

22
expo
acaire
2023



District Energy

LATAM Conference 2023

SEP 27 - 29 | CARTAGENA | CO



Cartagena

| 27, 28 y 29 de septiembre |

AI Applied to Chiller Plant Control

Rui Fonseca
Daikin Solutions Hub - Corporate Planning

Sept 28, 2023



Rui Fonseca

**Corporate Planning
Daikin Solutions Hub,
Singapore**



Frases como/Phrases like:

“Hable bien, como un varón” (“You should talk like a man”)

“Si fueras más femenina te iría mejor en la vida” (“You should act more feminine.”)

Son microagresiones (are microaggressions)



*“Limiting CO2 emissions-
driven global warming is one
the biggest challenges of our
time”*



COP27
SHARM EL-SHEIKH
EGYPT 2022

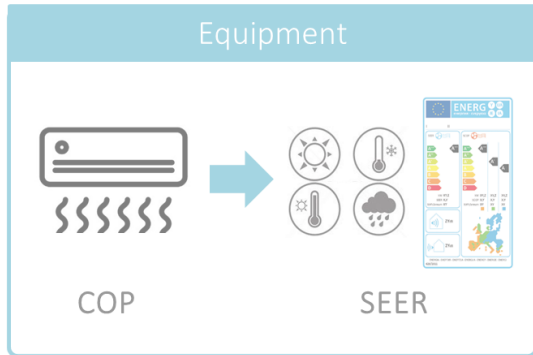


- Individual Pieces of Equipment are efficient in their own rights
- Variable Speed Drives (VSD) have been introduced to improve the part-load efficiency of chillers, pumps, and cooling towers

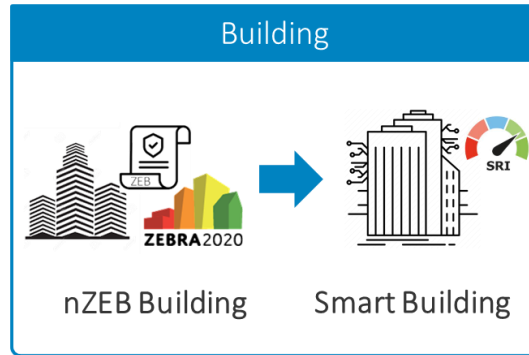
However

A **'conductor'** is needed to carefully consider each aspect of the system and how tweaking or adjusting one variable will affect the building operation as a whole

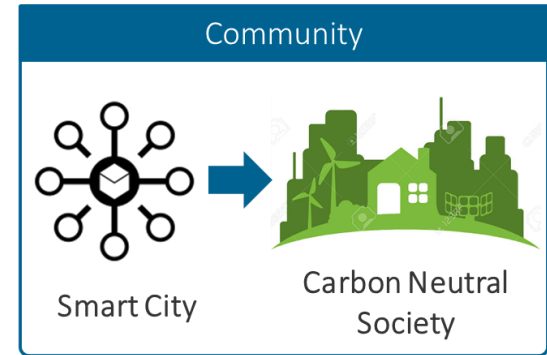
Paradigm shift how we design, build, operate & maintain buildings and their relationship on the wider community



- Individual product assessment
- Energy efficiency measured at rated points



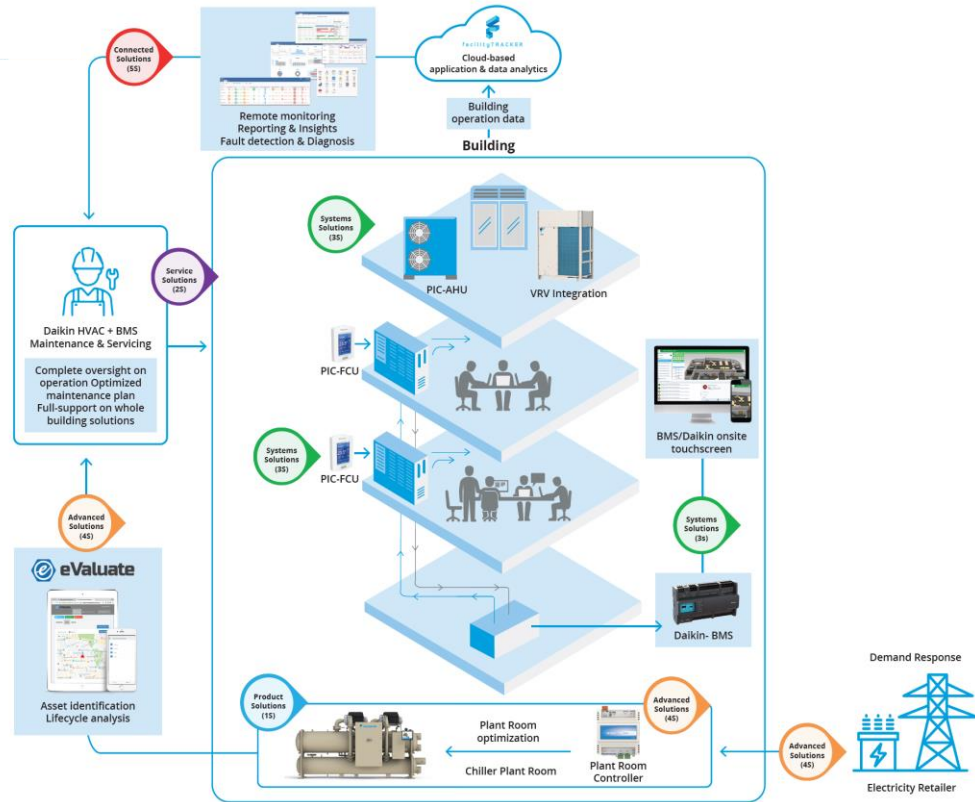
- Consider environmental impact of complete solution, with real operating conditions
- Connectivity and interoperability between building systems



- Integrated communication platforms
- Asset management to guarantee a seamless and efficient operation

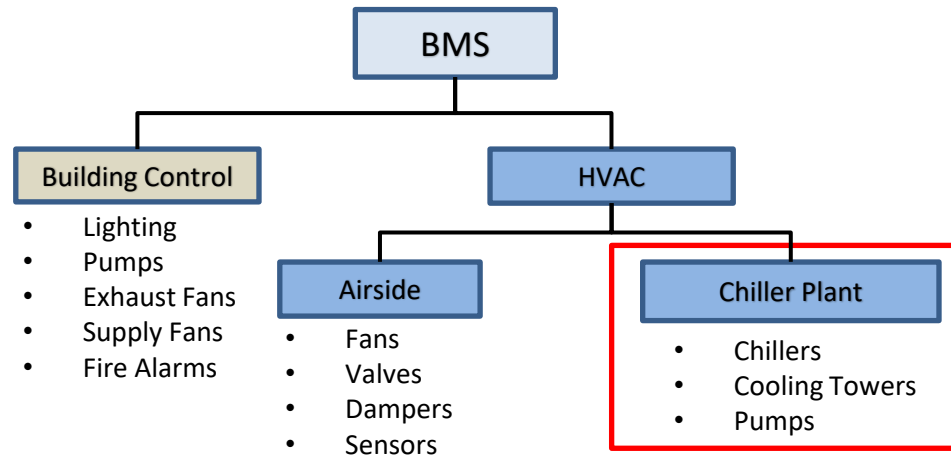
Where does the chiller plant sit within the operation?

