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# Gartner

## Preparing and Architecting for Machine Learning



Carlton E. Sapp Research Director

1 year at Gartner, 20 years industry experience

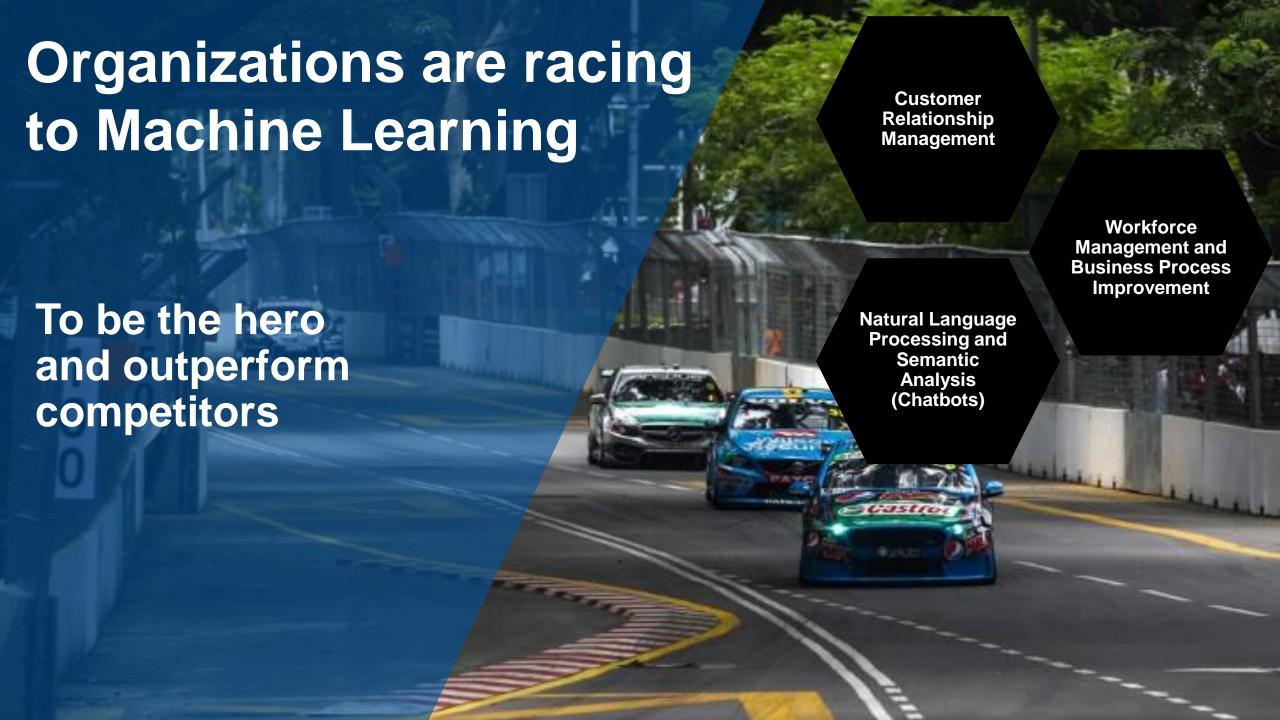
Carlton provides guidance on data ingestion strategies, streaming analytics, data integration, machine learning and cognitive analytics. He provides data-driven insights based on practical project experience and research. He is known for using systems thinking approaches to solve complex technical and business challenges.



https://https://www.linkedin.com/in/drcarltonesapp/







### Path to Being the Hero

- What is machine learning and why is it important?
- What is the impact to our architectures and how do we build for machine learning capabilities?
- How can I get started?



