



AUGUST 21-23, 2018 • CLEVELAND, OHIO

Metrics and Big Data



ALTRAMESE ROBERTS-TOMPKINS | *SENIOR ENGINEER*

Happy Target Setting

May the metrics be forever in your favor

TOYOTA

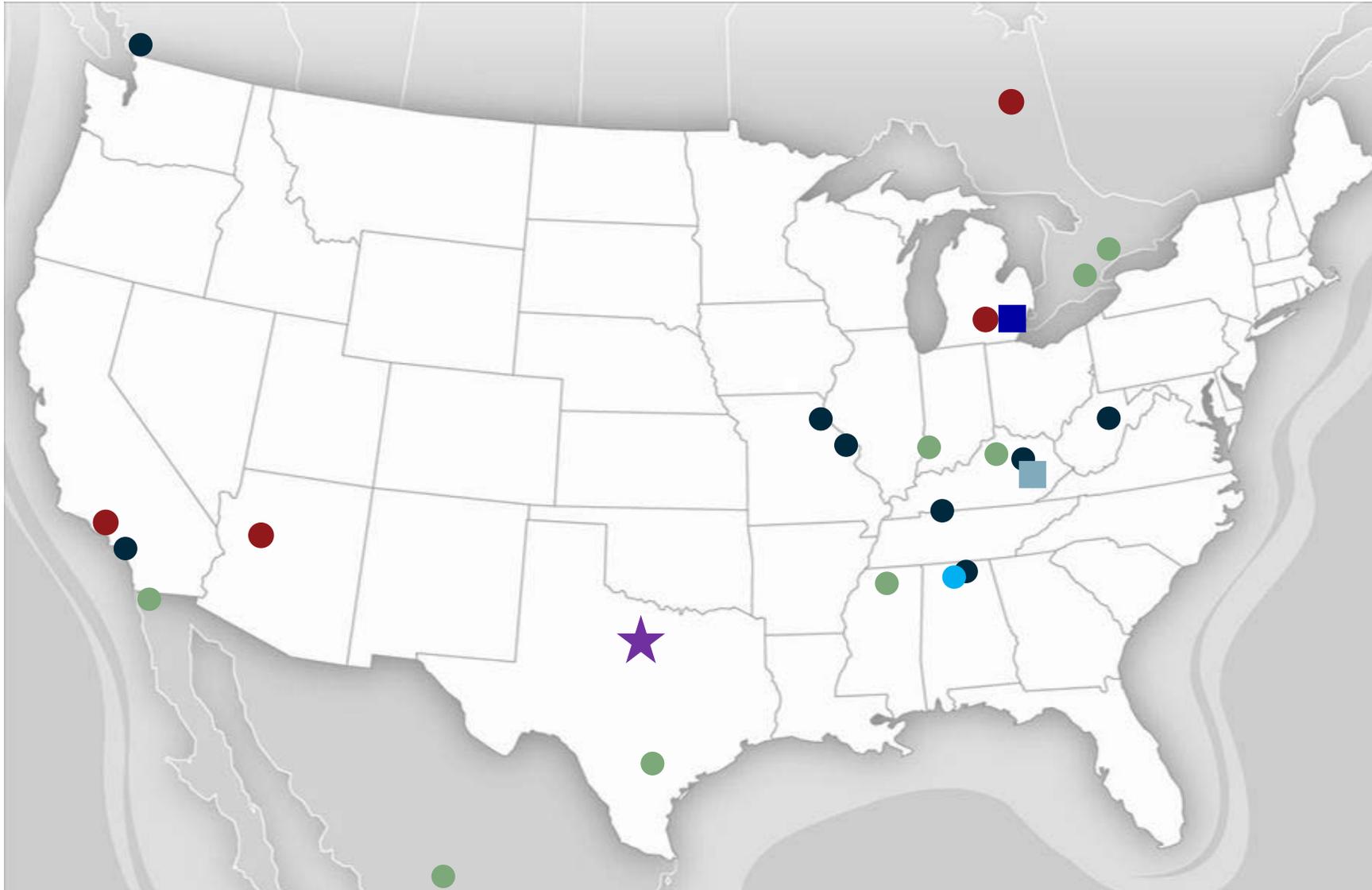


Background

Operations



TOYOTA NORTH AMERICA OPERATIONS



- Vehicle Assembly (8)
- Unit Plant (8)
- Research & Development (4)
- Manufacturing Purchasing
- Production Engineering
Georgetown, KY
- ★ Corporate Headquarters
Plano, TX
- Toyota & Mazda Joint Venture
Huntsville, AL

29 Million Sq.Ft.

Strategy

Plan for Action



Facilities Metrics

Metrics & Measures

- Unplanned Failures
- Repair Time (hours)
- In House PMs Planned
- In House PMs Completed
- Natural Gas Purchase Cost (\$)
- Electric Purchased Cost (\$)



KPIs

MTTR
(hours / # of failures)

PM Completion Rate
(%)

Total Energy Cost
(\$)



Quality

Production

Cost

Environmental Metrics

Metrics & Measures

- Electric Consumed (kWh)
- Natural Gas Consumed (MMBtu)
- Water Consumed (gal)
- Total Production (unit)
- Non Saleable Waste (kg)
- Volatile Organic Compounds (VOCs) (g/m²)

