

IAITAM ACE 2025

ITAM - Another Brick In The Wall

Integrating AI & Machine Learning into IT Asset Management



omnitza
Technology Asset Management

Speaker



AVIK GHOSH

VP, AI & Data Product Management

oomnitza

Technology Asset Management

Avik leads Oomnitza's efforts with helping companies drive measurable business value through AI, data solutions for modern IT Asset Management.

Agenda

- Introduction
- Identifying High-Value Use Cases
- Analyzing Data & Building ML Models
- Deploying Apps and Tools
- Key Takeaways & Open Discussion

Introduction

Objectives of This Session

- ❑ Understand how ITAM + AI can drive business value.
- ❑ Understand practical insights into deploying AI effectively in an enterprise IT environment.
- ❑ Discuss challenges and strategies for ensuring high-quality data.
- ❑ Open discussion on real-world challenges and use cases.

AI Is Expected To Drive Significant Business Value For ITAM

Key Areas of Business Value

Improve Operational Efficiency & Effectiveness

66% Increased productivity

30% Increased efficiency

Reduce Risk

100% Audit compliance

33% Improvement in operational reliability

28% Reduction in cybersecurity threats

Reduce Costs & Optimize Profit

70% Reduction in wasteful spending

30% Increase in profitability

5x Average ROI on technology investments

Accelerate Business & Drive Growth

150% Faster deployments of services

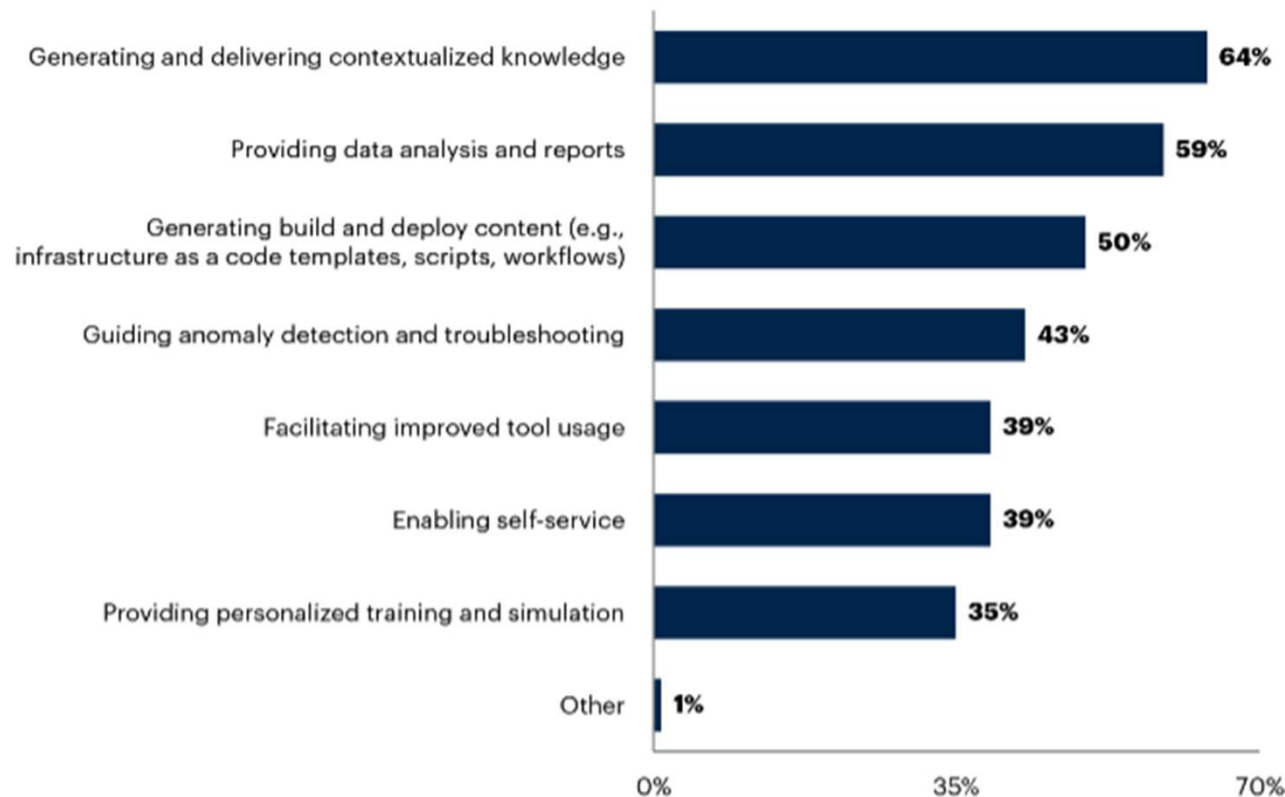
30% Faster decision making

12% Revenue increase from “tech-as-a-service” business models

ITOP Teams Are Accelerating Adoption of AI to Drive Value

Current and Intended Use Cases for GenAI in IT Operations

Multiple responses allowed

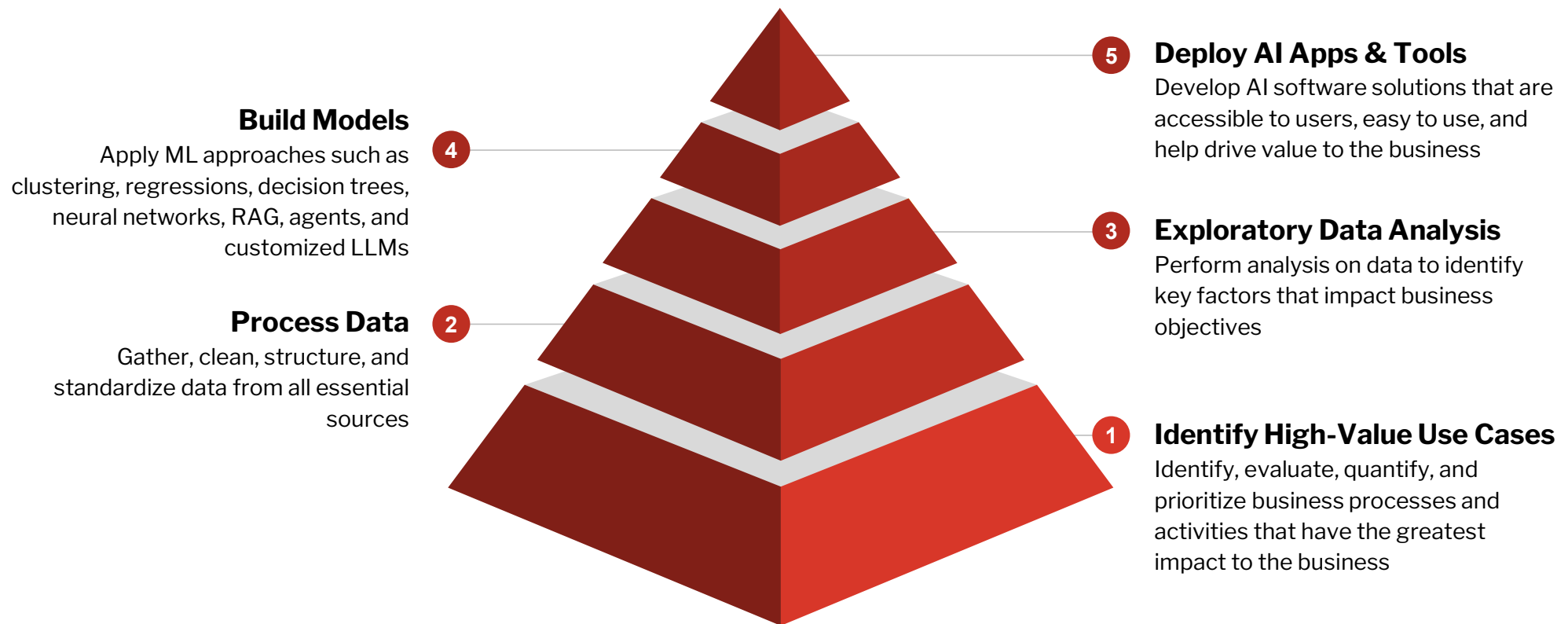


60%

of all IT operations
tools will include
AI Agents by 2028

Source: Gartner

Approach for Deploying AI in Business



Identify High-Value Use Cases

Start by Understanding Business Value Drivers for ITAM



Identify Themes & Value Drivers

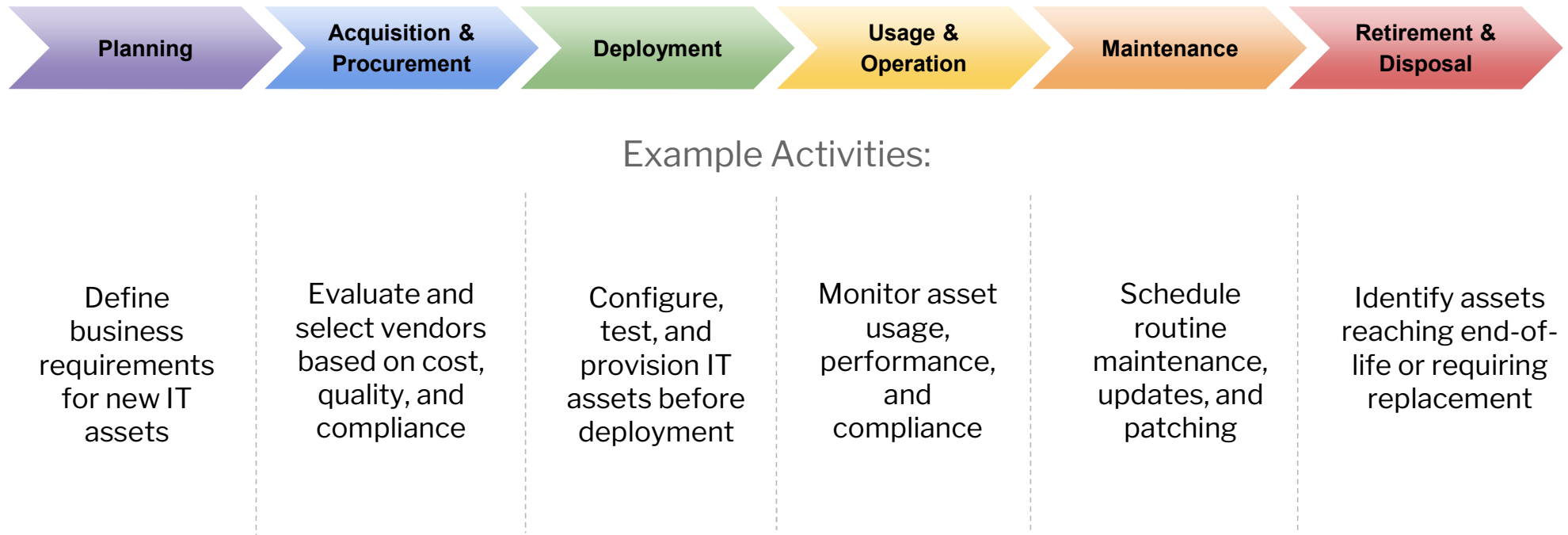
**Improve Operational
Efficiency &
Effectiveness**