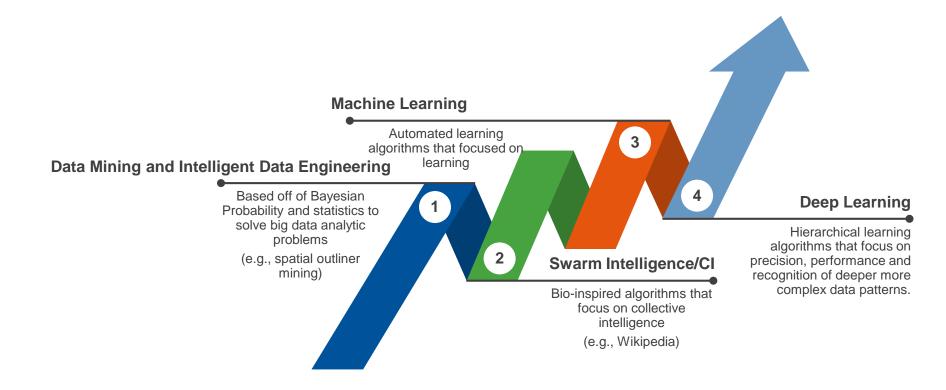
# Defining Machine Learning

• Machine learning (ML) — a subset of artificial intelligence (AI) — is more than a technique for analyzing data. It's a system that is fueled by data, with the ability to learn and improve by using algorithms that provide new insights without being explicitly programmed to do so

Note: There are many types of intelligence (swarm, learning, etc.)
Al vs. Machine Learning vs. Deep Learning vs. Data Mining



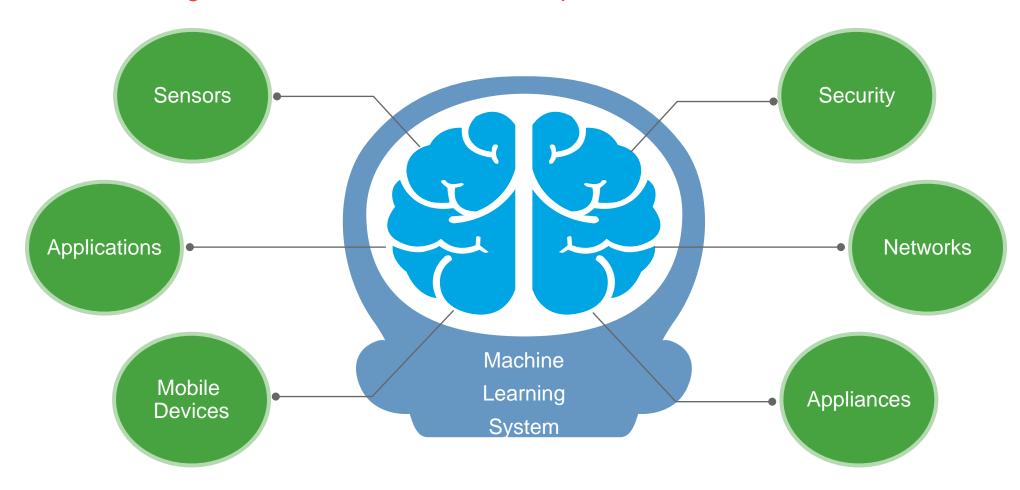
## **Machine Learning Lineage**





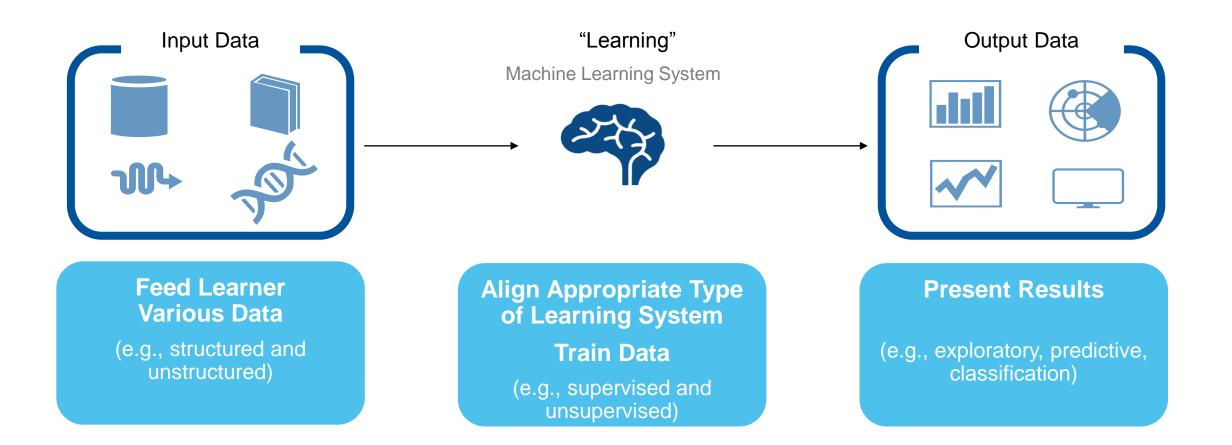
# Machines Learning From Data Without Being Explicitly Programmed