- Chiller plant is the **'heart beat'** of the overall building operation
- Optimisation and Energy Performance **starts here**
- Downstream systems can operate optimally if **consistent performance is achieved** from the chiller plant



Typical building management system design

- Standard Building Management Systems (BMS) are **generic in nature and not equipment or system specific**
- BMS focus on bringing in all disparate building systems together and the whole of building operation, based on **predefined rules of operation**
- Not usually built to **compare design operational data** vs actual onsite performance
- Chiller Plant is ‘just another’ system as part of the overall Building Management System



How to achieve an optimized chiller plant room?



Intelligent Plantroom Management - Beyond Traditional BMS approach -

- Real-time analytics
- Diagnostics and M&V system
- Continually readjusting of chiller plant
- Optimal performance



PLANT MEASUREMENT
AND VERIFICATION



PLANT PERFORMANCE
MONITORING



PLANT DIAGNOSTICS
AND REPORTING



PLANT CONTINUOUS
COMMISSIONING AND TUNING



PLANT CONTROL AND
AUTOMATION



PLANT
OPTIMISATION

Comfort:

Ensures the right operating conditions

Efficiency:

Optimized efficiency based on equipment actual performances & building profile & needs

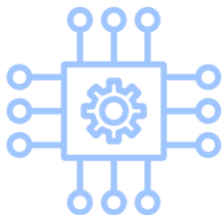
Cost:

Lowers operation & maintenance costs

Intelligent Plantroom Management

- The benefit of AI in Chiller Plants -

- Advanced mathematical optimization algorithms, primarily for 3 purposes
 - ✓ **Equip. data models:** Machine Learning to learn and minimize error to continuously calibrate the model
 - ✓ **Chiller Plant Optimization:** Finding optimal control performance (Temp. setpoints, target loads, chiller combinations...)
 - ✓ **Alerts:** Predictive fault detection, based on deviations from target efficiency



AI-Machine Learning
Backed Data Models



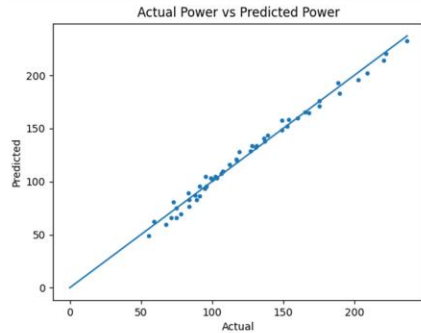
AI-Automatically Generated
Control Algorithms



Automated Alerts if Operation
Deviate from Target Efficiency

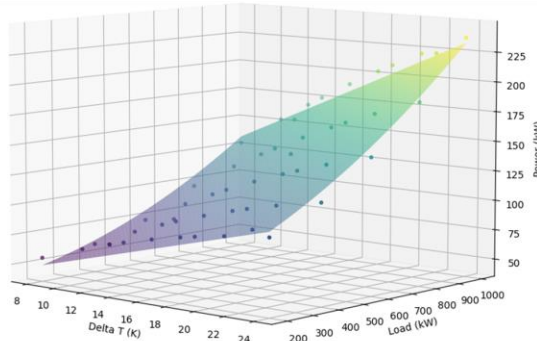
AI Machine Learning Strategies - Equipment Data Models -

Model



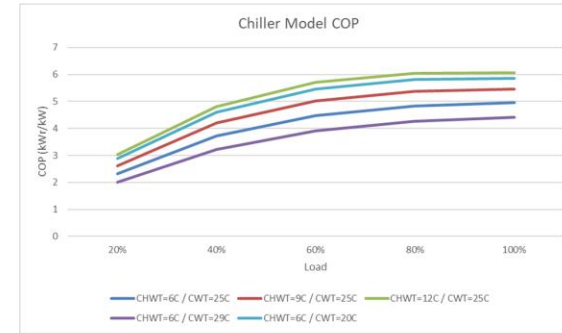
Optimization algorithm minimizes the difference between the actual power & model predicted power

Mathematical



Results in a mathematical model predicting the power as a function of cooling load, LCHWT & ECWT

Actual



The COP curves are then derived

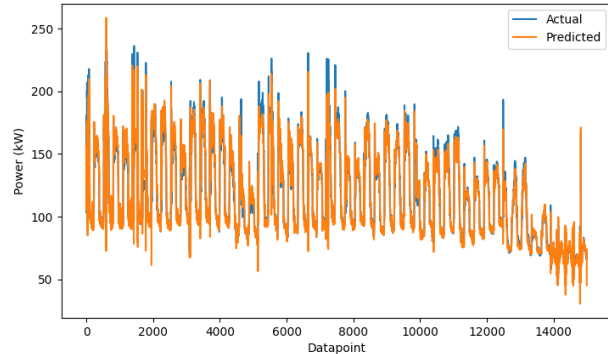
AI Machine Learning Strategies - Equipment Data Models -

Learning chiller performance:

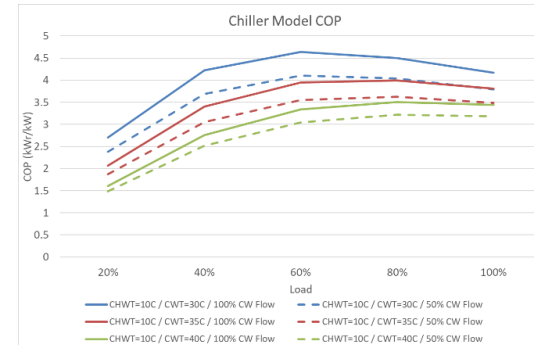
$$Power_{chiller} = P(\text{Load}, T_{ECW}, T_{LCHW}, Flow_{CW})$$