Reduce Risk

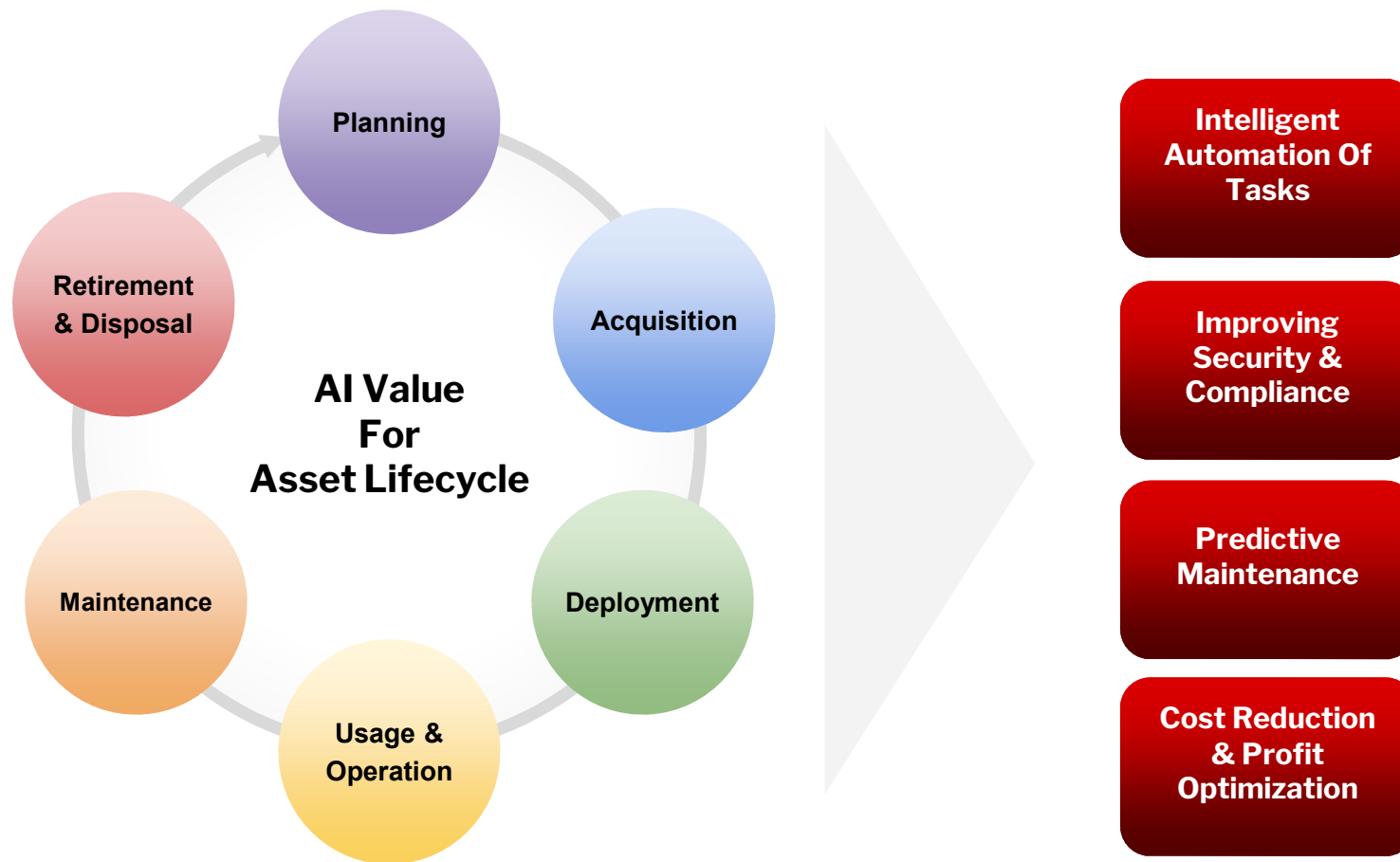
**Reduce Costs &
Optimize Profit**

**Accelerate Business &
Drive Growth**

Identify High-Value Activities Across Asset Lifecycle Mgmt



Identify How AI Can Enable & Accelerate Key Activities



Example AI Solution For ITAM:

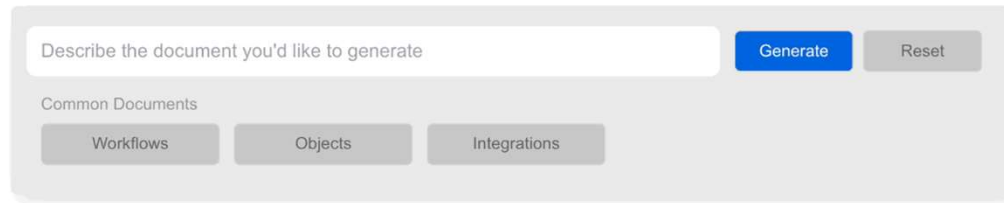
Oomnitza AI-Driven Workflow Generation



A screenshot of a web interface for AI-driven workflow generation. It features a light gray rounded rectangular container. Inside, on the left, is a white text input field with the placeholder text "Describe the workflow you'd like to generate" and a mouse cursor pointing at the start. To the right of the input field are two buttons: a blue "Generate" button and a gray "Reset" button.