Extracts Knowledge From Data to Transform Simple Machines Into Smart Machines:



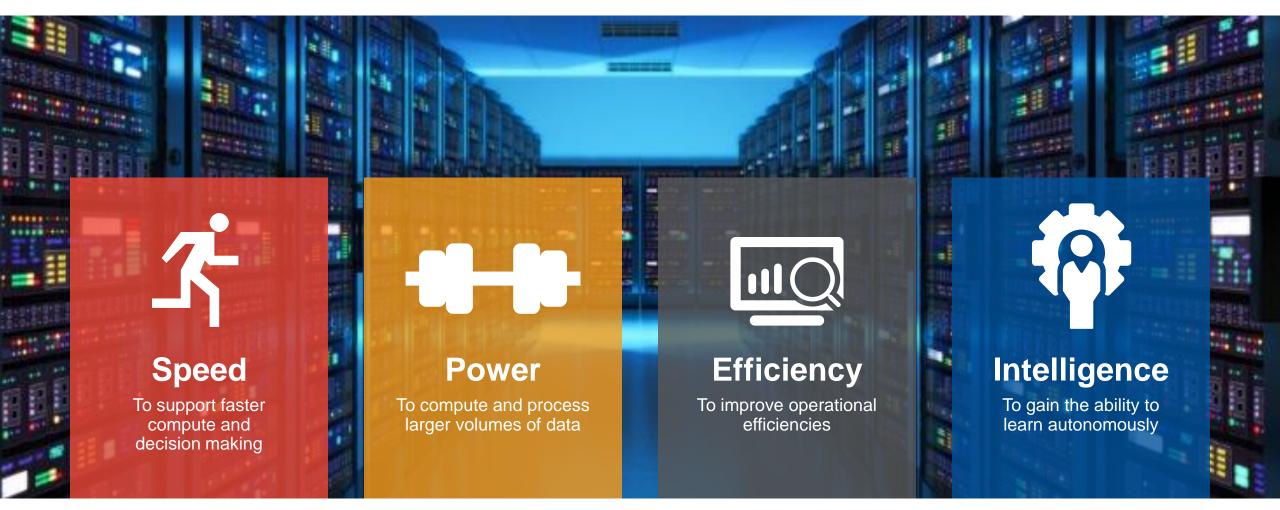


### The Basics of Machine Learning Technology



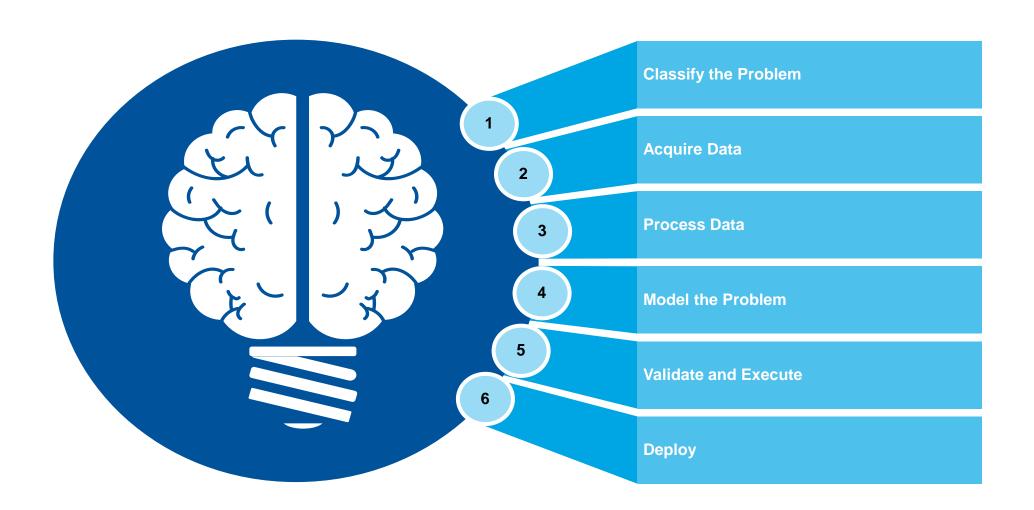


# Given Big Data, Machine Learning Offers:





## **Learn the Basics of Machine Learning Stages**





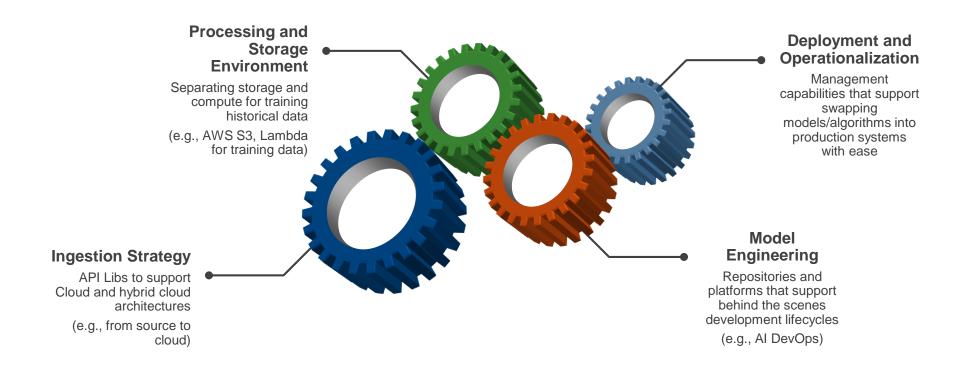