Table: Chiller learning results with site data

	Site A	Site B		Site C		
Chiller n°	2	1	2	3	1	2
MAE (kW)	4.1	2.6	2.8	3.9	11.0	10.6
MAE/Mean	4.6 %	2.4 %	2.4 %	3.4 %	6.5 %	6.4 %
R-squared	0.949	0.990	0.984	0.979	0.949	0.938



Site B, Chiller 1 predicted power compared to the actual measured power



Site B, Chiller 1 predicted COP from chiller model for a range of conditions

AI Optimization Strategies

- Mathematical optimization of plant room -

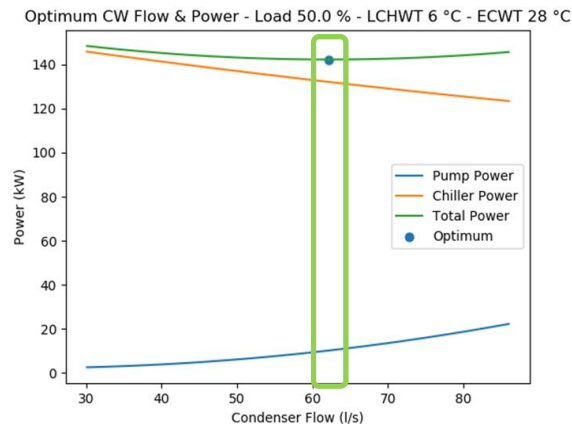
Minimize $Load_i, Flow_{CW,i}, T_{LCT}$ $Power_{plant} = \sum_{i=1}^n P_{chiller,i} + \sum_{j=1}^m P_{cw\ pump,j} + \sum_{k=1}^p P_{coolingtower,k}$

	Optimization Strategy	Achievable in PLC Control	Achievable in iPlant Manager	Estimated Savings (with iPM implementation)
(1) Optimization of plant room	Chilled Water Leaving Temperature Setpoint Optimization	Yes*	Yes	<ul style="list-style-type: none"> 1 to 1.5% savings in kW/TR in fixed speed chillers 2 to 3% savings in kW/TR in VFD chillers
	Condenser Water Entering Temperature Setpoint Optimization	Yes*	Yes	<ul style="list-style-type: none"> 1 to 1.5% savings in kW/TR in fixed speed chillers 2 to 3% savings in kW/TR in VFD chillers
	Smart Variable Flow	Yes*	Yes	<ul style="list-style-type: none"> Variable Primary Flow 3.0% Variable Secondary Flow 4.0%
(2) Smart Sequencing	Smart Sequencing	No	Yes	<ul style="list-style-type: none"> 2 to 4% typical savings for the whole plant
	Optimized Start/Stop	Partially*	Predictive Optimized Start/Stop and Staging Control	<ul style="list-style-type: none"> 1 to 1.5% typical savings for the whole plant

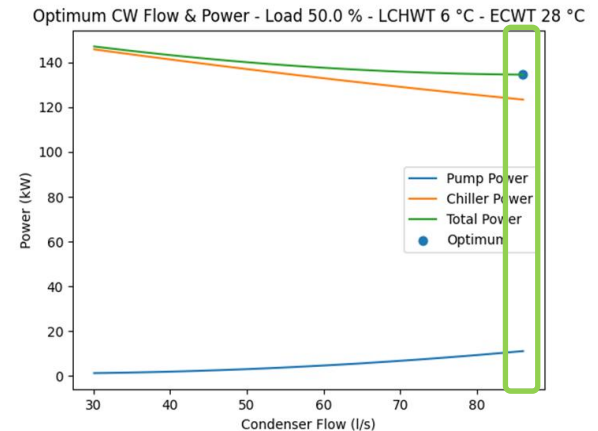
AI Optimization Strategies

- Mathematical optimization of plant room -

- Maximize energy & cost savings by determining the optimum operation to reduce the “lift”, or head pressure
 - ✓ Condenser Water Entering Temperature Setpoint Optimization
 - ✓ Chilled Water Leaving Temperature Setpoint Optimization
 - ✓ Smart Flow