Example AI Solution For ITAM:

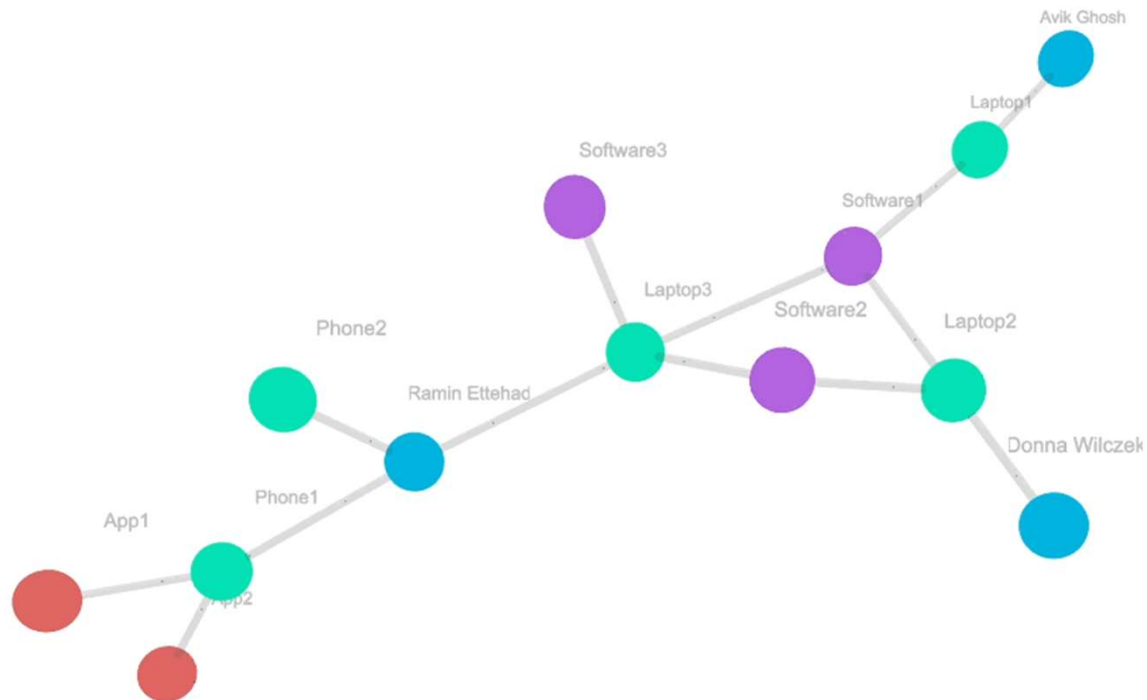
Oomnitza AI-Driven System Documentation For Audits



The screenshot shows a user interface for generating documentation. It features a text input field with the placeholder text "Describe the document you'd like to generate". To the right of the input field are two buttons: a blue "Generate" button and a grey "Reset" button. Below the input field, the text "Common Documents" is displayed. Underneath this text are three grey buttons labeled "Workflows", "Objects", and "Integrations".

