| 16 | Clear | | Standard Deviation | 0.00173 | | | | | | | | | |
| 17 | | | Standard Error | 0.00017 | | | | | | | | | |
| 18 | | | Execution Time (seconds) | 8.53215 | ← | | | | | | | | |
| 19 | | | | | | | | | | | | | |