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### Introduction

Autotune

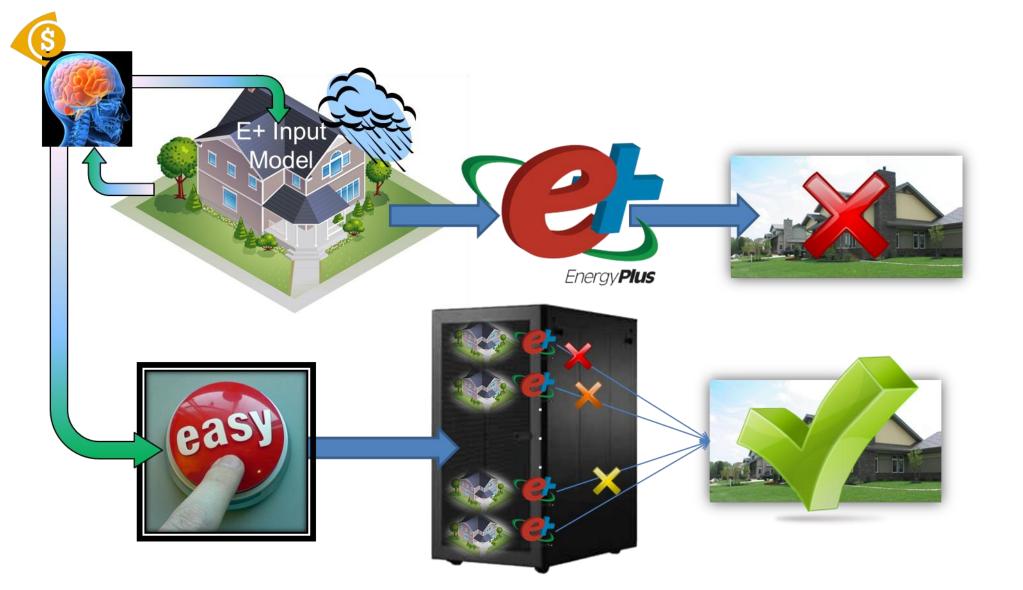
The major barriers to DOE's Building Technology Program (BTP) goals and the adoption of building energy modeling software are the user expertise, time, and associated costs required to develop a software model that accurately reflects reality. Typically, a building modeler uses the software tool they have most experience with to create the geometry of a building, layer it with detailed material properties, add equipment with anticipated operational schedules and run the simulation.

Machine Learning, and Supercomputing

Building Energy Model Calibration using EnergyPlus,

An E+ building model can easily have 3,000+ inputs for a normal residential building with very specific details that most energy modelers do not have data for.

The goal of the Autotune project is to save the time building modelers spend tweaking building input parameters to match ground-truth data by providing an "autotune" easy button for a standard desktop computer which intelligently adjusts model inputs.





## Approach

In order to achieve automatic tuning of buildings:

- Robotically emulated-occupancy 2800 ft<sup>2</sup> home with 269 channels of 15-minute sensor data
- Parametric E+ simulations, sensitivity analysis, and machine learning systems to characterize the effect of multi-variable perturbations on E+ simulations
- Intelligent adaptation of an existing E+ model to approximate sensor data for up to 96 E+ outputs

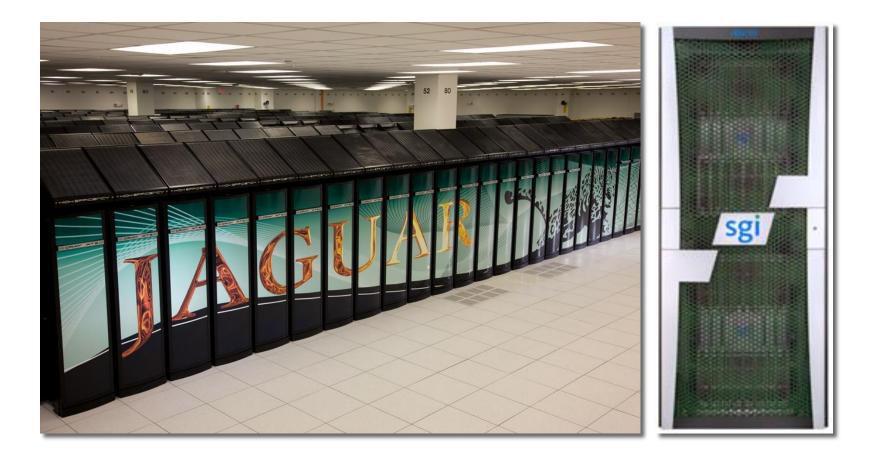
## **Machine Learning**

- Created MLSuite, an XML-based framework for flexible integration of 8+ machine learning algorithms via available tools such as Matlab (and several toolboxes), opensource libsvm for support vector machines, evolutionary computation with inspyred (on github), and others
- Multiple methods supported to alter input order and cross-validation methodologies for robust classification

#### Traditional model tuning vs. "Autotune"



Heavily instrumented ZEBRAlliance houses



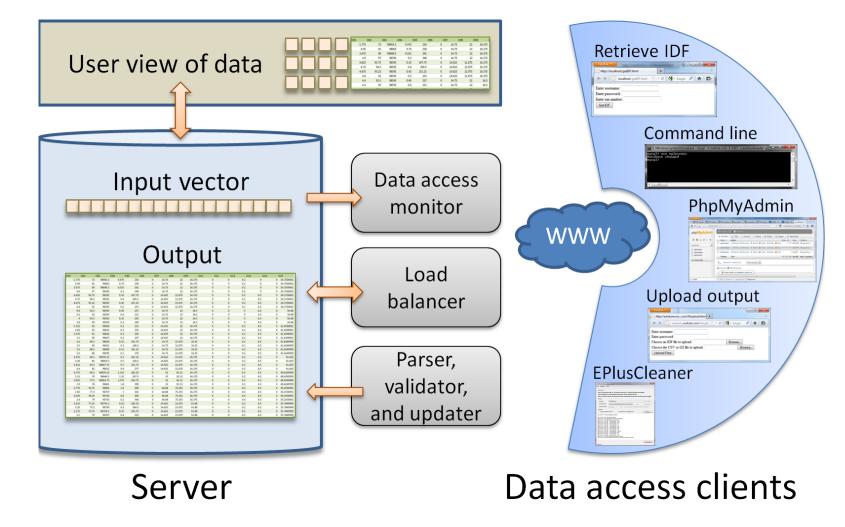
## Supercomputing and Big-data

- ORNL's Jaguar/Titan (299,000 cores), Frost (2,048 cores) and NICS's Nautilus (1,024 cores) systems in use
- Markov order sampling for increasing combinatorial effects of 156 important and widely used E+ inputs
- 4 most popular US buildings: residential home, warehouse, stand-alone retail, and medium office
- Anticipated ~270 TB of raw data for ~8 million simulations constituting 26.9 trillion data points
- Advanced database engines and architectures for hosting queryable data for power-users
- Parametric E+ data is available for public download now!



More information and data access at http://autotune.roofcalc.com

# ORNL's Jaguar and UT's Nautilus supercomputers



Public access to simulation data through use of advanced database engine technologies

...making software accountable.