What is the impact of Machine Learning to our architecture?

How do we build for Machine Learning capabilities?

### Impact of AI on the Data and Analytics Architecture

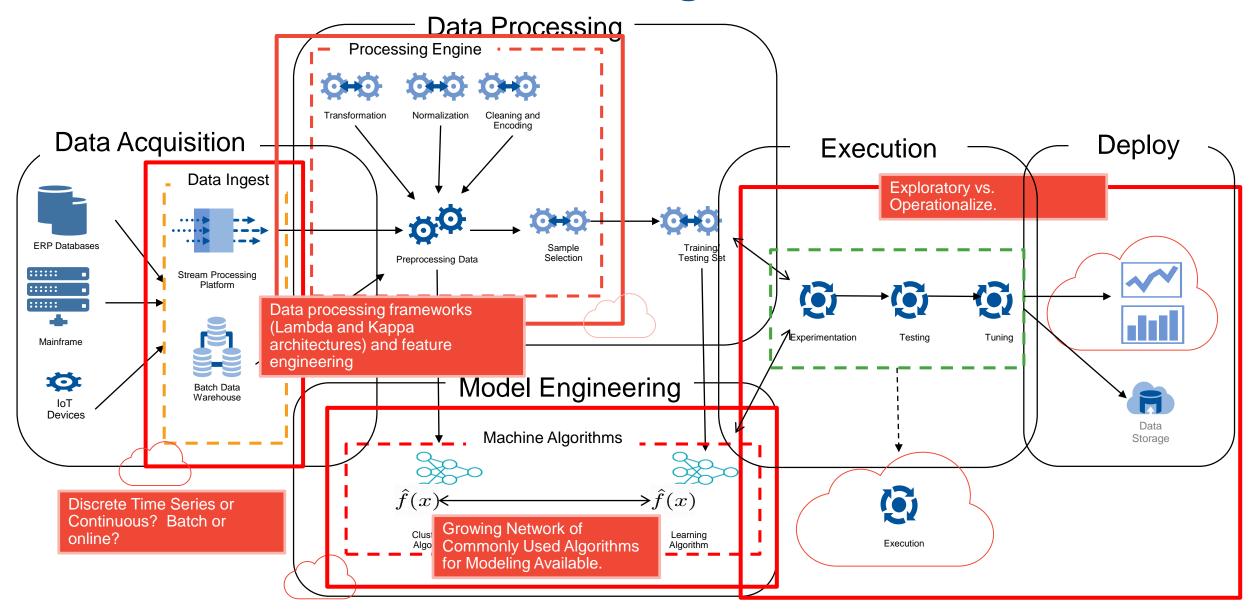
What changes should I make to my architecture to prepare for AI?