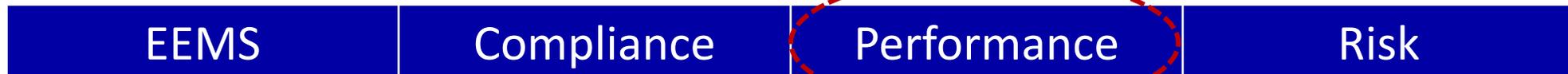
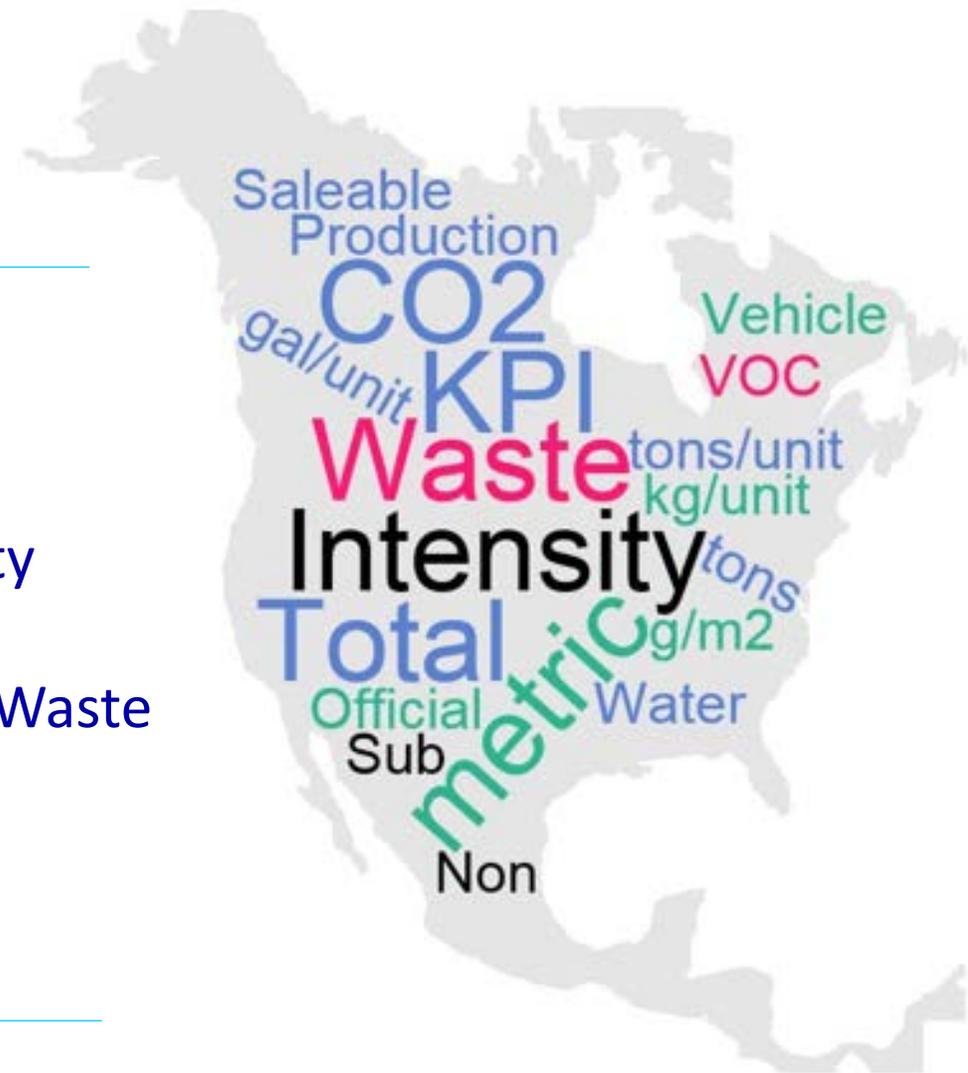
KPIs

CO₂ Absolute (MT)

Water Intensity (gal / unit)

Non Saleable Waste (kg / unit)

VOCs (g / m²)



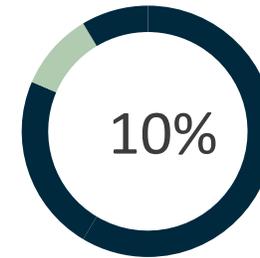
5 Year Strategy

The image shows a screenshot of a table titled 'Environmental Action Plan (EAP)'. The table has multiple columns and rows, with some rows highlighted in green. The text 'Environmental Action Plan (EAP)' is overlaid in red on the table.

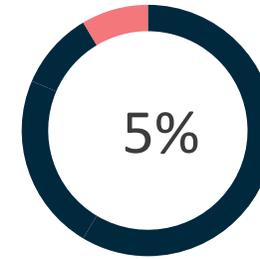
Environmental Action Plan (EAP)

Activities planned to achieve targets

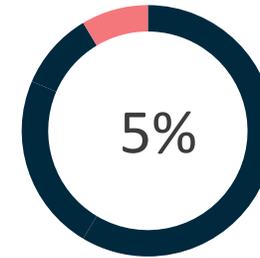
% Baseline reduction by 5th year



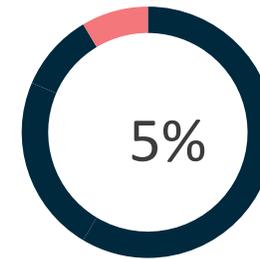
CO₂:
2% Per Year



Water:
1% Per Year



VOC:
1% Per Year



Waste:
1% Per Year

Annual Planning

Yearly Checkpoint



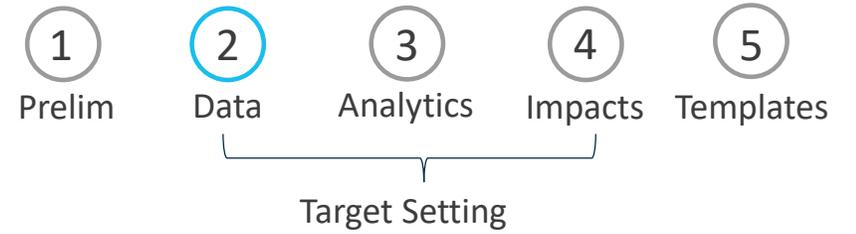
Preparation

- 1 Preliminary
- 2 Data Gathering
- 3 Linear Regression
- 4 Project Impacts
- 5 Target Templates



Target Setting Process

Data Gathering



- EPI Database
- BizEE Degree Day Website

Degree Days.net - Custom Degree Day Data

Degree Days.net calculates degree-day data for energy-saving professionals worldwide. The software is developed by [BizEE Software](#) based on temperature data from [Weather Underground](#).

Degree Days.net

Enter a weather station ID if you have one, or search for any city, state, ZIP code, or airport code.

