Towards a Big Data-enabled Digital Twin for Large-Scale Infrastructure Planning

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EDS

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Digital Twin – Key Concepts

- "An integrated multiphysics, multiscale, probabilistic simulation of an as-built system, enabled by Digital Thread, that uses the best available models, sensor information, and input data to mirror and predict activities/performance over the life of its corresponding physical twin." (DAU Glossary)
- A Digital Twin is created through a combination of
 - Data: sensors data, model data and domain knowledge
 - Subject matter expertise
 - Modeling: physics-based models (electrical, hydraulic, etc.), simulation/kinematic models, discreteevent, etc.
- Mirrors replication of a physical asset; Mimics its structure, context and behavior; connects different value chains
- Provides a window to the past and present states and conditions of the physical asset
- Dynamic in nature: is constantly updated with data coming from the system/physical asset → When coupled with the right level of intelligence, allows us to look into the future

Towards A Digital Campus

GT Campus as a Digital Twin of a System of Systems



Endgame: Campus of the Future

A Multi-layered System of Systems

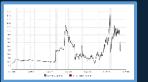


Sustainable



Emerging Tech

Resilient





Weather Threats

Rising Energy Costs



Changing Policy



Cyber Threats Emergency Events

Adaptable



Campus Growth





Engaged

Stakeholders

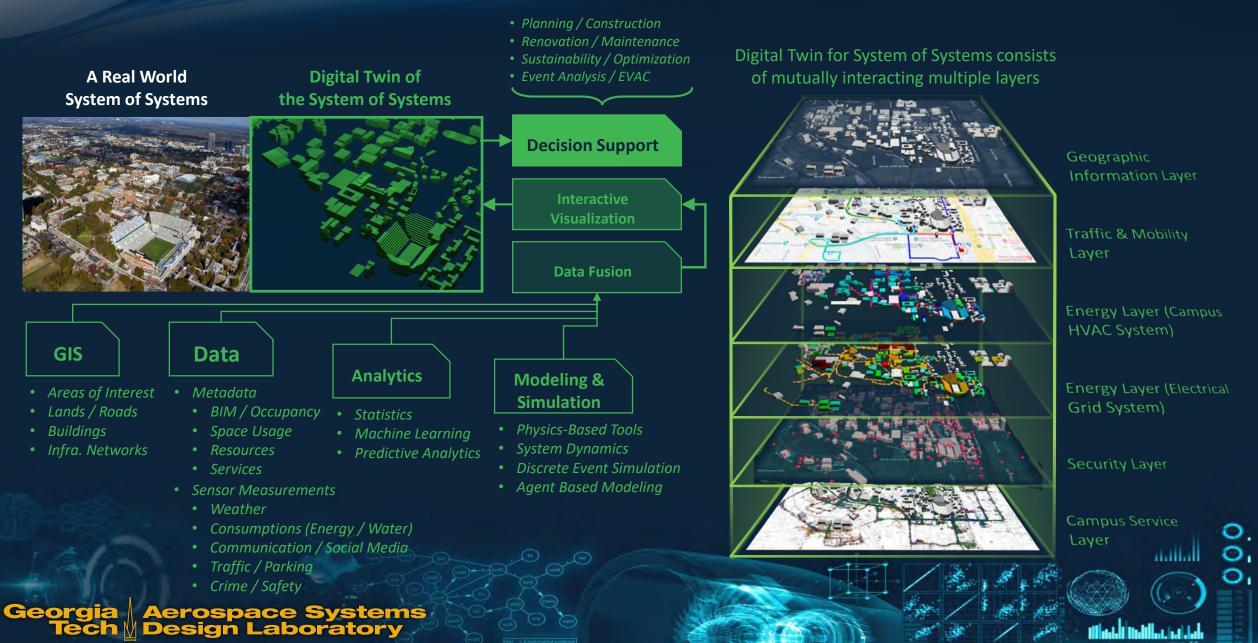




Evolving Cityscape



Vision: Digital Twin of Georgia Tech Campus



Towards A Digital Campus

Focus on Infrastructure & Energy

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Smart Campus: Targeted Capabilities

