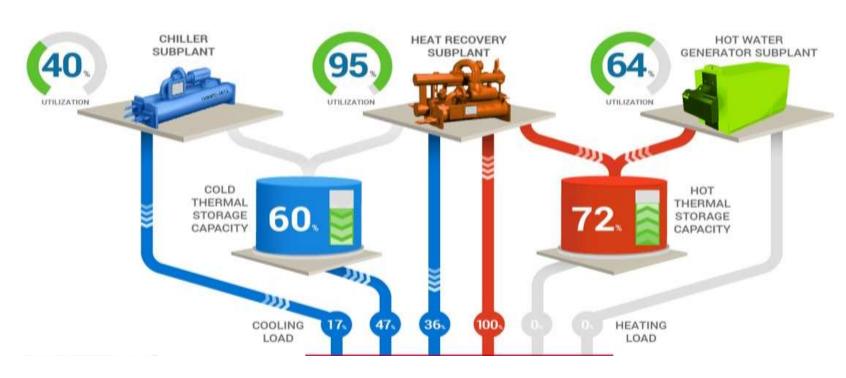


## Thermal microgrids can be highly cost effective part of a "total energy microgrid"

### What is a "thermal microgrid"?

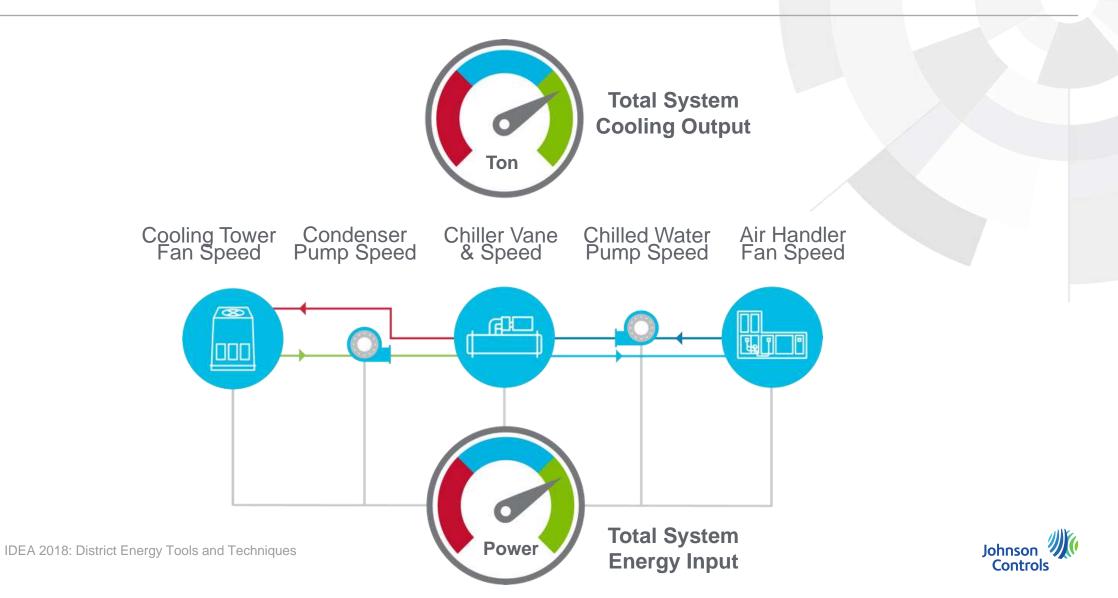
A district energy system that combines system-level optimized heating and cooling services, distributed generation, heat recovery, and electric power management through smart control technologies.





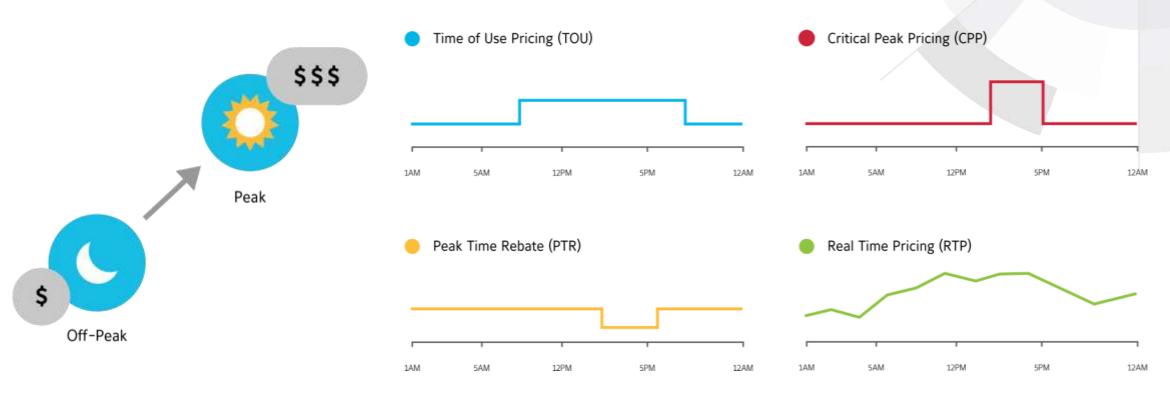
Key Principles: Maximizing efficiency at component level doesn't optimize at the system level