**Top concerns from technical professionals (n=148)** 





# **Evaluate Basic Machine Learning Architecture**



### **Critical Architecture Components Exposed**

**Goal: Flexible integration policies** that cross data and ML platforms

Integration of streaming data with data at rest

> Goal: Model repository with metadata

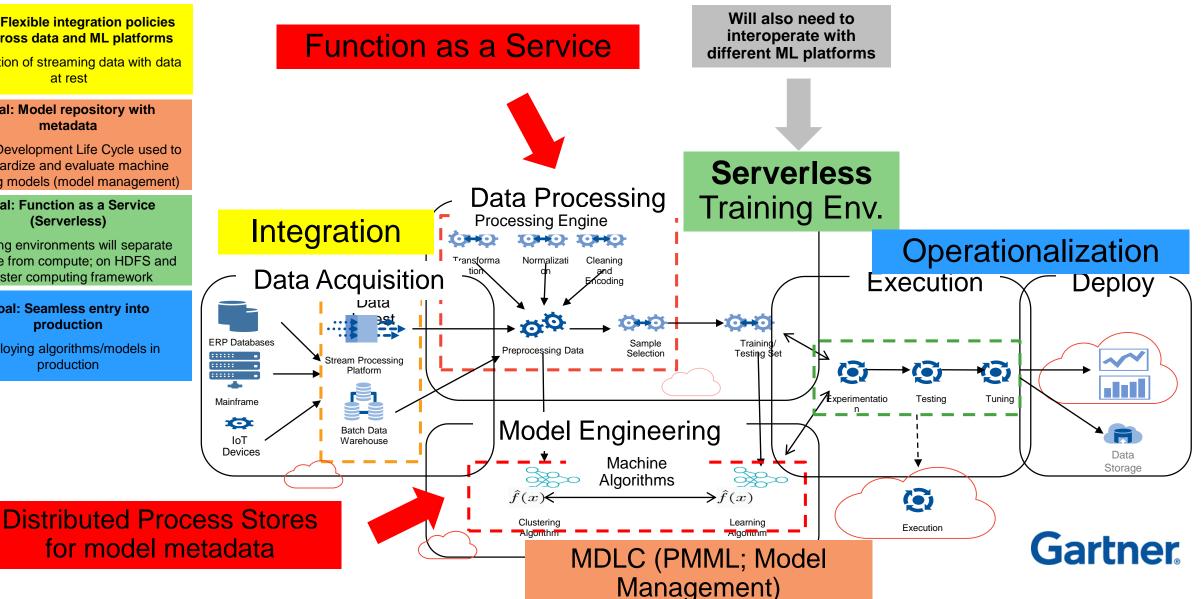
Model Development Life Cycle used to standardize and evaluate machine learning models (model management)