Chiller/Pump set 1: Pump rated Power 22 kW
Optimal CW Flow 60 L/s

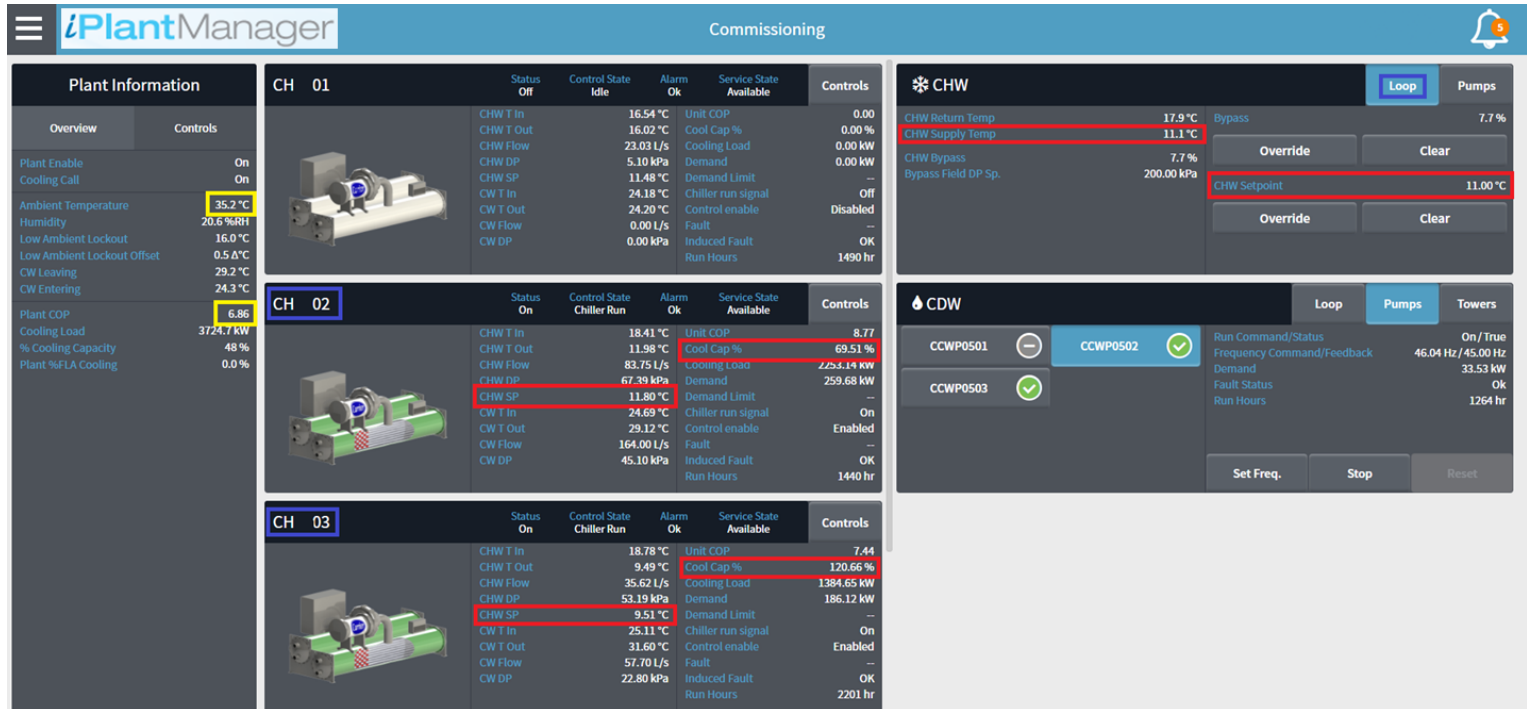


Chiller/Pump set 2: Pump rated Power 14 kW
Optimal CW Flow 85 L/s

AI Optimization Strategies

- Smart Sequencing -

- Select the most efficient load point for each chiller and chiller combination based on each chiller performance curves at different CHWT and CWT



The screenshot displays the iPlantManager interface for Commissioning. On the left, the Plant Information panel shows various parameters: Overview (Plant Enable: On, Cooling Call: On), Controls (Ambient Temperature: 35.2°C, Humidity: 20.6%RH, Low Ambient Lockout: 16.0°C, Low Ambient Lockout Offset: 0.5Δ°C, CW Leaving: 29.2°C, CW Entering: 24.3°C), Plant COP: 6.86, Cooling Load: 3724.7 kW, % Cooling Capacity: 48%, and Plant %FLA Cooling: 0.0%.

The main area shows three chiller control panels (CH 01, CH 02, CH 03) and two water control panels (CHW, CDW).

CH 01

Parameter	Status	Control State	Alarm	Service State	Controls
CHW T In	Off	Idle	Ok	Available	Unit COP: 0.00
CHW T Out					Cool Cap %: 0.00%
CHW Flow					Cooling Load: 0.00 kW
CHW DP					Demand: 0.00 kW
CHW SP					Demand Limit: --
CW T In					Chiller run signal: Off
CW T Out					Control enable: Disabled
CW Flow					Fault: --
CW DP					Induced Fault: OK
					Run Hours: 1490 hr

CH 02

Parameter	Status	Control State	Alarm	Service State	Controls
CHW T In	On	Chiller Run	Ok	Available	Unit COP: 8.77
CHW T Out					Cool Cap %: 69.51%
CHW Flow					Cooling Load: 2293.14 kW
CHW DP					Demand: 259.68 kW
CHW SP					Demand Limit: --
CW T In					Chiller run signal: On
CW T Out					Control enable: Enabled
CW Flow					Fault: --
CW DP					Induced Fault: OK
					Run Hours: 1440 hr

CH 03

Parameter	Status	Control State	Alarm	Service State	Controls
CHW T In	On	Chiller Run	Ok	Available	Unit COP: 7.44
CHW T Out					Cool Cap %: 120.66%
CHW Flow					Cooling Load: 1384.65 kW
CHW DP					Demand: 186.12 kW
CHW SP					Demand Limit: --
CW T In					Chiller run signal: On
CW T Out					Control enable: Enabled
CW Flow					Fault: --
CW DP					Induced Fault: OK
					Run Hours: 2201 hr

CHW

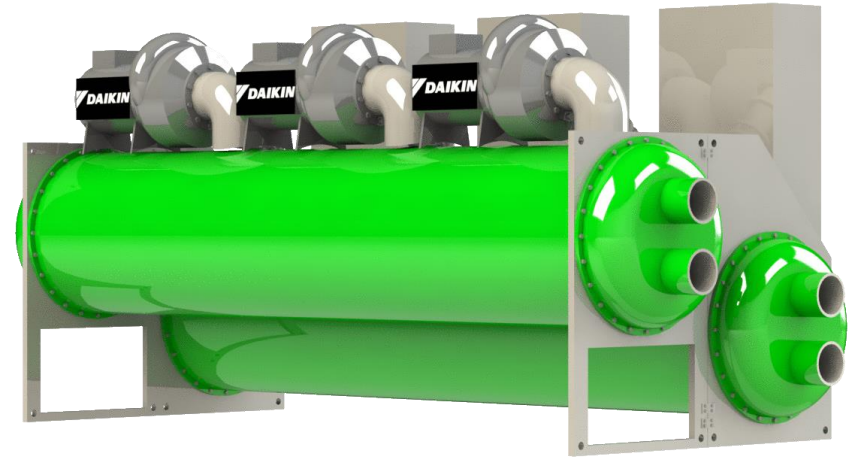
Parameter	Value	Control State	Alarm	Service State	Controls
CHW Return Temp	17.9°C	Bypass			7.7%
CHW Supply Temp	11.1°C				
CHW Bypass	7.7%				
Bypass Field DP Sp.	200.0 kPa				
CHW Setpoint	11.00°C				

CDW

Parameter	Value	Control State	Alarm	Service State	Controls
CCWP0501		Off			
CCWP0502		On			
CCWP0503		On			
Run Command/Status	On/True				
Frequency Command/Feedback	46.04 Hz / 45.00 Hz				
Demand	33.53 kW				
Fault/Status	Ok				
Run Hours	1264 hr				

AI-based predictive models - Fault Detection & Diagnosis -

- Autonomously defines the “Rating” in which the units are operating
 - ✓ **GREEN, YELLOW and RED** operating areas
 - ✓ Units are still running but underperforming with respect to design conditions