Weather station ID

Data type Heating Cooling Regression(beta)

Temperature units Celsius Fahrenheit

Base temperature Include base temperatures nearby

Breakdown Daily Weekly Monthly Custom Average

Period covered

Regression Analysis



Actuals thru Dec. 2017 + FY18/FY19 Projected Production						Actual	Forecast				
FY	Month	ELE (kWh)	Vehicles	CDD	3YR CDD	FY	Month	NG (MMBtu)	Vehicles	HDD	3YR HDD
FY18	Jan-17	19,567,901	37,012	0	0	FY18	Jan-17	148,818	37,012	855	965
	Feb-17	17,996,885	34,874	5	2		Feb-17	121,952	34,874	575	820
	Mar-17	20,246,456	40,718	10	8		Mar-17	124,659	40,718	582	574
	Apr-17	16,928,296	26,484	78	54		Apr-17	53,442	26,484	207	256
	May-17	21,302,846	38,015	153	159		May-17	63,136	38,015	130	122
	Jun-17	23,389,839	38,889	305	342		Jun-17	51,911	38,889	29	17
	Jul-17	22,190,833	24,026	365	375		Jul-17	40,838	24,026	6	4
	Aug-17	24,789,006	38,961	232	296		Aug-17	50,124	38,961	33	22
	Sep-17	21,970,158	34,982	197	211		Sep-17	56,984	34,982	82	74
	Oct-17	20,531,234	37,347	83	75		Oct-17	83,790	37,347	276	253
	Nov-17	17,960,582	32,880	6	11		Nov-17	112,802	32,880	591	541
	Dec-17	17,476,206	28,253	0	1		Dec-17	149,351	28,253	990	882
Jan-18	0	37,402	0	0	Jan-18	0	37,402	1,166	1,024		
Feb-18	0	34,285	0	0	Feb-18	0	34,285	0	0		
Mar-18	0	37,747	0	0	Mar-18	0	37,747	0	0		
FY19	Apr-18		37,434			FY19	Apr-18		37,434		
	May-18		39,061				May-18		39,061		
	Jun-18		35,652				Jun-18		35,652		
	Jul-18		26,763				Jul-18		26,763		
	Aug-18		40,785				Aug-18		40,785		
	Sep-18		32,300				Sep-18		32,300		
	Oct-18		37,492				Oct-18		37,492		
	Nov-18		32,408				Nov-18		32,408		
	Dec-18		24,485				Dec-18		24,485		
	Jan-19		38,054				Jan-19		38,054		
	Feb-19		35,680				Feb-19		35,680		
	Mar-19		35,040				Mar-19		35,040		

Variables

- Electricity (kWh)
- Natural Gas (MMBtu)
- Water (gal)
- CDD (cooling degree days)
- HDD (heating degree days)
- Production (units)

Regression Analysis



NG (MMBtu) Regression

SUMMARY OUTPUT

Regression Statistics	
Multiple R	0.99336664
R Square	0.98677727
Adjusted R Square	0.98383889
Standard Error	5189.11599
Observations	12

ANOVA					
	df	SS	MS	F	Significance F
Regression	2	18085371444	9042685722	335.823188	1.00000
Residual	9	242342323.2	26926924.8		
Total	11	18327713767			

	Coefficients	Standard Error	t Stat	P-value	Lower 95%	Upper 95%	Lower 95.0%	Upper 95.0%	
Baseload Intercept	3872.89012	10278.20273	0.376806161	0.71504822	-19378.0198	27123.8	-19378.0198	27123.8	0.28495
Production X Variable 1	1.21203434	0.290109445	4.177852073	0.00238367	0.55576119	1.8683075	0.55576119	1.8683075	0.99762
HDD X Variable 2	117.461625	4.567872474	25.71473386	9.7997E-10	107.12838	127.79487	107.12838	127.79487	1.00000

- Perform linear regression
- Review summary output

ELE (kWh) Regression

SUMMARY OUTPUT

Regression Statistics	
Multiple R	0.95888739
R Square	0.91946503
Adjusted R Square	0.90156837
Standard Error	778525.191
Observations	12

ANOVA					
	df	SS	MS	F	Significance F
Regression	2	6.22786E+13	3.11393E+13	51.3763472	0.99999
Residual	9	5.45491E+12	6.06101E+11		
Total	11	6.77335E+13			

	Coefficients	Standard Error	t Stat	P-value	Lower 95%	Upper 95%	Lower 95.0%	Upper 95.0%	
Baseload Intercept	9763704.33	1566371.835	6.23324748	0.00015265	6220325.06	13307083.6	6220325.06	13307083.6	0.99985
Production X Variable 1	250.954427	43.85044406	5.722962049	0.00028588	151.757831	350.151023	151.757831	350.151023	0.99971
cDD X Variable 2	16515.723	1829.825343	9.025846668	8.34E-06	12376.3705	20655.0755	12376.3705	20655.0755	0.99999

Regression Analysis



Use variables from regression to model ELE, NG and Water

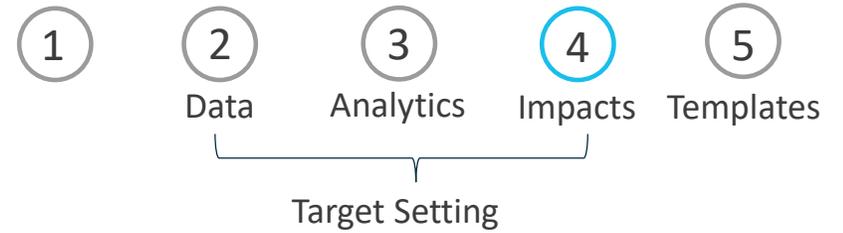
Regression Models		ELE and NG Variables		FY18	
		ELE Model	NG Model	ELE (lbs/kWh)	NG (lbs/MMBtu)
	Base	9,763,704 kWh/month	3,873 MMBtu/month	117	117
	Production	251 kWh/vehicle	1 MMBtu/vehicle		
	CDD	16,516 kWh/CDD	117 MMBtu/HDD		
Variables Converted (MT)					
	FY18 Base	7,305 MT/month			
	FY19 Base	5,711 MT/month			