Running Campus Smartly

Focus: Existing campus, improving it as it is



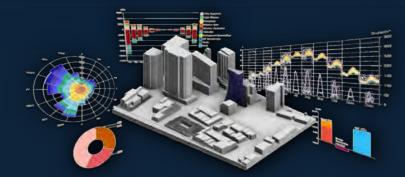




Goals: Cost avoidance, energy savings, reliability, safety

Planning Campus Smartly

Focus: Future campus, Future scenario forecasting

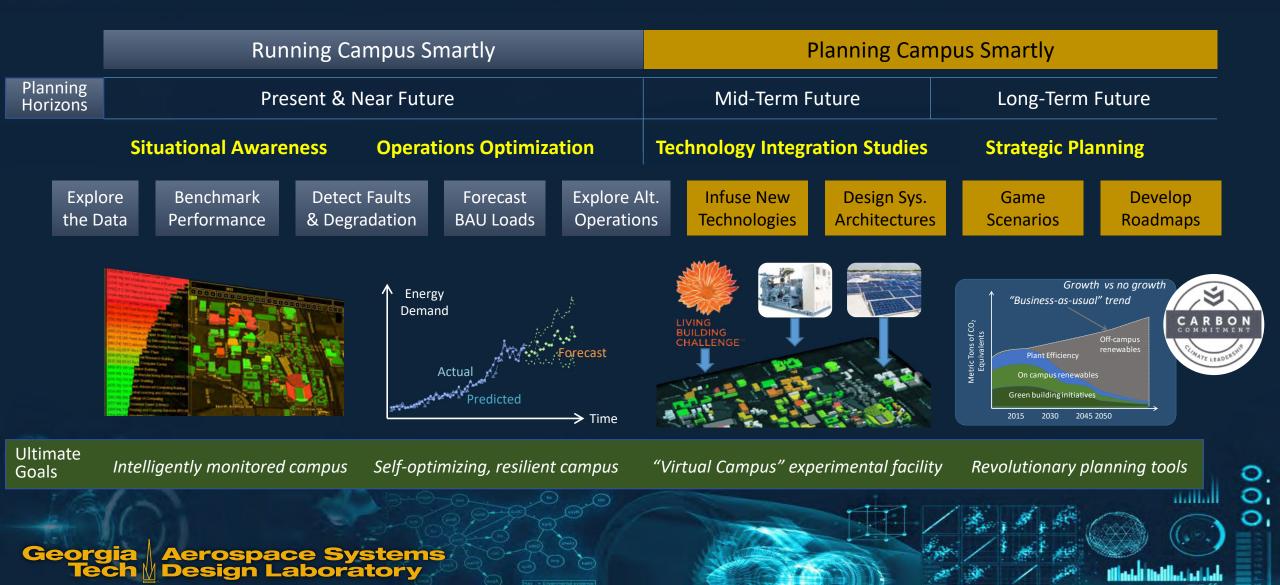


Goals: Data-driven decision making, strategic gaming, etc.

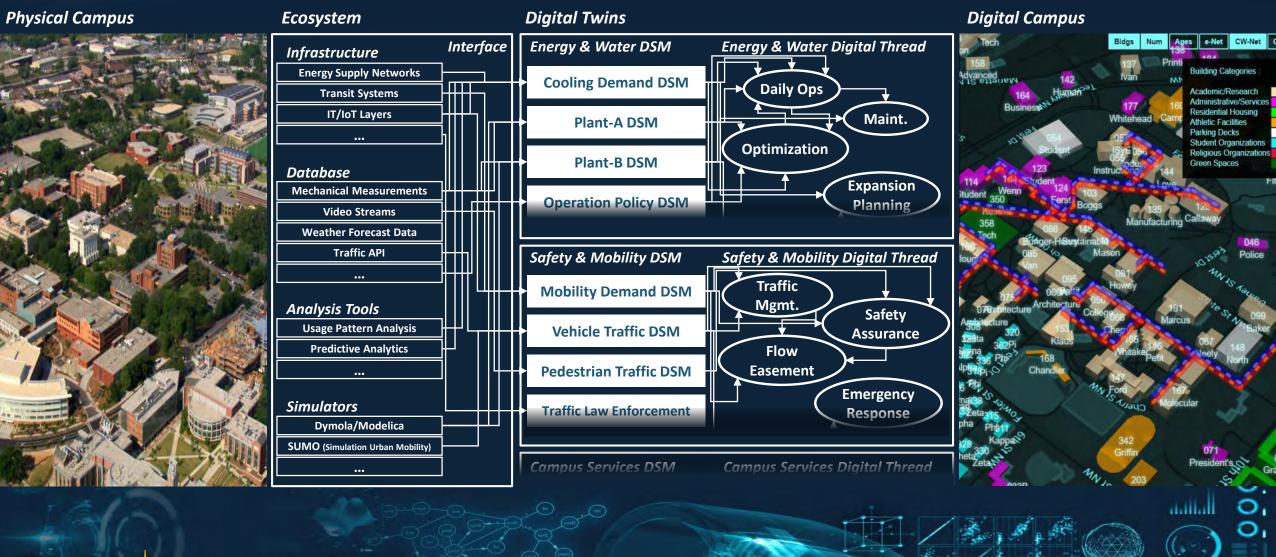
Smart Campus Data Analytics & Simulations support decision making at several horizons