> Goal: Function as a Service (Serverless)

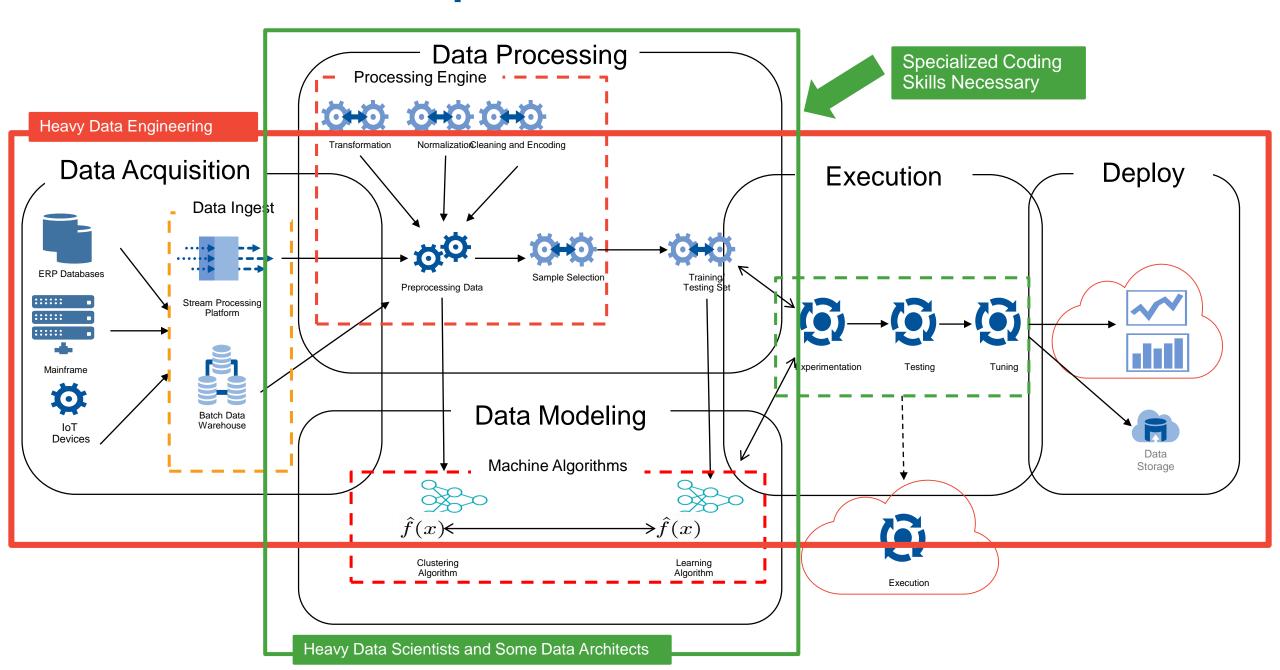
Training environments will separate storage from compute; on HDFS and cluster computing framework