FY	Month	Month	CDD	HDD	3YR CDD	3YR HDD
FY17	Apr-16	19,294,986	46	342	46	274
	May-16	19,967,152	104	163	182	114
	Jun-16	23,856,342	375	8	360	10
	Jul-16	22,494,725	352	2	348	14
	Aug-16	27,748,058	362	5	337	14
	Sep-16	23,090,819	233	53	209	82
	Oct-16	20,714,618	97	192	66	248
	Nov-16	17,960,582	112,802	17,222	6	
	Dec-16	17,476,206	149,351	18,858	0	

FY18		ELE (kWh)	NG (MMBtu)	Total MT	MT/veh	Prod	CDD	HDD
Month								
Apr-17	16,928,296	53,442	13,426	0.51	26,484	78		
May-17	21,302,846	63,136	16,677	0.44	38,015	153		
Jun-17	23,389,839	51,911	17,387	0.45	38,889	305		
Jul-17	22,190,833	40,838	16,050	0.67	24,026	365		
Aug-17	24,789,006	50,124	18,168	0.47	38,961	232		
Sep-17	21,970,158	56,984	16,768	0.48	34,982	197		
Oct-17	20,531,234	83,790	17,291	0.46	37,347	83		
Nov-17	17,960,582	112,802	17,222	0.52	32,880	6		
Dec-17	17,476,206	149,351	18,858	0.67	28,253	0		
Jan-18	0	0	25,054	0.67	37,402	0		
Feb-18	0	0	23,027	0.67	34,285	2		
Mar-18	0	0	22,376	0.59	37,747	8		
			222,305	0.54	409,271	1,429		
			87,657	7,305				

FY19		ELE (kWh)	NG (MMBtu)	Total MT	MT/veh	Prod	CDD	HDD
Month								
Apr-18	$=\$D\$4 + \$D\$5 * O21 + \$D\$6 * P21$	15,515	40,998	15,515	0.41	37,434	54	
May-18		22,186,124	65,527	15,988	0.41	39,061	159	
Jun-18		24,384,550	40,998	16,345	0.46	35,652	242	