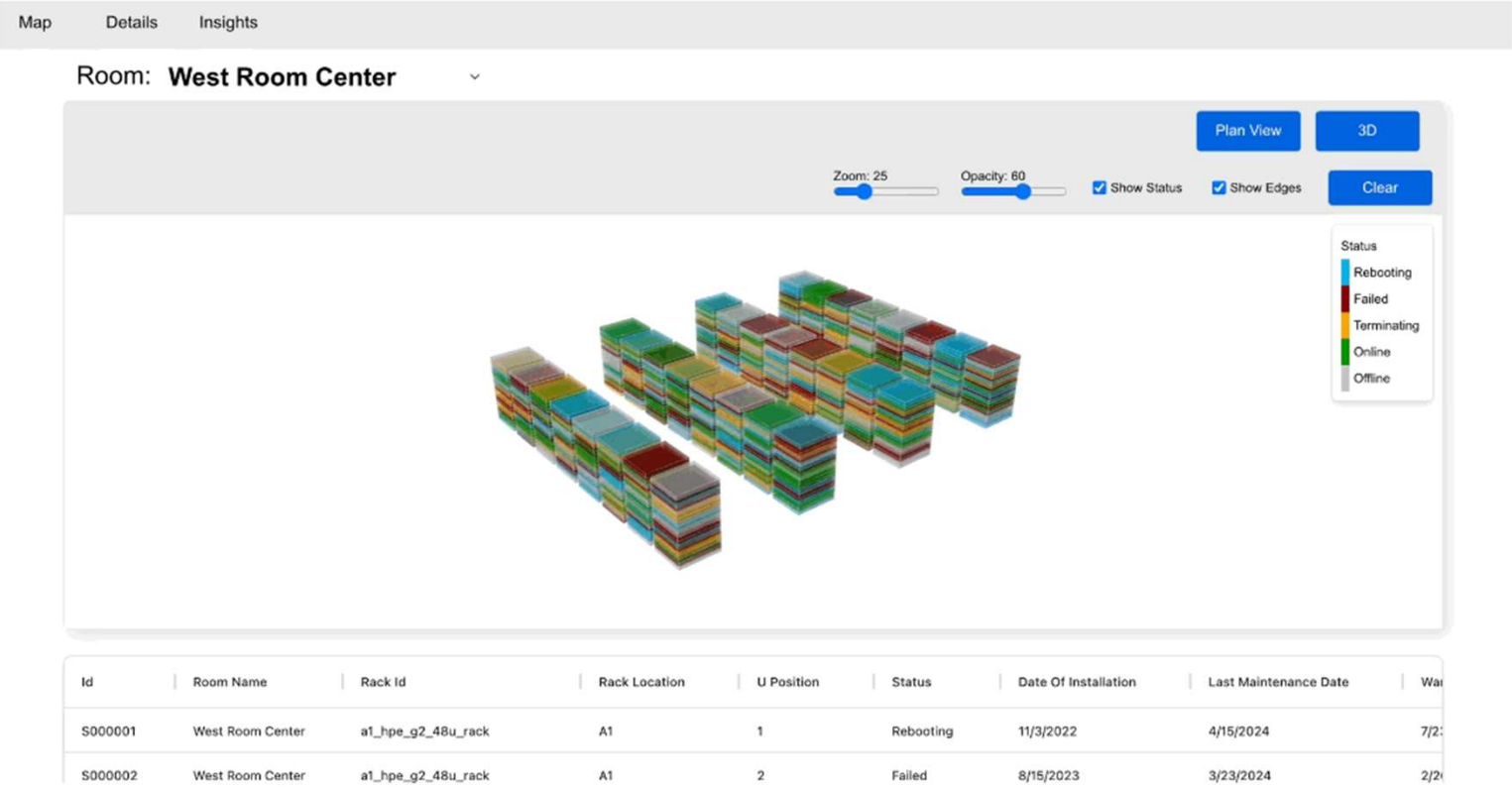
Example AI Solution For ITAM:

Oomnitza AI-Driven Relationship Modeling (Graph RAG)



Example AI Solution For ITAM:

Oomnitza AI-Driven Data-Center Infrastructure Management



Analyzing Data & Building ML Models

Basic Concept & Approach

Given several variables (x), predict a target variable (y)

$$y = \mathbf{w_1}x_1 + \mathbf{w_2}x_2 + \mathbf{w_3}x_3 \dots + \mathbf{w_n}x_n$$

w= weights. Since x and y are determined from data, **the goal is to calculate weights**



Most Businesses Struggle With Data Quality & Processing

**Errors With
Data Collection**

**Limited
Automation**

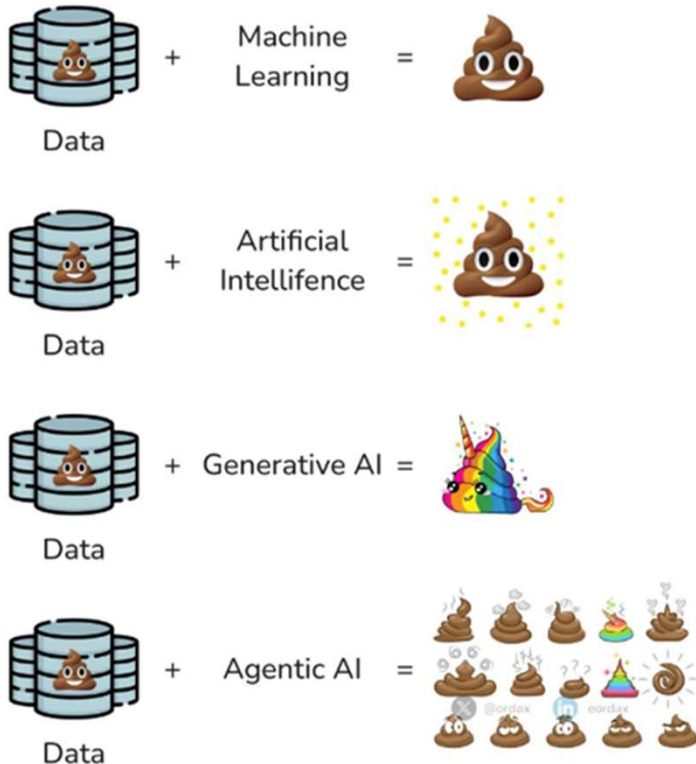
**Poor Data
Model Design**

**Shadow IT and
Untracked
Assets**

**Siloed Data
Sources**

**Poor Data
Governance**

Bad Data Leads to Bad AI Leads to Critical Business Issues



“Highly capable **agentic AI** using inaccurate data will cause **outages, exposed attack vectors and budget overruns**. I&O leaders must take immediate action to address data quality in technology asset repositories, observability and IT monitoring tools, configuration management databases (CMDBs), and other systems of record.”