Initiatives Align with Decision-Making Horizons



Digital Architecture (DA) of Smart Campus



Massive Data Available

Sensor data, collected and overlaid onto a digital facsimile of campus, to support infrastructure monitoring and planning

ASDL started with access to:

Meso-scale: 20,000 streams

Data from 150+ buildings & 2 plants Archived every 15 min for >7 years

- Energy & water metering
- People counters on newer entryways



CENTRAL HEATING & COOLING PLANT



Georgia Aerospace Systems Tech Design Laboratory ...cleaned, normalized, and mapped to drive:

Macro-scale Situational Awareness

- Campus-level visualizations
- Baselining for detecting degradation
- Modeling to project future scenarios



...soon to be supplemented with

Micro-scale data streams

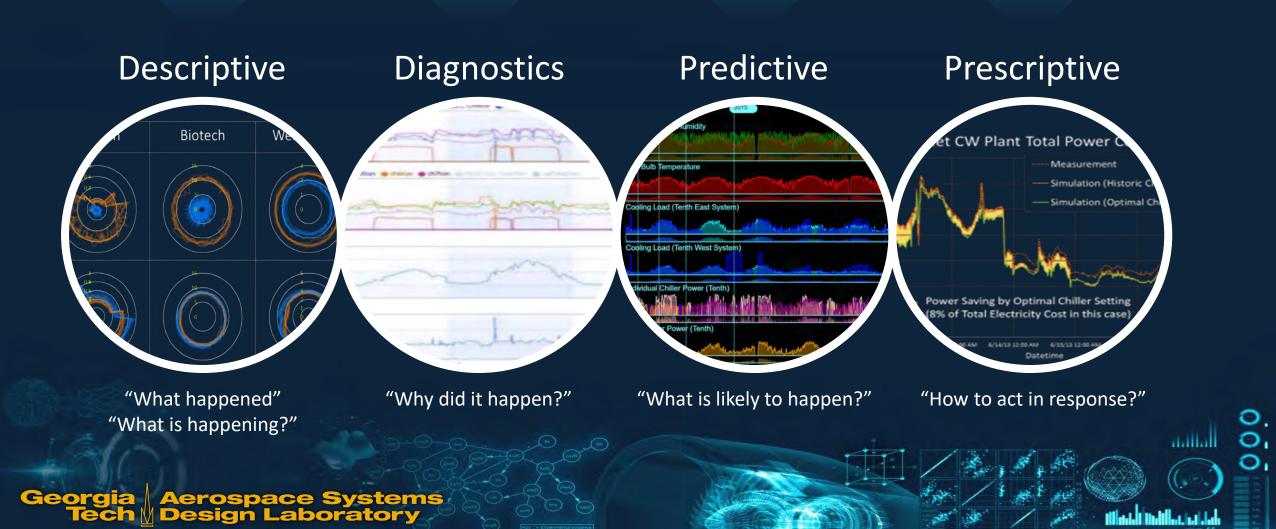
- ~10k end points per building
 - Internal temp, air quality
 - HVAC hardware states
- IoT sensors from maker spaces
- Mobile device locations, anonymized



Creating this digital version of campus required that ASDL:

Map & verify data sources Tap into data streams and databases Clean erroneous data Normalize data by weather and campus schedule Interpret data with help from GT Facilities Management engineers

DIGITAL TWIN - ANALYTICAL CAPABILITIES



DIGITAL TWIN - ANALYTICAL CAPABILITIES

Descriptive



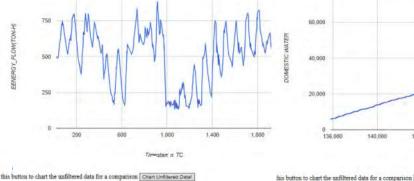
- Focus on summarizing and visualizing historical data to provide insights into the past and present
- Provides necessary context and foundation for further analysis
- Seeks answer about "What happened/What is happening?"