3



## Key Principles: Minimizing energy at each instant doesn't always optimize economics over time

A push for smarter utility grids is resulting in more rate structures that vary by time of day and more demand response incentive programs.





# Truly minimizing utility costs or carbon footprint while meeting priority #1: *reliable service delivery* is a complicated challenge

Every hour of every day plant engineers could consider *hundreds* of possible equipment combinations and set points

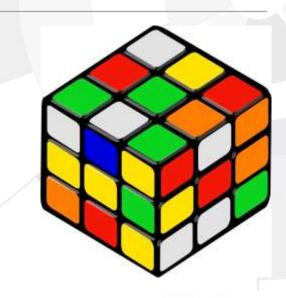
Imagine a simple plant with 5 chillers and 3 boilers. . .

- Number of boiler and chiller combinations =  $2^5 + 2^3 = 40$
- Select part-load level for each variable-speed "prime mover" based on multidimensional and nonlinear efficiency curves
- Auxiliary equipment like pumps and towers add layers of complexity.
- Ever-changing loads, weather and sometimes even time of use utility prices add to the challenge.

The number of decisions an operator *could* consider is staggering.

Imagine being able to optimize dispatch decisions every 15 minutes.

"That's something we just couldn't do manually."







Model Predictive Control (MPC) provides a math-based and physics-based "classic optimization" approach for control of thermal microgrids

#### Controls the variables based on sophisticated rules:

- **Reset Sequences of** Operation
- **Iterative Step & Wait**
- **Relational Control, and/or**
- Natural Curve Fit / Equal **Marginal Principle**

## **Supervisory Control** (Rules-based)

**Supervisory Control AND Optimization** 

> **Applies "classic" optimization** procedures with objective functions (e.g. minimize cost) subject to constraints (e.g. laws of physics)

Objective Function:

$$\begin{split} \min(J) &= \sum C_{ch} P_{ch} \Delta t + \sum C_{gas} F_{gas} \Delta t + \sum C_{nr} P_{nr} + \\ &\sum C_{pmp} P_{pmp} + \sum C_{dc} \max(P) + \sum C_{hr} P_{hr} \Delta t \end{split}$$

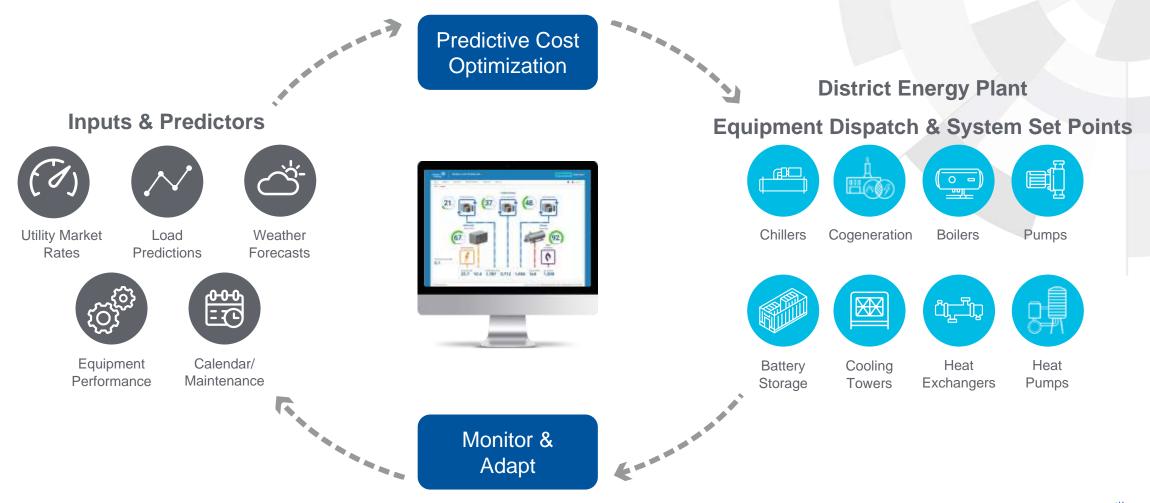
Equality Constraints:

$Q_{ch} + Q_{hr} = Q_{loadew_{-t_0}} \pm Q_{TEScw}$	$Q_{inv} + Q_{ir} = Q_{ioadnr_t_b} \pm Q_{TES_{inv}}$
$Q_{ck} + Q_{hr} = Q_{loadew_h} \pm Q_{7EScv}$	$Q_{inv} + Q_{iv} = Q_{loadinv_{-}i_1} \pm Q_{TES_{loc}}$
Inequality Constraints:	

$Q_{ch} \le \max(Q_{ch})$	$Q_{\scriptscriptstyle TES} \leq \max(Q_{\scriptscriptstyle TES})$	$Q_{\scriptscriptstyle TES} \leq \min(Q_{\scriptscriptstyle TES})$



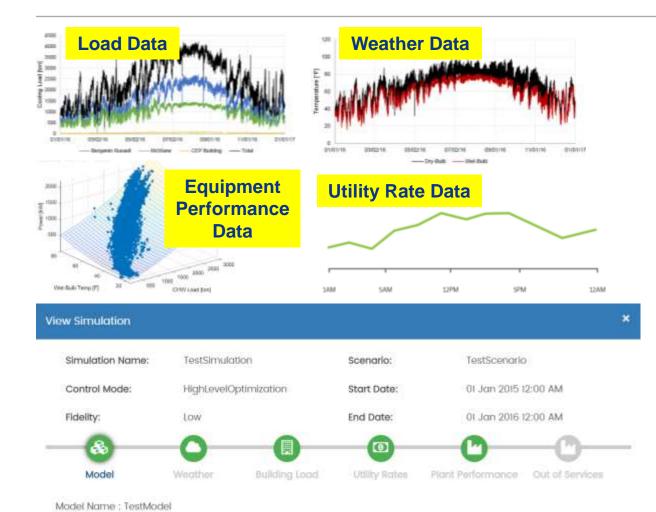
Predictive cost optimization algorithms respond to myriad ever-changing inputs







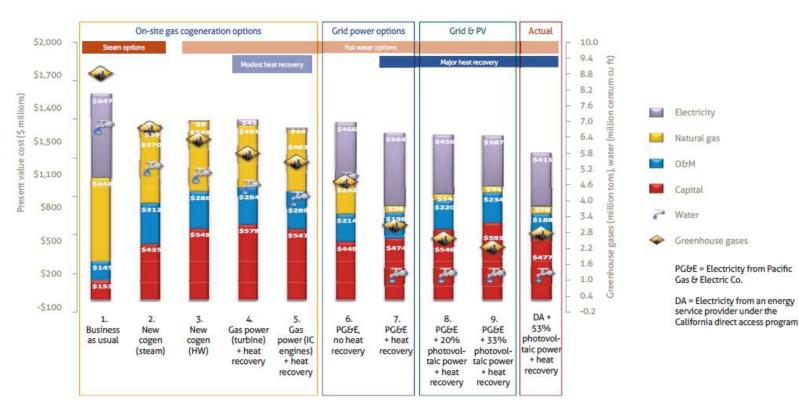
# Emerging District Energy "Planning Tools" for detailed energy cost simulation



- Develop 8760 models of utility usage, peak demand, and costs
- Perform what-if scenarios based on actual operating conditions
- Inform design decisions to reduce lifecycle operating costs
- Helps to right-size plant equipment to reduce first cost
- Supports annual budgeting and ongoing O&M decisions
- Enables measurement & verification (M&V) of savings for project investments



Using Model Predictive Control (MPC) as a simulated control strategy enables robust lifecycle cost and GHG analysis to select best combination of properly-sized equipment



Stagner (2016) "Stanford University's Fourth Generation District Energy", IDEA District Energy Quarterly

"I knew when we had a model that would show how the system would work every hour of the year, no engineer could look at it and cast doubt...

...In the end, the design withstood peer reviews from external consultants, professors, and very knowledgeable board members. It breeds a lot of confidence to actually have the data to back up a recommendation."