– Roger Williams | VP, Research at Gartner

Source: *Innovation Insight: Agent-Native I&O*

February 26, 2025

omnitza
delivers accurate,
high quality ITAM data

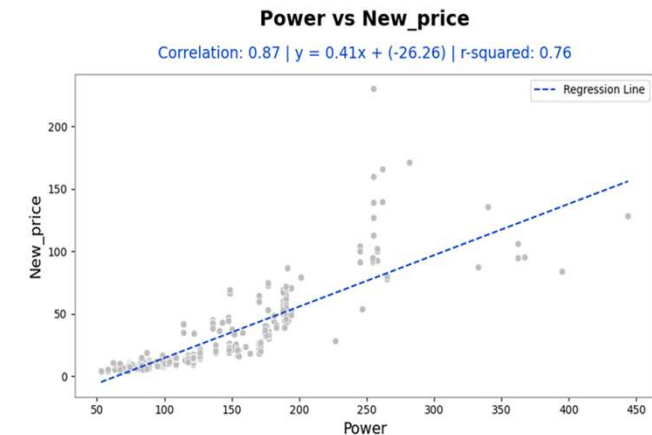
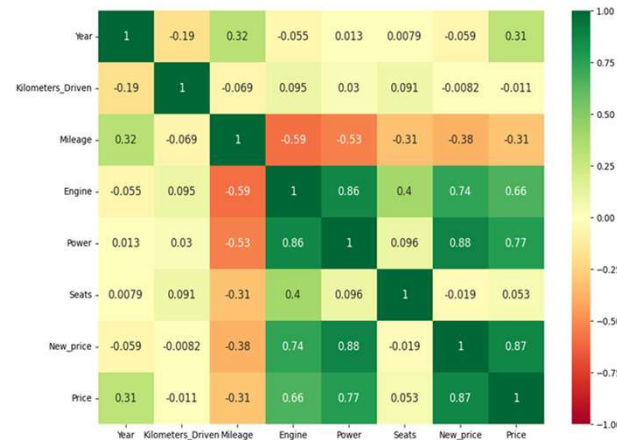
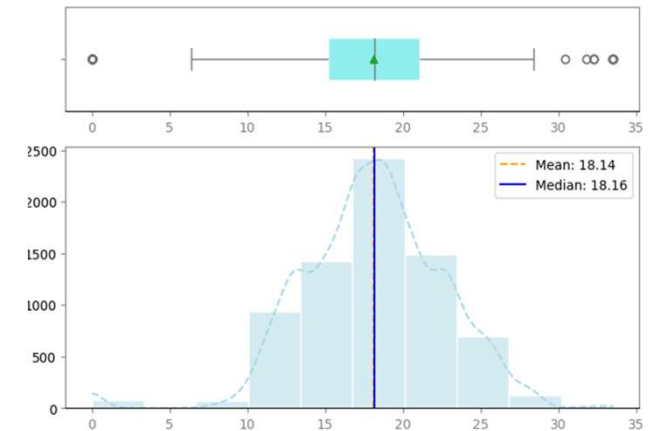
Exploratory Data Analysis (EDA)

What is EDA?

- Combination of visualization techniques and statistical methods
- Exploring and summarizing key information within the data

Why EDA is Useful?

- Gain good insights into the data
- Uncover underlying structure of the data
- Determine the best strategy to handle unclean data (**missing values, outliers** etc.)
- Identify initial set of observations and insights
- Establish intuitive understanding of data to help validate results derived from ML models later



Preparing Data For Machines To Understand

Raw Data

Features
(also Dimensions)

Goal:
Predict
This

Observations

SquareFeet	Bedrooms	Bathrooms	Neighborhood	YearBuilt	Price
2,666	3	1	Suburb	1985	\$170,070
2,827	3	2	Urban	2001	\$346,339
2,514	5	2	Rural	1996	\$210,800
1,779	5	1	Suburb	2011	\$127,488
2,389	3	1		1986	\$307,190
2,897		3	Urban	1995	\$305,172
2,000	4	3	Rural	2009	\$286,479
1,604	3	3	Urban	1978	\$87,906
1,011	5	2	Suburb	1994	\$145,770
2,207	3	1	Rural	1984	\$317,883

Numerical
Feature

Numerical
Feature

Numerical
Feature

Categorical
Feature

Numerical
or
Categorical

Numerical
Feature

Ensures that the model can better fit the data. If the intercept is omitted, the model may perform poorly and not capture the true relationship between the features and the target variable, especially if the data does not inherently pass through the origin.

Encoded Data

Independent Variables
(X values)

Dependent
Variable
(Y values)