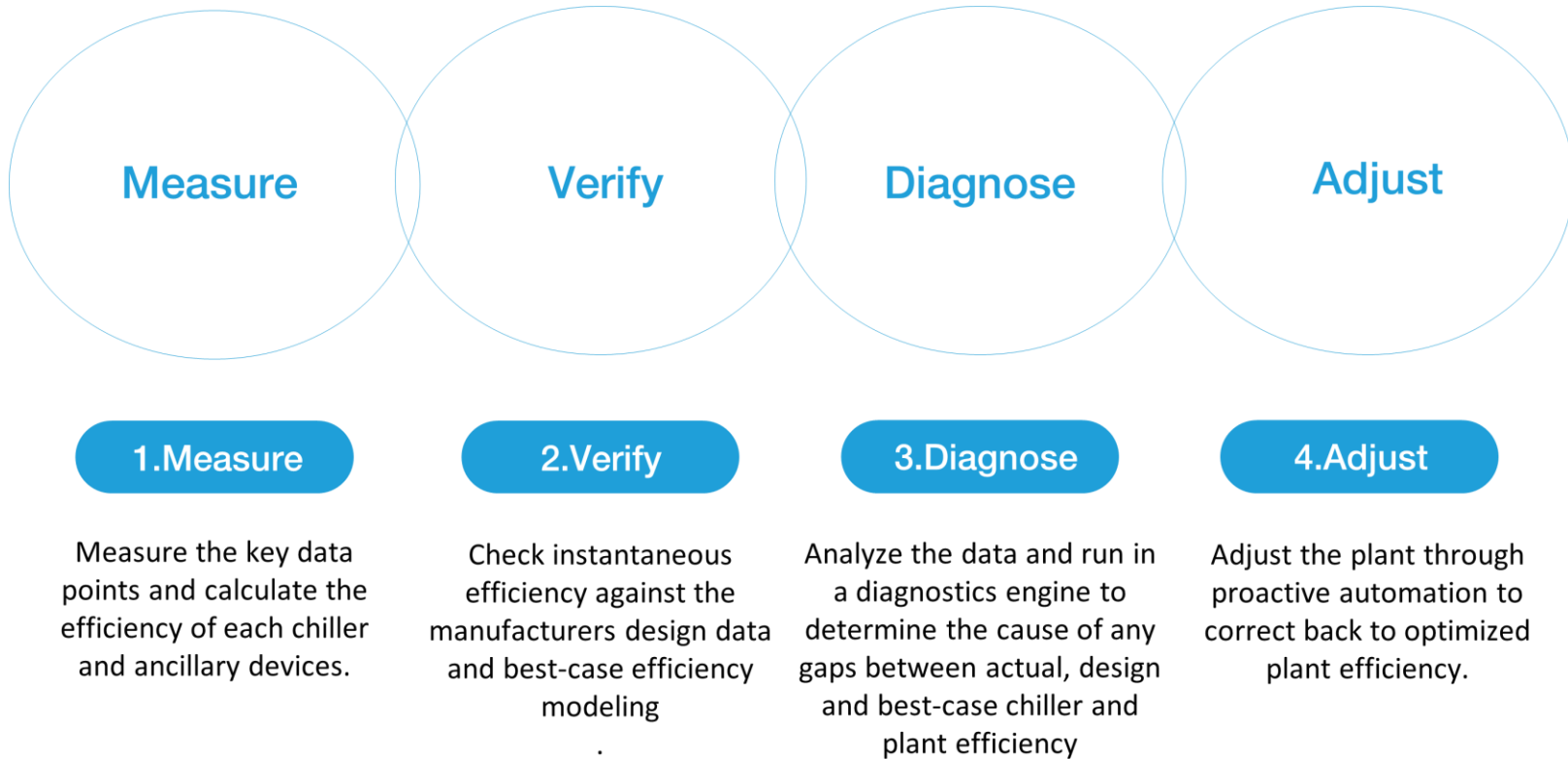


- **Diagnostics & Analytics**
 - ✓ Detects variables from the plant and provide info on the system status

	Month To Date			Year To Date			
	Current	Previous	Variation	Current	Previous	Variation	
Unit 1:Chiller1 (Water Cooled Flooded Chiller)							
Total Cooling Produced	[kWh]	106,610	36,668	191%	106,610	72,733	47%
Total Electricity Consumed	[kWh]	21,513	7,660	181%	21,513	26,989	-20%
Total Electricity Cost	[Dollar]	4,195	1,494	181%	4,195	5,263	-20%
Average Unit Efficiency		1.6	6.5	-75%	1.6	0.0	0%
Specific Energy Cost	[c/kWh]	3.9	4.1	-3.4%	3.9	7.2	-45.6%
Diagnostics							
Chiller Short Cycling		19%					
Power Surging		18%					

Intelligent Plantroom Management

- In summary -



Case Study University, Singapore



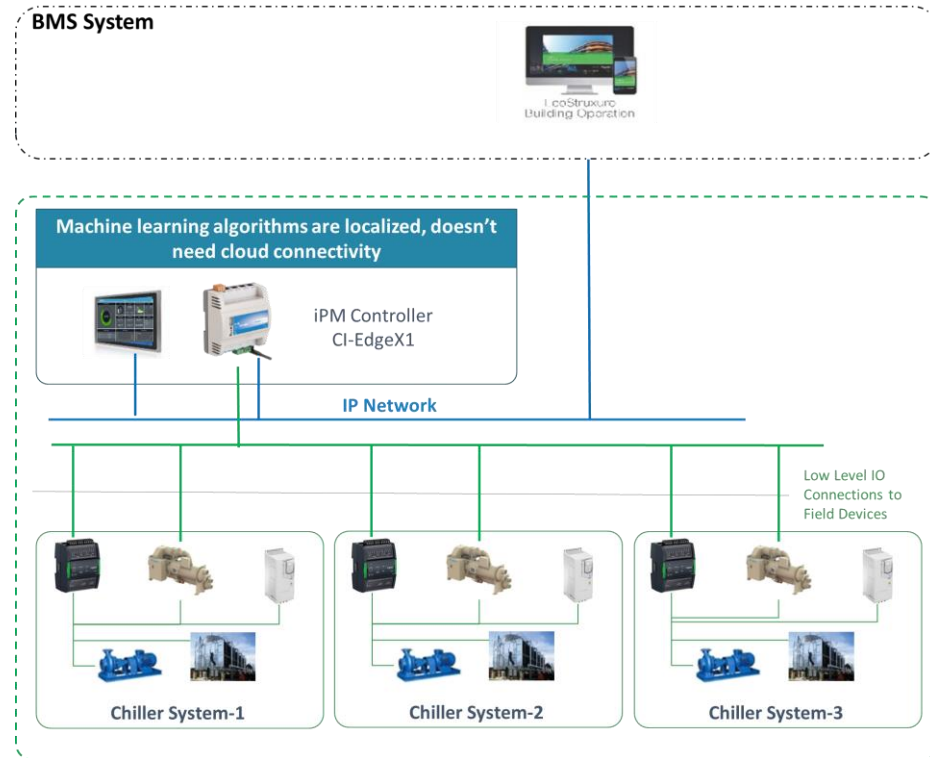
Case Study – University, Singapore

Chiller Plant overview

- **Chiller:** 3x 250 RT
Water cooled, Magnetic bearing chillers
- **CHWPs:** 3x 18.5 kW (40.6L/s, 30m head)
- **CDWPs:** 3x 18.5 kW (55.3L/s, 25m head)
- **Cooling towers:** 3x 11kW (55.3L/s)