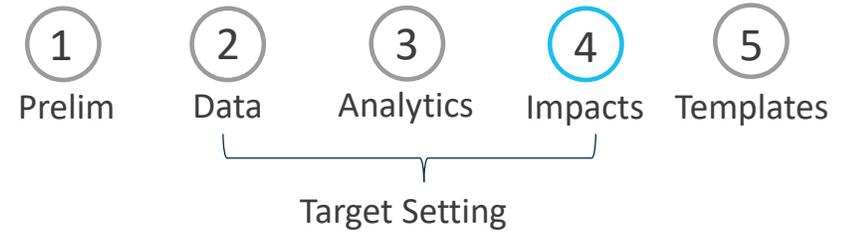
Project Impacts



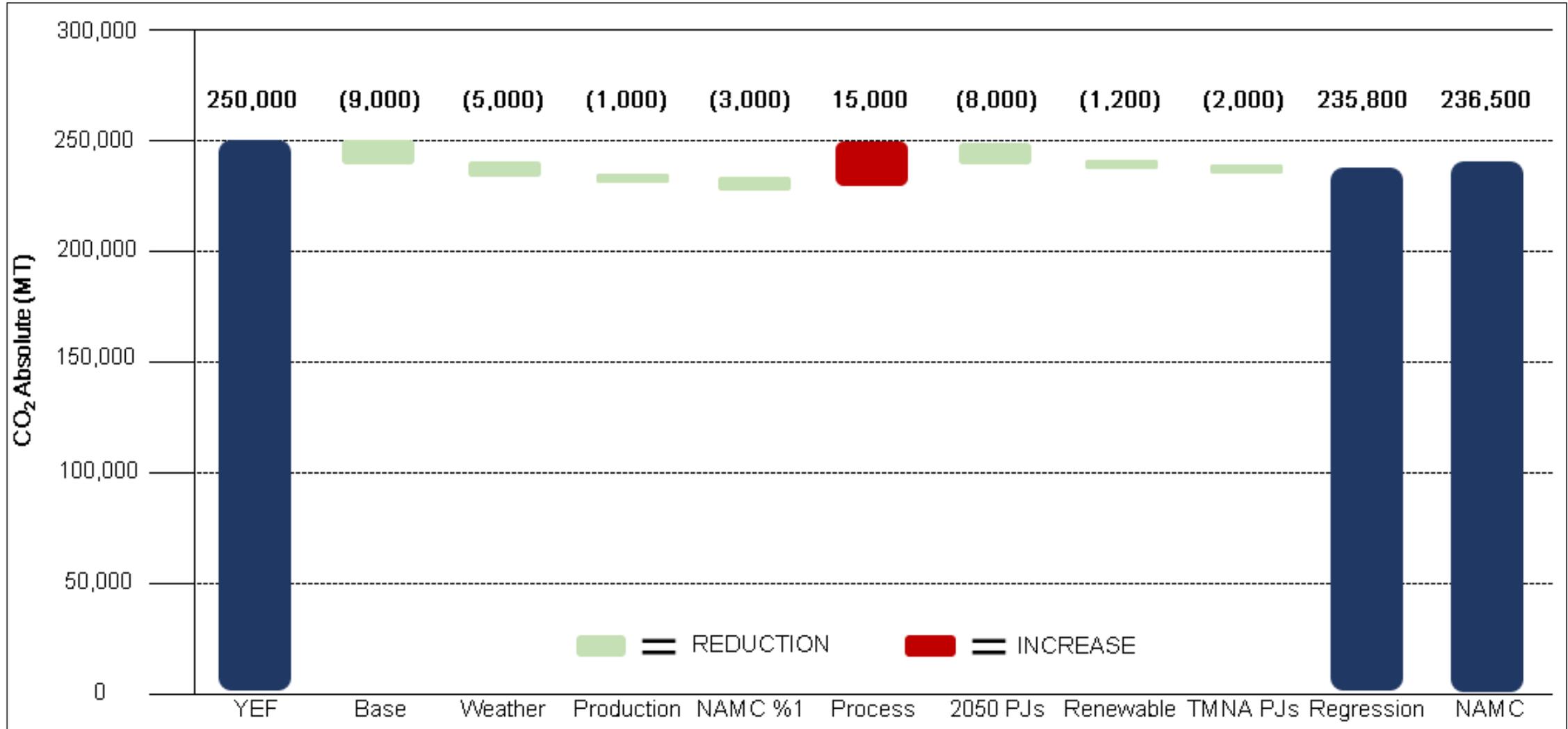
Target table and project impact

Total CO2 Absolute Table												
Month	Previous FY	Forecasted FY										
		Base Usage	Weather Impact	Production Impact	NAMC Kaizens	Process Changes	PE Technology	2050 Projects	Renewable	TMNA Projects	Total Monthly	Total Cumulative
April	13,426	5,711	2,099	7,705	(134)	3,540	0	0	0	0	18,921	18,921
May	16,677	5,711	2,237	8,040	(167)	3,548	0	0	0	0	19,369	38,290
June	17,387	5,711	3,295	7,338	(174)	3,764	0	0	0	0	19,935	58,225
July	16,050	5,711	3,512	5,509	(160)	4,127	0	(77)	0	0	18,621	76,846
August	18,168	5,711	2,895	8,395	(182)	4,815	0	(77)	0	0	21,557	98,404
September	16,768	5,711	2,421	6,648	(168)	4,081	0	(77)	0	0	18,617	117,021
October	17,291	5,711	2,278	7,717	(173)	3,520	0	(77)	0	0	18,976	135,996
November	17,222	5,711	3,471	6,671	(172)	4,172	0	(77)	0	0	19,776	155,772
December	18,858	5,711	5,503	5,040	(189)	4,029	0	(77)	0	0	20,017	175,789
January	25,054	5,711	6,014	7,833	(251)	3,244	0	(591)	0	0	21,960	197,749
February	23,027	5,711	5,130	7,344	(230)	3,244	0	(591)	0	0	20,608	218,358
March	22,376	5,711	3,652	7,212	(224)	3,242	0	(591)	0	0	19,002	237,360
FY Total	222,305	68,537	42,507	85,451	(2,223)	45,327	0	(2,239)		0	237,360	237,360

Project Impacts



Waterfall Chart



Target Templates



What do we propose?

- Review files
- One on one calls/meetings
- Sent out templates

PERFORMANCE Target Template

SIGNATURE BLOCK

Fiscal Year FY19

NAMC NAME : _____

Production (veh)

FY / Month	A	M	J	J	A	S	O	N	D	J	F	M	5YR AP	CURRENT YEF	FY19 Target	Comment/Reason
TMNA Monthly Forecast (Vehicle)																
TMNA Monthly Forecast Cumulative (Vehicle)																
NAMC Monthly Forecast (Vehicle)																
NAMC Monthly Forecast Cumulative (Vehicle)																

Absolute CO2

FY / Month	A	M	J	J	A	S	O	N	D	J	F	M	5YR AP	CURRENT YEF	FY19 Target	Comment/Reason
TMNA Monthly Target																
TMNA Cumulative Target																
NAMC Monthly Target (if different)																
NAMC Cumulative Target (if different)																

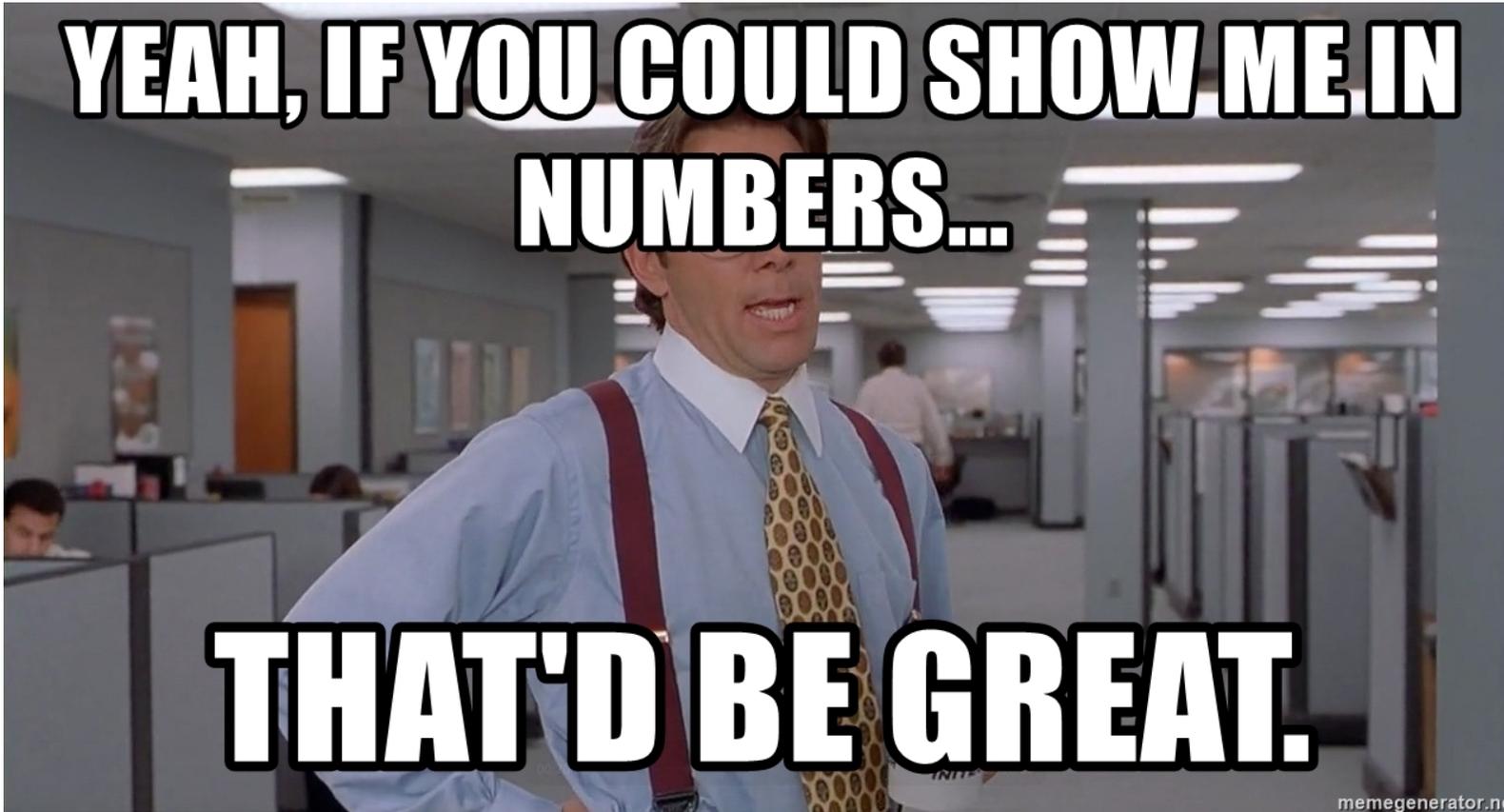
Energy (MMBtu/veh)