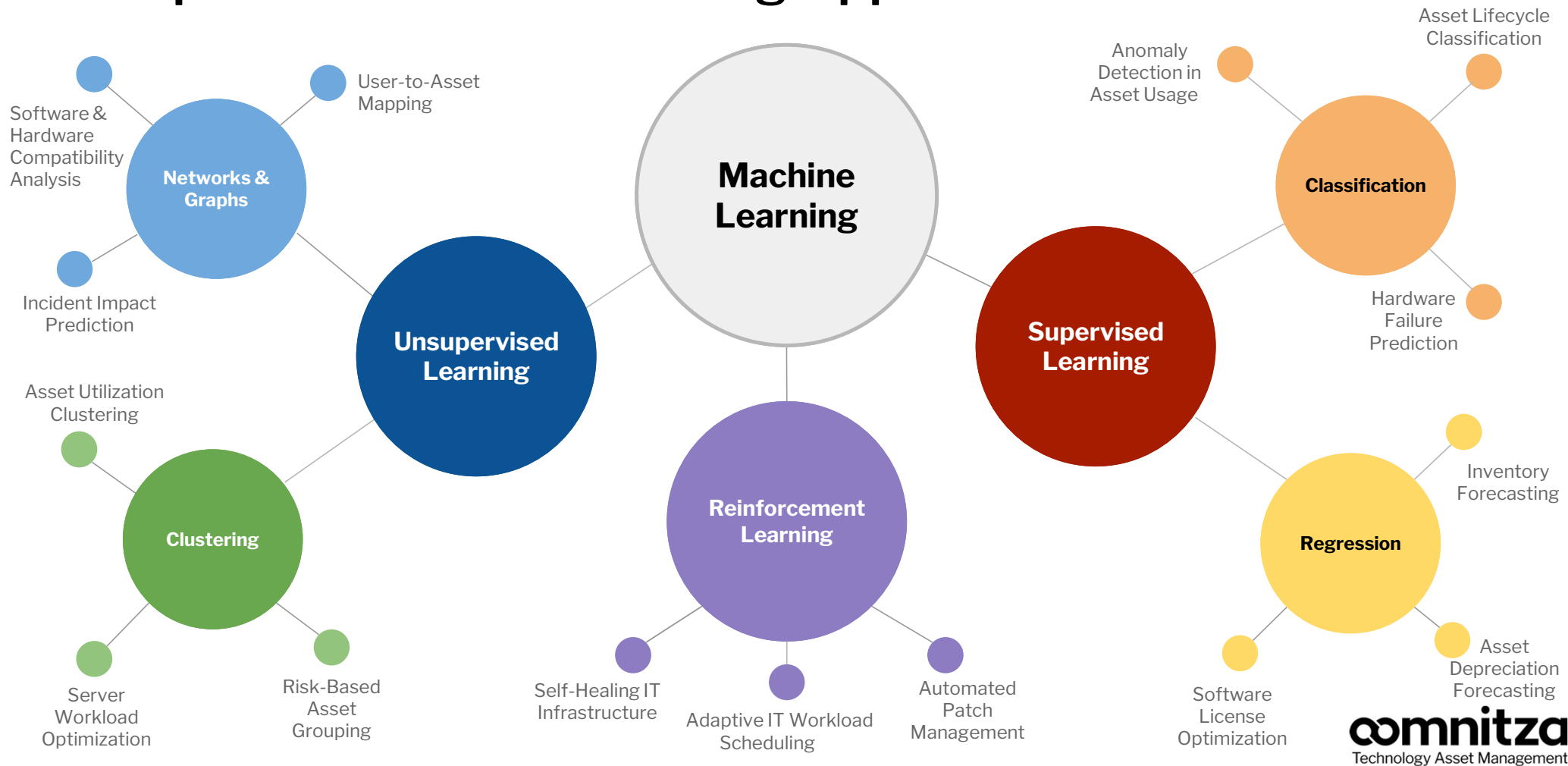
Intercept	SquareFeet	Bedrooms	Bathrooms	YearBuilt	Neighborhood_Suburb	Neighborhood_Urban	Price
1	2666	3	1	1985	1	0	170070
1	2827	3	2	2001	0	1	346339
1	2514	5	2	1996	0	0	210800
1	1779	5	1	2011	1	0	127488
1	2389	3	1	1986	1	0	307190
1	2897	3	3	1995	0	1	305172
1	2000	4	3	2009	0	0	286479
1	1604	3	3	1978	0	1	87906
1	2011	5	2	1994	1	0	145770
1	2207	3	1	1984	0	0	317883

Treat null values:

- Assumed **mode** for **categorical features**
- Assumed **median** for **numerical features**
- Since ML models are trained to look at variation, median and mode don't have negligible impact on predictions. Just address data gaps.

Categorical features need to be represented numerically as binary number (1 or 0).

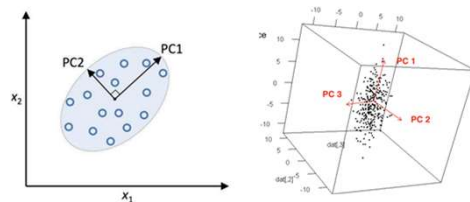
Examples of Machine Learning Approaches in ITAM



Common Machine Learning Approaches

Principal Component Analysis

Identifies the most significant features that impact the prediction variable



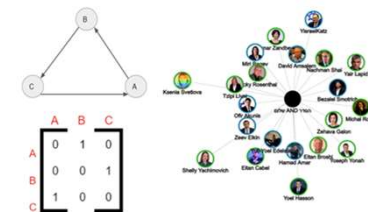
Clustering

Identifies patterns in data without prior labels or categories.



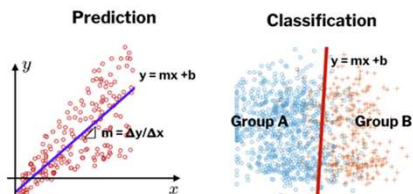
Network Analysis (Graph)

Models complex relationships



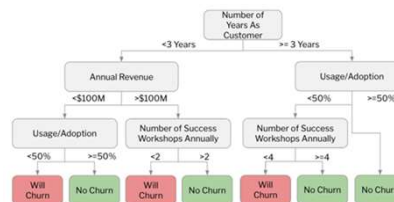
Regressions

Predicting continuous and discrete outcomes of linear relationships.