Solution scope:

- **Controller:** 1 x Master iPlant Manager for controlling and optimising all plant equip.
- Connection by ModBus/BACnet to controllable equipment
- Seamless integration with **SE BMS on site**

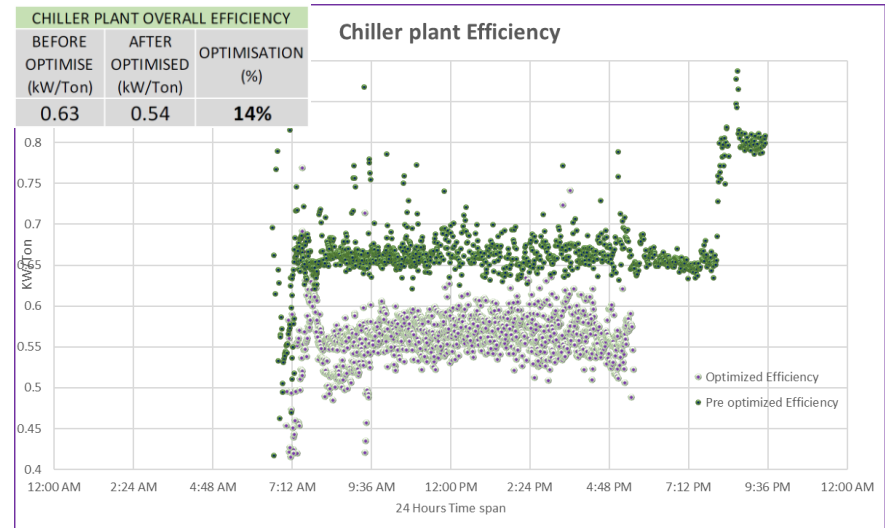


Case Study – University, Singapore

		BMS	iPM
1	Chiller operation		
	Day time operation	2 Nos. Of Chiller Based on Run time Priority	Maximum 2 Nos Based on Runtime priority
	Evening / Night operation	1 Nos.Of Chiller Based on Run time Priority	Max 2 Nos, Most Efficient chiller from priority chiller 1 and 2, if only one chiller to operate. Based on chiller performance analytics
2	Chilled water pump		
	Day time operation	2 Nos Of Pump	2 or 1 Nos. of Pump based on Cooling load, flow and Differential pressure requirement
	Evening / Night operation	1 Nos of Pump	2 or 1 Nos. of Pump based on Cooling load, flow and Differential pressure requirement
3	Condenser water pump		
	Day time operation	2 Nos Of Pump	3 or 2 Nos. of Pump based on condenser water system Optimization analytics
	Evening / Night operation	1 Nos of Pump	3 or 2 Nos. of Pump based on condenser water system Optimization analytics
4	Cooling tower		
	Day time operation	2 Nos Of Cooling tower	3 or 2 Nos. of Cooling tower based on heat rejection, Condenser water return temperature and cooling tower approach
	Evening / Night operation	1 Nos of cooling tower	3 or 2 Nos. of Cooling tower based on heat rejection, Condenser water return temperature and cooling tower approach