FY / Month	A	M	J	J	A	S	O	N	D	J	F	M	5YR AP	CURRENT YEF	FY19 Target	Comment/Reason
TMNA Monthly Target (MMBtu / Vehicle)																
TMNA Cumulative Target (MMBtu / Vehicle)																
NAMC Monthly Target (if different)																
NAMC Cumulative Target (if different)																

Water (Gal/veh)

FY / Month	A	M	J	J	A	S	O	N	D	J	F	M	5YR AP	CURRENT YEF	FY19 Target	Comment/Reason
TMNA Monthly Target (Gal / Vehicle)																

Negotiation & Consensus



Let's Talk

- Method used?
- What is included?
- Align with 5 Year Plan?
- Unplanned conditions?

Tracking Progress

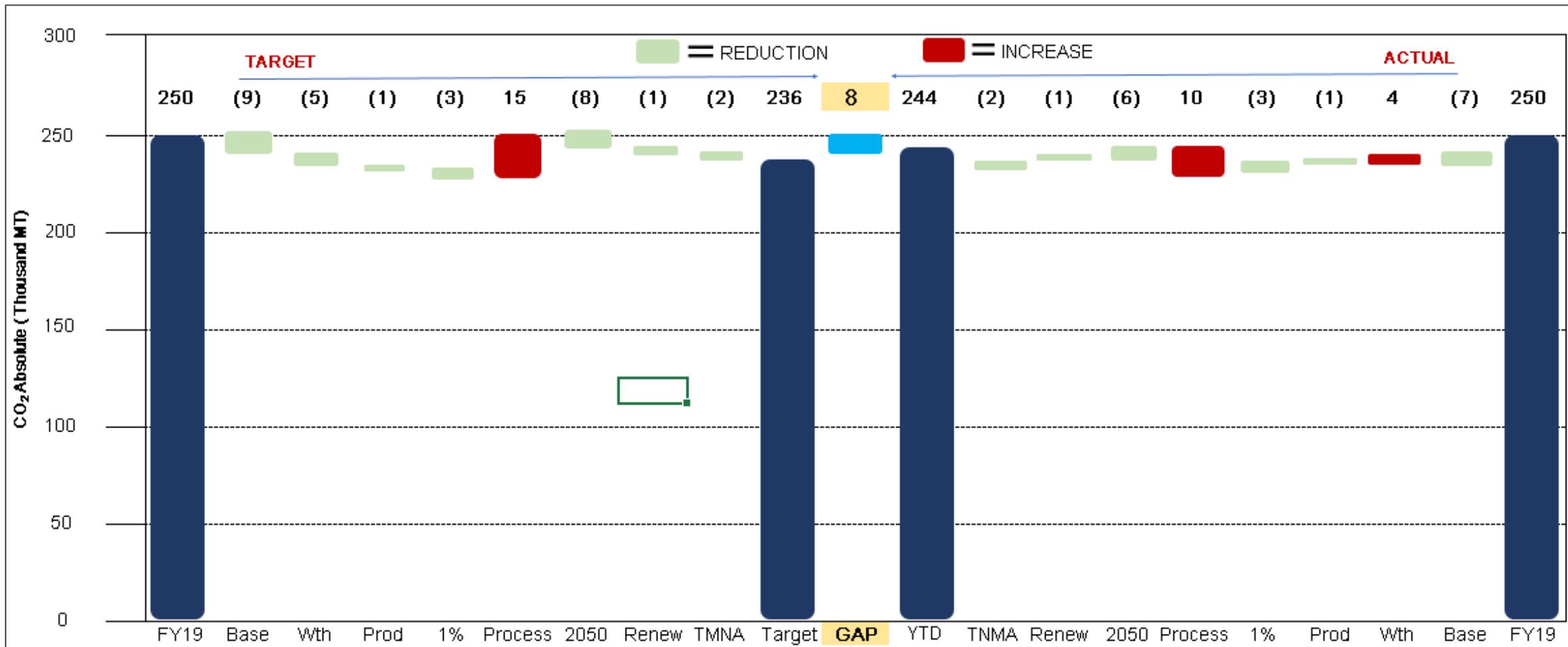
Action



Tracking Targets

○ or ✗ condition

How are we doing?



If target setting is productive...

“May the metrics be forever in your favor”



Best Practices...

- Ideal Metrics
- Have a Plan
- Have Consensus
- Track Progress

THANK YOU





Energy & Water Analytics at P&G

August 2018

Journey

The P&G logo is a white, italicized serif font centered within a dark blue circular graphic that has a 3D effect. This graphic is positioned at the intersection of two large, overlapping blue shapes that form a stylized 'C' or 'J' shape on the left side of the slide. The background of the slide is a vibrant blue, and a circular cutout on the left reveals a scenic landscape of a river winding through a dense green forest under a blue sky with scattered white clouds.

- Big Data & Analytics
- What we learned
- Recommendations

P&G AT A GLANCE

Countries of Operations: **~70**

Countries Where Our Brands Are Sold: **180+**

Consumers Served by Our Brands:
Nearly **5 Billion**



SOME OF THE WORLD'S MOST TRUSTED BRANDS



2020 GOAL PROGRESS

Within our operations we strive to grow responsibly, constantly improving our efficiency while reducing our footprint. Production has increased since 2010; however, we have successfully decoupled that growth from our environmental footprint.