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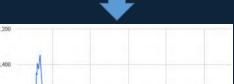
Descriptive Dia

Answers: "What Ha

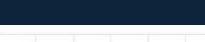


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Problem data

Corrected data

Anomaly treatment



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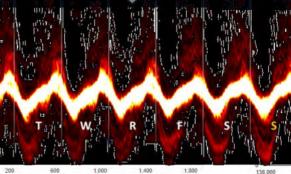


Preparing Big Data



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Clean data sets



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Visualizing Data for Infrastructure

Descriptive Dia

Diagnostics

Prescriptive

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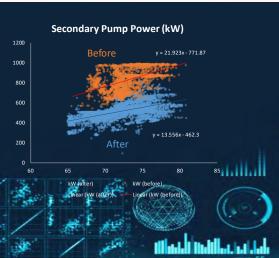
Answers: "What Happened?"

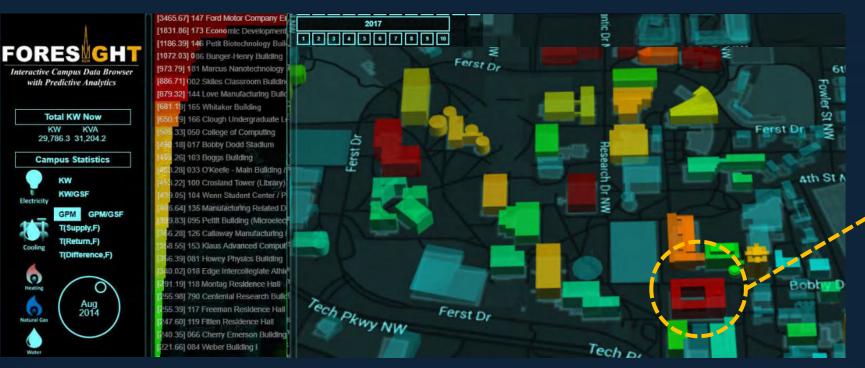
Value Propositions:

Predictive

- Common Operating Picture
 for team discussion
- Cost avoidance, identifying inefficiency









Visualizing Data From Many Domains

Descriptive Diag

Diagnostics

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Answers: "What Happened?"

Value Proposition:

Discovery of interactions between systems-of-systems

Predictive

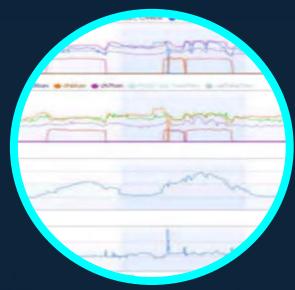




Safety (Incidents)

DIGITAL TWIN - ANALYTICAL CAPABILITIES

Diagnostics



- Drill down into the data, identify correlations and dependencies, and identify patterns for the purpose of identifying causes
- Seeks answer about "Why did it happen?"



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Descriptive Diagnostics

Prescriptive

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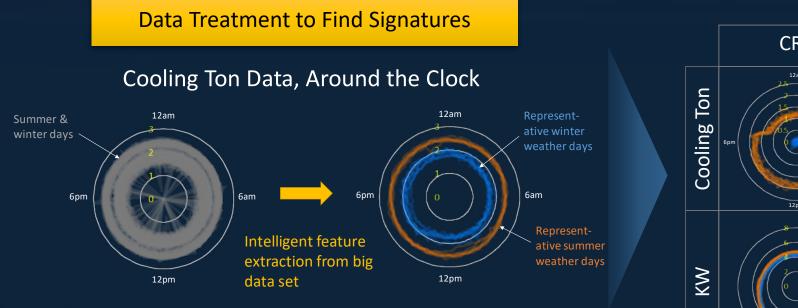
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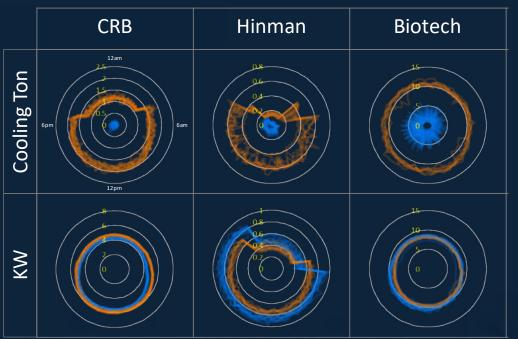
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Extracting Features