- Joe Stagner Executive Director Department of Sustainability and Energy Management Stanford University



# **Kent State University, OH**

- 2<sup>nd</sup> largest in OH public university system: 43,000 enrollment
- Kent campus district energy plants serve 110 buildings, 6 million sq ft
- 12MW cogeneration system (2 gas turbines)
- 13,000 tons of chilled water capacity across 7 separate plants serving 3 loops
- Combination of electric centrifugal, steam turbine, and absorption chillers
- 4 boilers, 2 duct burners, and 1 heat recovery steam generator



# Thermal Microgrid Optimization at Kent State University's Kent Campus

Completion of CPO system commissioning and transition to Auto mode began in Dec 2017

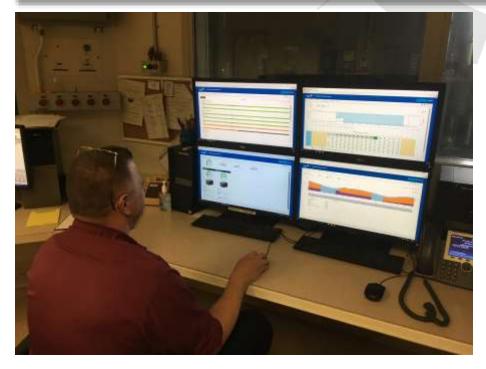
System continuously monitors ~1,000 input variables:

- Current loads and operating conditions
- Seven-day weather forecasts
- Real-time market prices for economic load demand response (ELDR)
- Electricity, water, and natural gas rates
- Campus events that may impact loads
- Equipment maintenance schedules

Dispatches ~150 control decisions every 15 minutes to *minimize cost*, not just energy use, over 7-day horizon

Expected to help KSU generate over **\$1 million** annually in utility savings and added ELDR revenues "This system brings together all seven separate chiller plants into one platform that we can view easily, simultaneously, and from any remote location. Then it finds the optimal solution across the system, and accounts for changes every 15 minutes. That's something we just couldn't do manually...

- Dr. Frank Renovich, Associate Director of Energy Operations



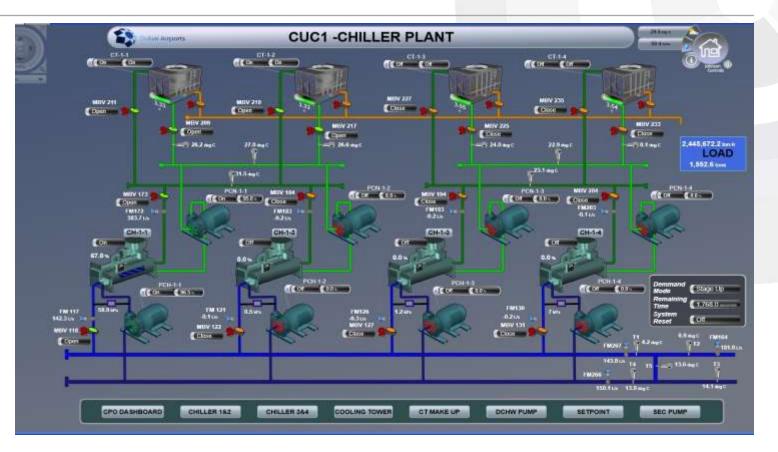
Controls



### District Energy for Warmer Climates

# Al Maktoum International Airport CUC-1 Plant, United Arab Emirates

- Chilled water district energy
- 8000 ton refrigeration capacity including 4 water-cooled chillers
- Constant Primary Flow / Variable Secondary Flow
- Baseline Annual Average System
  Efficiency = 1.018 kW/ton





### District Energy for Warmer Climates Al Maktoum International Airport CUC-1 Plant, United Arab Emirates







### District Energy for Warmer Climates SCOPE International AG, Chennai, India

- Chilled water plant serving office environment
- 1,725 Ton-R plant including both water-cooled and aircooled chillers of differing vintages
- Variable primary flow configuration
- Baseline Annual Average Plant-wide Efficiency = 1.25kW/ton
- SCOPE International sought to upgrade two chillers, add variable speed drives, and implement controls to holistically optimize:
  - $\circ$  chiller selection,
  - VSD chiller speed,
  - o chilled water and condenser water pump speed, and
  - $\circ$  cooling tower fan speed





### District Energy for Warmer Climates SCOPE International AG, Chennai, India