GHG

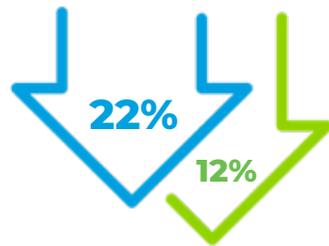


25% reduction per unit of production

16% absolute reduction



ENERGY



22% reduction per unit of production (exceeding goal early)

12% absolute reduction



WASTE



80% reduction in solid waste with 72% of our sites achieving zero manufacturing waste to landfill

77% absolute reduction

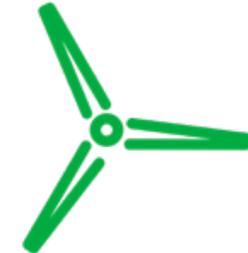


WATER



27% reduction per unit of production (exceeding goal early)

18% absolute reduction



RENEWABLE ENERGY



Currently **10%** with large scale wind farm and biomass projects coming online in the back half of the fiscal year that will double our usage to ~20%



Data

“From the dawn of civilization until 2003, humankind generated five exabytes of data. Now we produce five exabytes every two days...and the pace is accelerating.”

Eric Schmidt, Executive Chairman, Google



Value of data

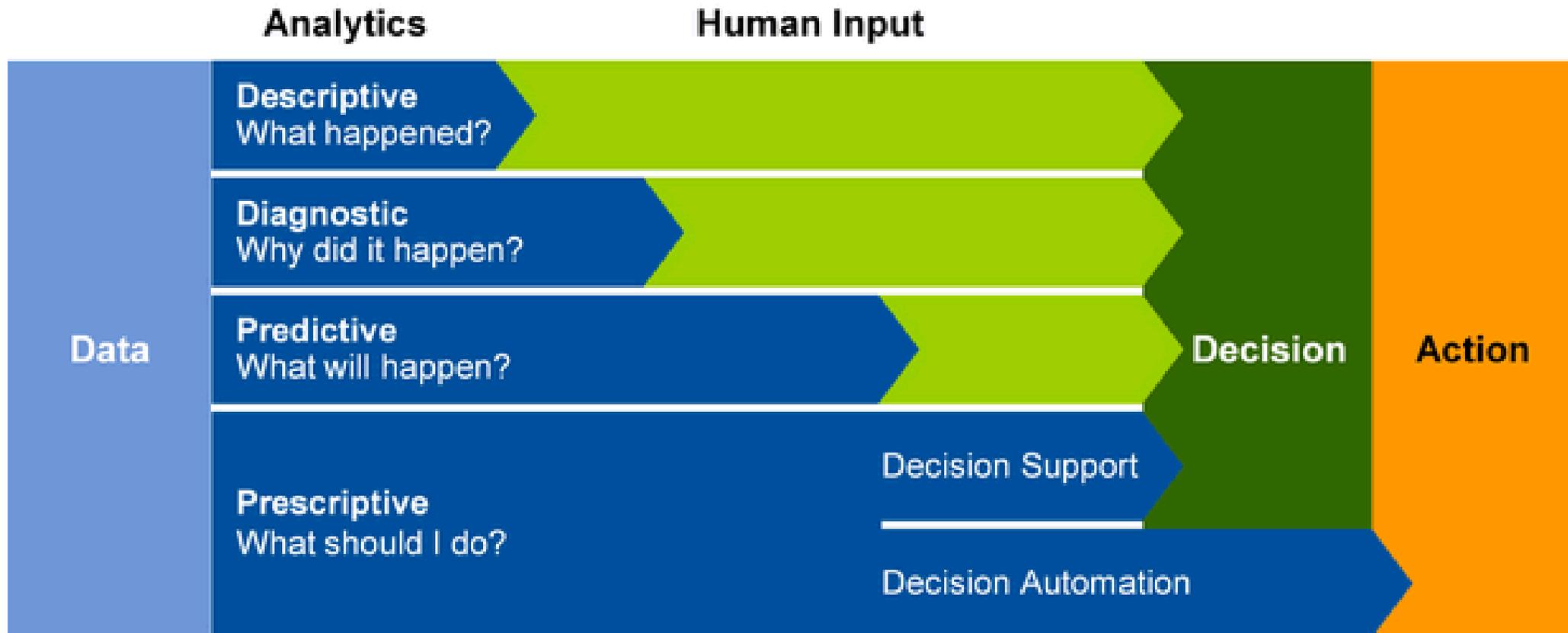
Ability to hear the
signal from the noise
is crucial...



P&G



It is all about ACTIONABLE INSIGHTS



Source: Gartner (October 2014)