Answers: "Why did it happen?"

• SMEs identify performance signatures & root causes





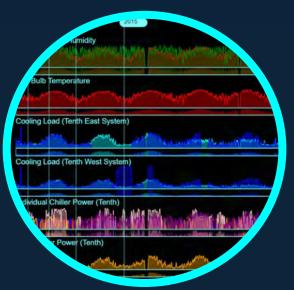
Predictive



DIGITAL TWIN - ANALYTICAL CAPABILITIES

- Leverages descriptive and diagnostic analytics to predict future trends
- Probabilistic in nature
- Seeks answer about "What is likely to happen?"

Predictive



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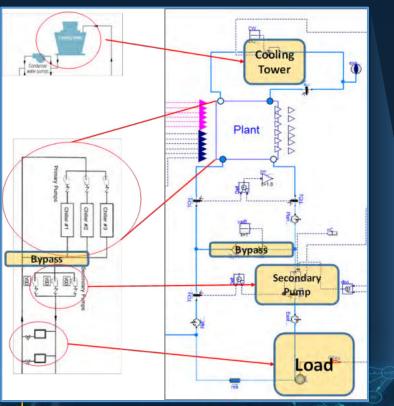
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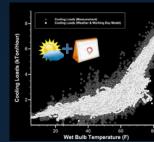
Predictive Modeling Capabilities

 Calibrated, physics-based models to forecast loads for chiller plant energy

Chiller Plant model (Modelica)



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AN COMPLETE

Neural Net Load

Model

Calculation by

M&S Tool (Dymola)

Past with Data ! Future with Model

Temperature & Humidity

Cooling Load for Sub-Network 1

Cooling Load for Sub-Network 2

Total Chiller Power Consumption

Load model (Neural Net)

Value Proposition:

• Forecasting future demand and system response

Example: Given a weather forecast...

...predict the energy usage of the central chiller plants to meet the thermal demand

FORESIGHT Predictive Campus Browser

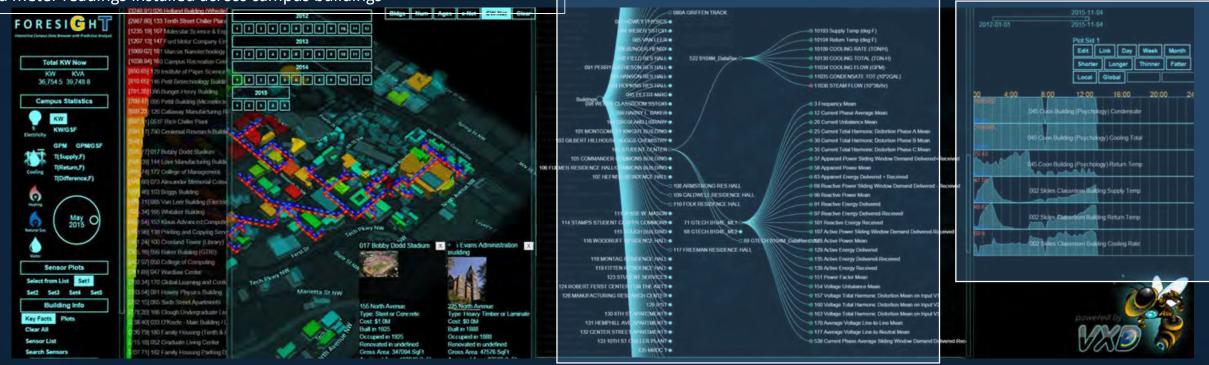
Interactive, visual-analytics based campus data browser, supporting real-time situational awareness, campus-level energy usage monitoring and model-based energy usage predictions, based on real time data streams