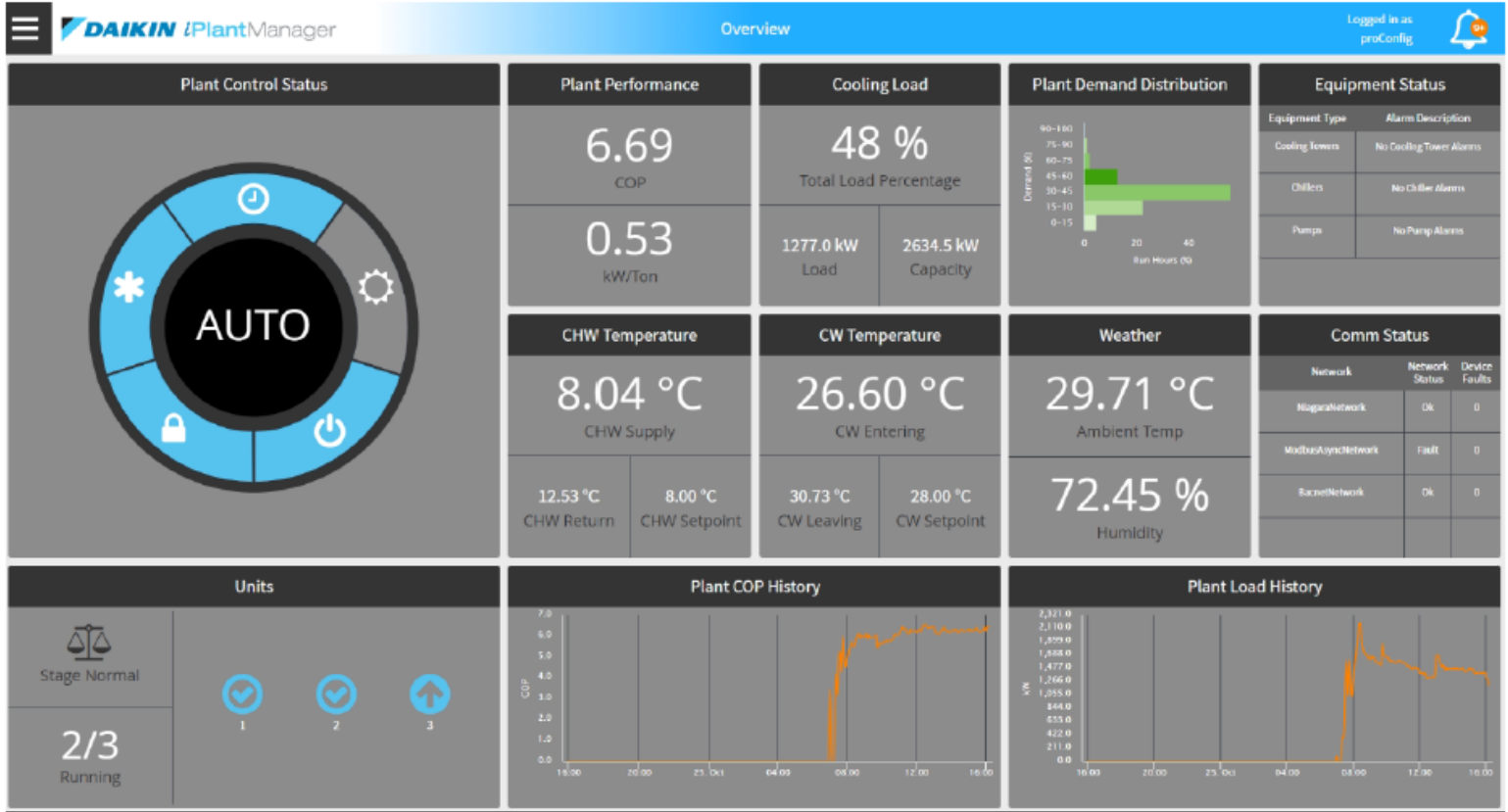
Case Study – University, Singapore

- iPM assumes full plant control and optimization of the plant
- It utilizes on-board, real-time analytics, diagnostics and M&V system to continually readjust chiller plant for optimal performance
- **Optimized performance:** 0.54 kW/RT
- 14% savings in average, with up to 22% reduction on energy consumption during normal operation (lower savings during school holiday periods)



ROI: 14 months

Case Study – University, Singapore



Case Study – University, Singapore

Learning chiller data models

Unit Learning

Heat Balance Cutoff: 10 %

Model Generation Interval: 24 hrs

Estimated Pump Power: 33.00 kW

Model Chart

Chiller Model based on User-Input Coefficients



Chiller Model w/ CW Flow

Data Sampling Interval: 5 min | Model Type: Manual

Dataset Start time: 05/08/2022 07:00 am

Dataset End time: 01/10/2022 07:00 am

Model Generation Interval: 24 hr

Heat Balance Cutoff: 10 %

Prioritized Data Time Window: 12 mo

A0	7.604085146256	A1	0.000000011323
A2	0.044097911449	A3	0.043345676995
A4	0.000072734419	A5	0.000710874761
A6	-0.258181316120	A7	0.002191506001
A8	0.000000000025	A9	-0.000235040099

Generate Model | Save

User-defined Model



Model Coefficients

A0	7.604085146256
A1	0.000000011323
A2	0.044097911449
A3	0.043345676995
A4	0.000072734419
A5	0.000710874761
A6	-0.258181316120
A7	0.002191506001
A8	0.000000000025
A9	-0.000235040099

User-defined Model

A0	2.480991999170993	A3	0.024529525346587
A1	5.286475530077219	A4	0.000050271940591
A2	0.008448548355203	A5	0.002888826874455