> Goal: Seamless entry into production

Deploying algorithms/models in production



### **Understand the Enterprise Roles Involved**



### **Build A Machine Learning Stack**

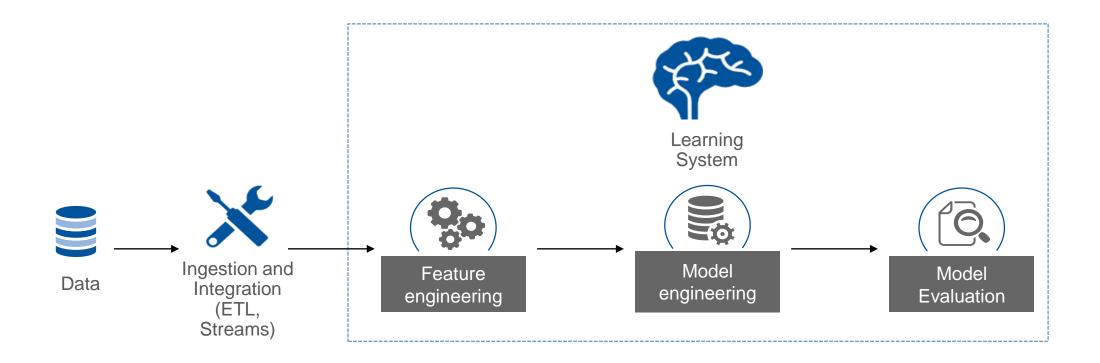
Distributed Processing Test (Serverless Training Environment) Environment (e.g, Apache Spark/Storm/Cloud for hyperparameter optimization) (A/B Testing) **Data Pipelines** Machine Learning Platform/ Repository/Management Dev Transport Layer ML Lib, AI/ML/DL Framework **Environment** (online vs batch) (includes integration, PaaS, PMML, or Proprietary) (sensitivity analysis) **Production** Distributed Storage **Environment** (Analytical Data Lake; Heavy Read/Write; ex: Apache Cassandra, Hadoop) (e.g., PMML)



Establish a data and analytics lifecycle model for machine learning



## **Example: Model Development Lifecycle for Platforms**





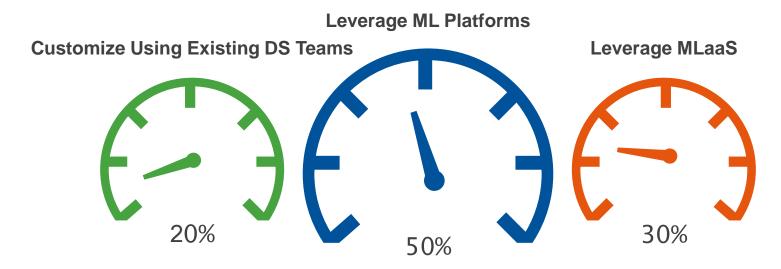


# How to get started?



### **Getting Started**

Clients initiating Machine Learning initiatives (n=148)



#### **Customizations**

Customized solutions with existing data science teams leverage existing knowledge in quantitative statistics leveraged by machine learning

### **Leveraging ML Platforms**

Many ML platforms currently on the market provide pre-build toolkits that can be leverage by data science teams

#### Leveraging MLaaS with **API and SDK**

Leveraging ML as a service provides API and SDK for environments where integrating numerous services is required



# **Polling Question 1 of 1**

### Rate your level of interest in developing or coding machine learning algorithms from scratch:

- Very interested
- B. Somewhat interested
- C. No opinion
- Not really interested
- Absolutely not interested



### **Roadmap To Success**

Clients initiating successful Machine Learning initiatives (n=148)



### **Four Steps To Success**

### Step 01

#### Collaborate

Work closely with data science teams and business analysts to identify use cases

#### Step 03

#### **Iterative Rollout**

Iteratively expand your ML platform and services over time

#### Step 02

#### Leverage Cloud and Platforms

Build use cases in the public cloud or using toolkits to avoid excessive systems engineering work

#### Step 04

#### Integrate with **DevOps**

Develop a process for operationalizing ML applications



## **Examples of Machine Learning Delivering Value**

Case Studies by SRS





### Recap

- 1. Machine learning is about acquiring knowledge through data
- 2. Machine learning technologies offers us speed, power, efficiency and intelligence
- 3. Machine learning is the next-generation analytic tradecraft for the digital business architect
- 4. We build for machine learning by understanding the problem, acquiring data, processing data, modeling, executing and deploying
- 5. Start with a business challenge, build sample programs, and execute use cases iteratively





### **Recommended Gartner Research**

- Solution Path for Evolving Your Business Analytics Program Carlie J. Idoine (G00292685)
- Big Data Analytics Failures and How to Prevent Them Svetlana Sicular (G00272497)
- Advancing Business With Advanced Analytics
   Alexander Linden (G00277713)
- Machine Learning Drives Digital Business
   Alexander Linden, Noha Tohamy and Others (G00263964)



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