User can navigate through time/campus location and observe past energy performance trends for any building of interest

FORESIGHT Predictive Campus Browser

Real-time measurements and historic data queried from repositories maintained by campus facilities, and is sourced from sensor measurements and meter readings installed across campus buildings Interactive, visual-analytics based campus data browser, supporting real-time situational awareness, campus-level energy usage monitoring and model-based energy usage predictions, based on real time data streams



FORESIGHT Predictive Campus Browser

Interactive, visual-analytics based campus data browser, supporting real-time situational awareness, campus-level energy usage monitoring and model-based energy usage predictions, based on real time data streams



Comprehensive prediction capability for campus-wide energy usage that includes varying energy demand, accounts for total campus cooling load fluctuations, and utilizes weather forecast data

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DIGITAL TWIN - ANALYTICAL CAPABILITIES

- Integrates data, mathematical models (optimization, etc.) and business rules to advises on possible outcomes / recommends one or more courses of action
 → Support decision making
- Seeks answer about "How to act in response?"



Diagnostics Model-Based Scenario Exploration: Energy Demand vs. Supply

• Calibrated, physics-based models used to plan playbooks & responses

Descriptive

Given a weather forecast & cooling demand



...where does it make sense to curtail load?

...how should plant chillers be optimally staged?

Predictive



If a chiller went offline unexpectedly...

Value Propositions:

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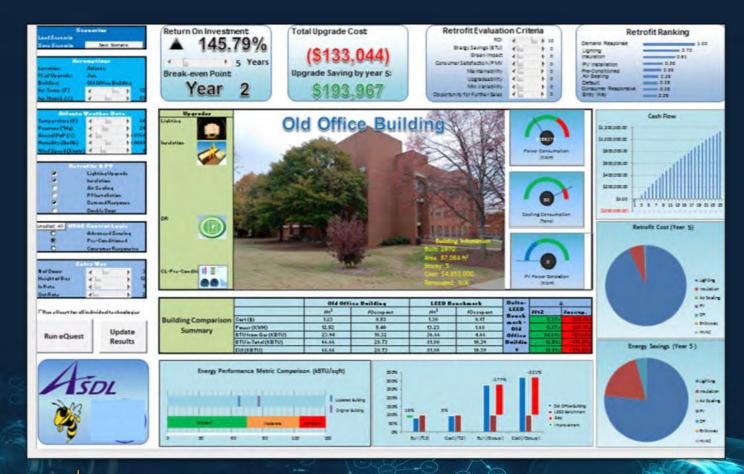
Optimization of efficiency Playbooks for resilient response Prescriptive

Answers: "How to act in response?"

Model-Based Scenario Exploration: Upgrade Portfolios

• Calibrated, physics-based models used to rank choices for retrofitting buildings

Descriptive



A tool to answer:

Diagnostics

• For an existing system in need of energy retrofits,

Predictive

- given a set of technology candidates,
- \rightarrow What should I invest in?
- → And what are their: ROI, Payback period, Energy savings, etc.
 - Either: absolutely, normalized per ft², compared to best in class, etc.

Georgia Aerospace Systems Tech Design Laboratory Answers: "How to act in response?"

Prescriptive

Digital Twin of a System-of-Systems



LEADER-X: Campus as Experimental Apparatus

Goal: Elevate ASDL's Smart Campus platforms into a Living Testbed for scientific study of socio-cyber-physical systems

Vision: a test bed with parallel facets:

"Digital twin" virtual test facility

Calibrated models, update with real conditions *Used to:*

- Plan out and run virtual experiments
- Create multiple instances in parallel, for comparison
- Prove-out proposed changes before implementing
- Create a baseline for understanding real measurements?

Questions to Address:

- With a campus as your test bed, how would you design living experiments to characterize the complex interactions between technology, infrastructure, and people? *Relevant to understanding Smart Cities, mobility systems, human space habitats, etc.*
 - How good are the current measurements of your experimental instrument? *i.e.,* What can already be done with current campus data sets, for selected case studies?
- What other data would you collect, i.e., what sensors to add would improve it?
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Used to:



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- Run parallel experiments on
 - Infrastructure
 - Community response shaping
- Calibrate the virtual apparatus

Thank You

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New Technolog

Material

Stability