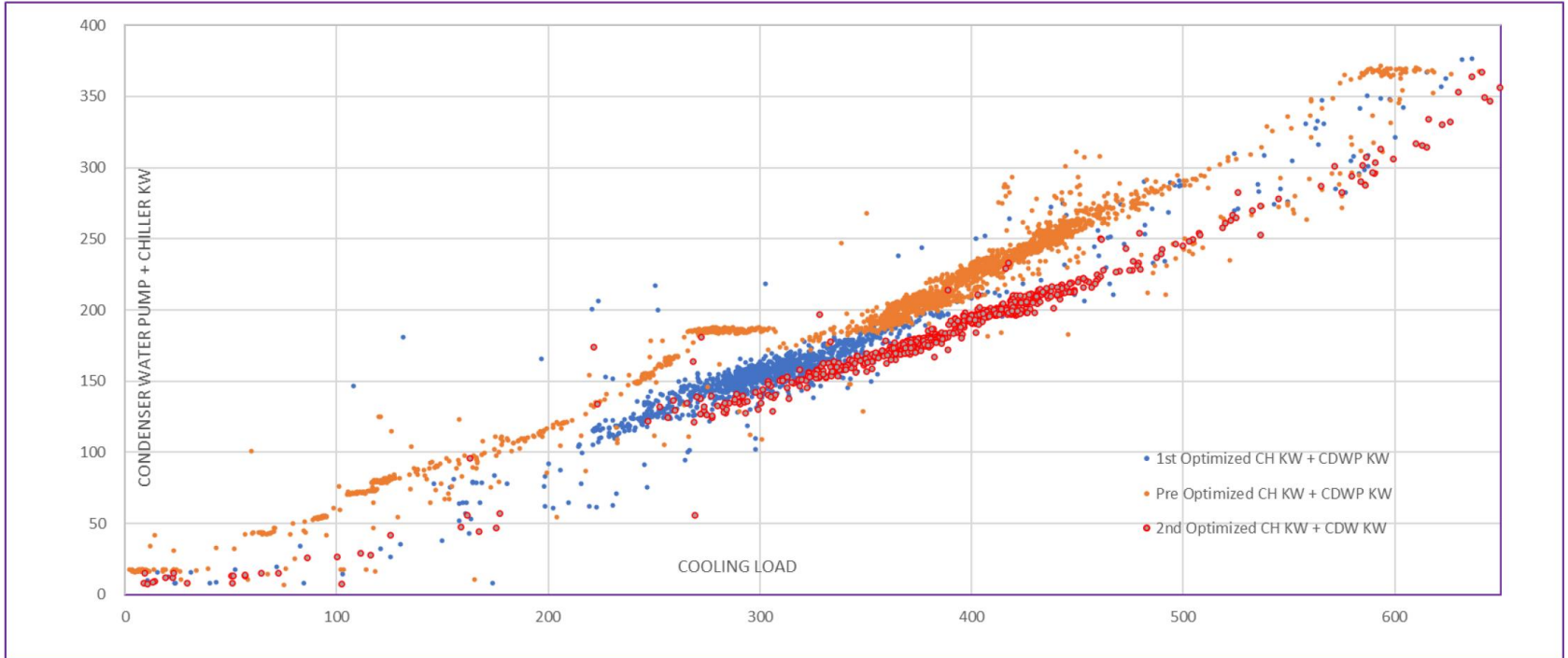
Case Study – University, Singapore

Condenser Water System Optimisation



Case Study – University, Singapore

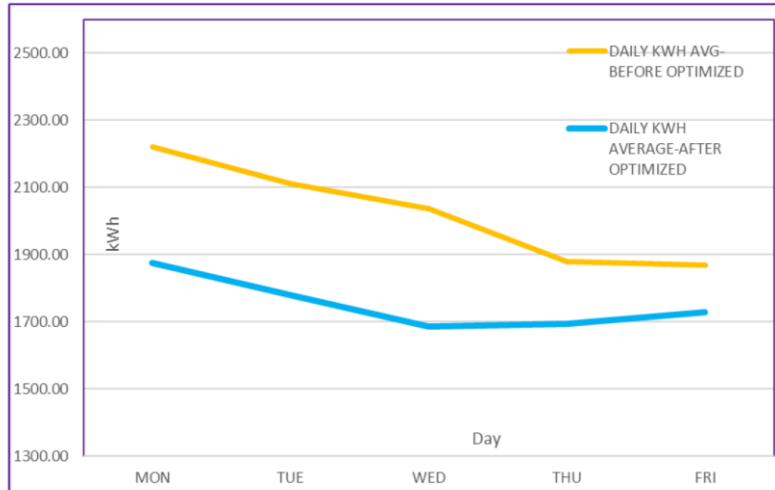
Condenser Water System Optimisation



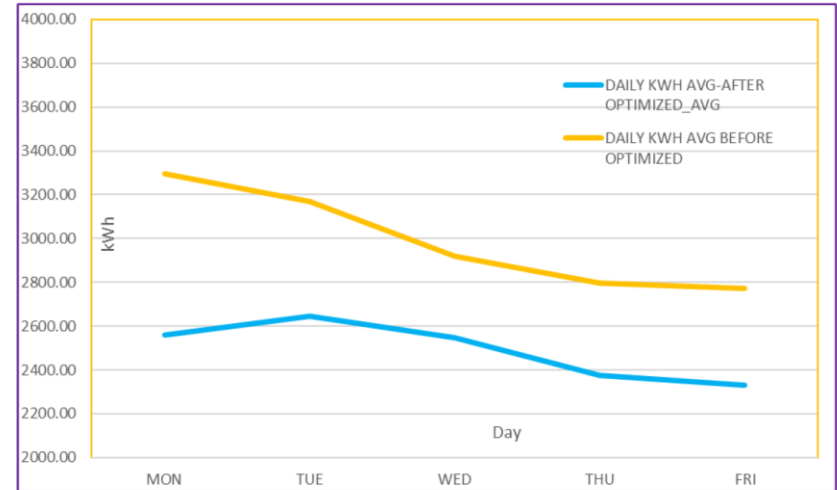
Case Study – University, Singapore

Power consumption comparison

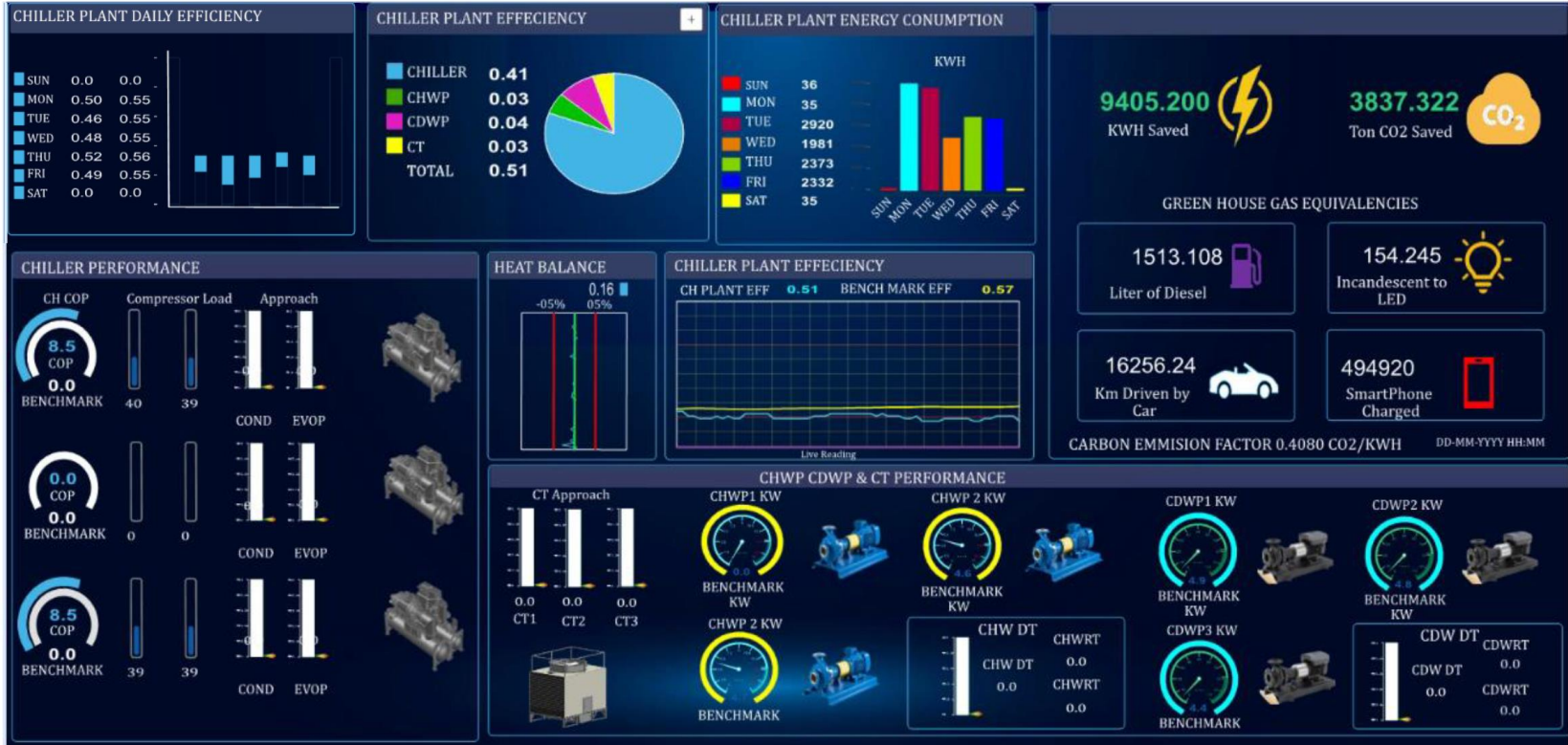
Before and after daily average kWh consumption of Block 04 chiller plant during term holiday



Before and after daily average kWh consumption of Block 04 chiller plant during normal operation



Case Study – University, Singapore





In summary

- Towards a sustainable, carbon-neutral world, efficient equipment must not be considered independently
- **Integration is key**, with simpler configuration and plug-&-play solutions
- **Data-driven operations:** Next-Gen smart energy savings sit between digital expertise and core control know-how
- **Paradigm shift:** From equipment to smart buildings to smart cities, combine technology for innovative solutions and dynamic adaptation to changing conditions



District Energy

LATAM Conference 2023

SEP 27 - 29 | CARTAGENA | CO

¡GRACIAS!

Rui Fonseca
rui.fonseca@daikin.com.sg