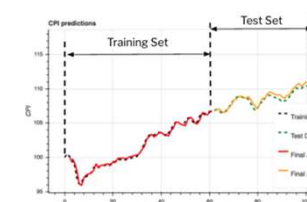
Decision Trees

Provides clear, interpretable models for predicting complex non-linear relationships



Time Series Forecasting

Predicting future values based on previously observed data over time.



Recommendation Systems

Collaborative filtering based on item characteristics to make recommendations

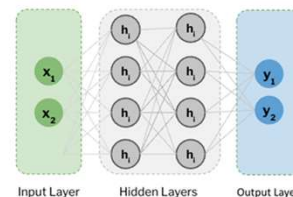
UserID	Role	Tenure	Level
1	Eng	5	NY
2	Mktg	3	LA
3	Sales	10	NY
4	Ops	8	LA
5	Support	1	SF

AssetID	Brand	Model
1	Apple	MacBook Air M4
2	Asus	Zenbook 14 OLED (Q425M)
3	Acer	Swift Go 14
4	Apple	MacBook Pro 14 (2024)
5	Dell	XPS 13

UserID	Asset 1	Asset 2	Asset 3	Asset 3	Asset 3
1	4	5	3	5	5
4	4	3	5	2	4
5	4	1	4	2	2

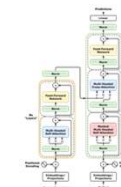
Neural Networks

Uses interconnected nodes to learn complex patterns and make predictions



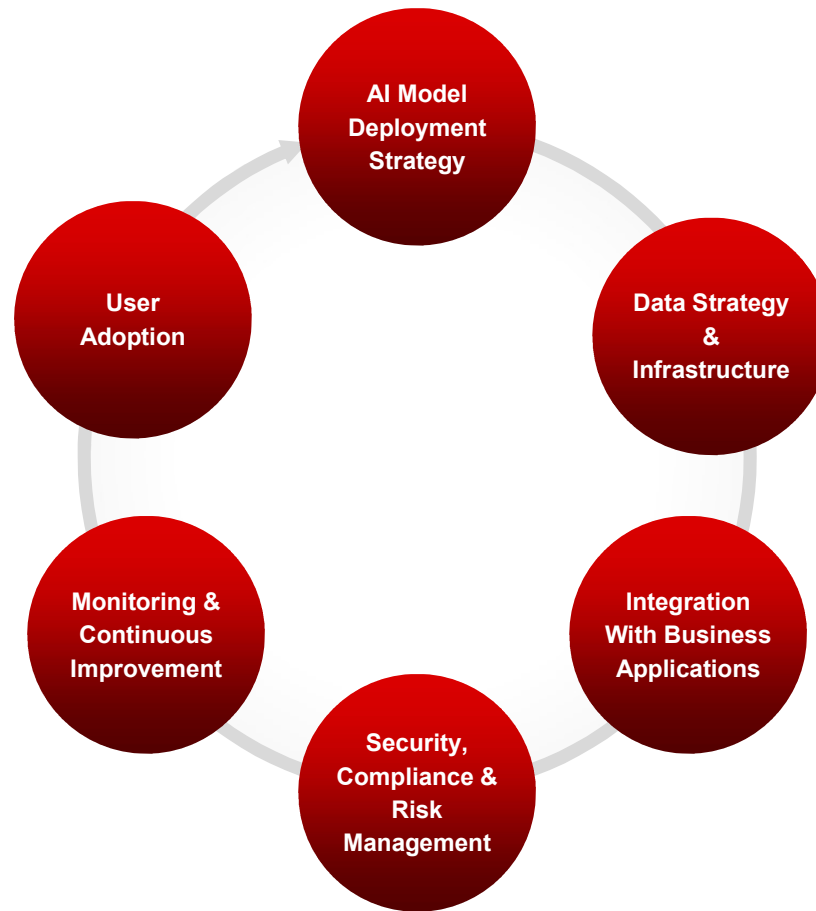
Transformers & LLMs

Natural language processing (NLP) for next-word prediction



AI Apps and Tool Deployment

Key Factors to Consider for AI Deployment



Key Takeaways & Open Discussion

Three Key Takeaways

1. AI is Only as Good as Your ITAM Data

Without accurate, complete, and well-structured IT asset management data, AI solutions will struggle to provide meaningful business value. Poor data quality leads to unreliable predictions, inefficiencies, and failed AI initiatives.

1. Business Context is Critical

If you don't have a clear understanding of how ITAM drives business value—such as cost optimization, security, compliance, or efficiency—AI-driven solutions will lack the right focus, leading to misaligned outcomes and wasted investment.

2. Data Governance and Strategy are Prerequisites for AI Success

AI isn't a magic fix; it depends on a strong ITAM foundation. Prioritize data governance, normalization, and integration across IT ecosystems to maximize AI's potential to enhance decision-making and drive business impact.

Connect With Me



AVIK GHOSH

VP, AI & Data Product Management

oomnitza

Technology Asset Management

Email: **avik.ghosh@oomnitza.com**

LinkedIn: **avikghosh03**