THE CHALLENGES

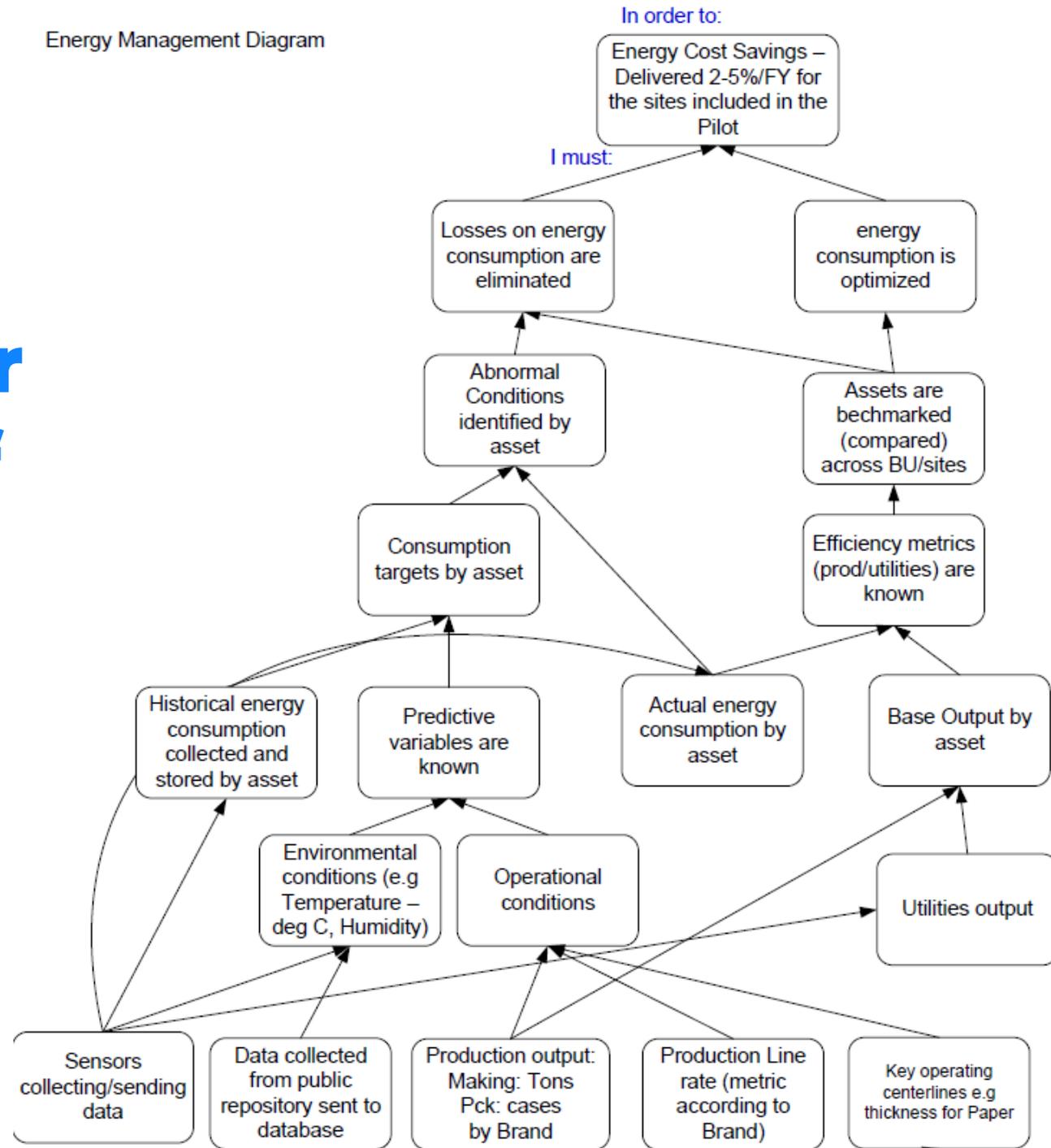
ARCHITECTURE: Centralizing data requires a clear flow of data from operating floor.

STORAGE INFRASTRUCTURE: Enterprise infrastructure is evolving rapidly

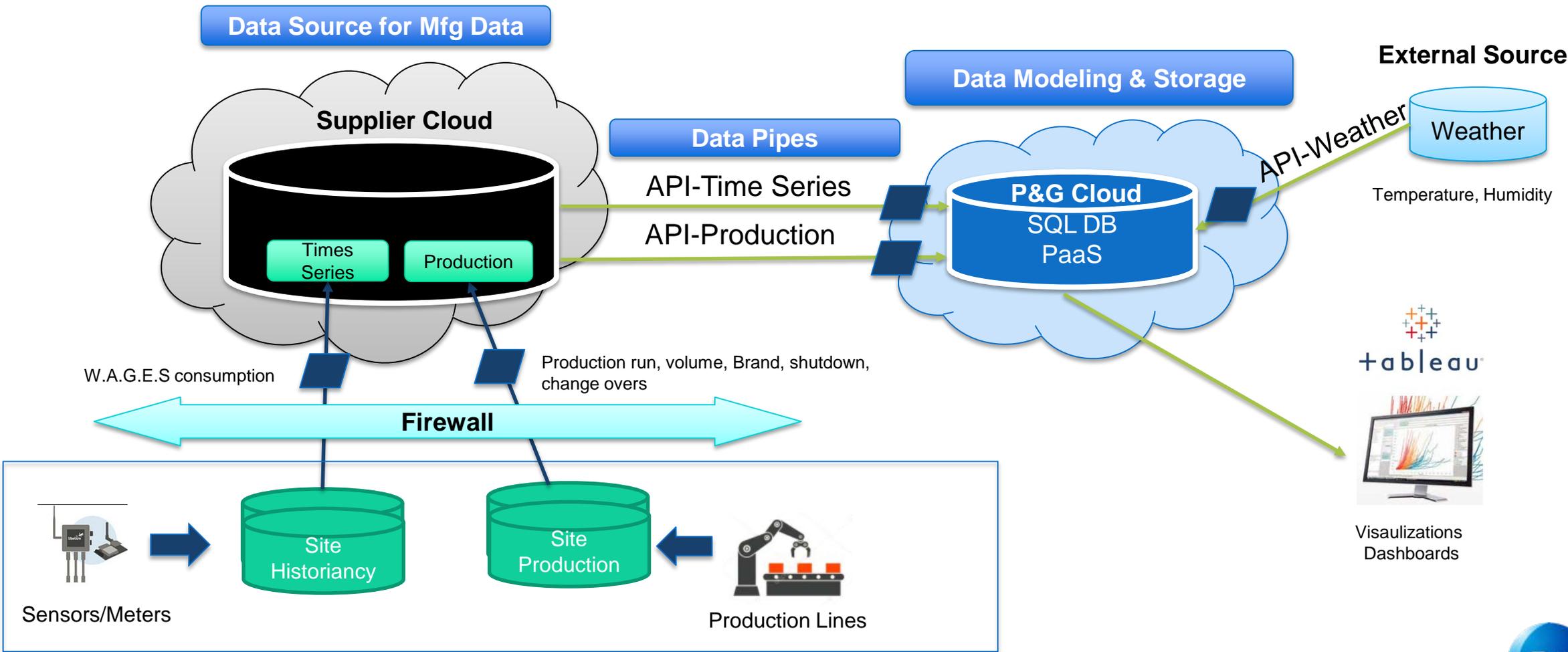
DATA HARMONZIATION: Data was there for 20-30 years. First time pulling it all together.

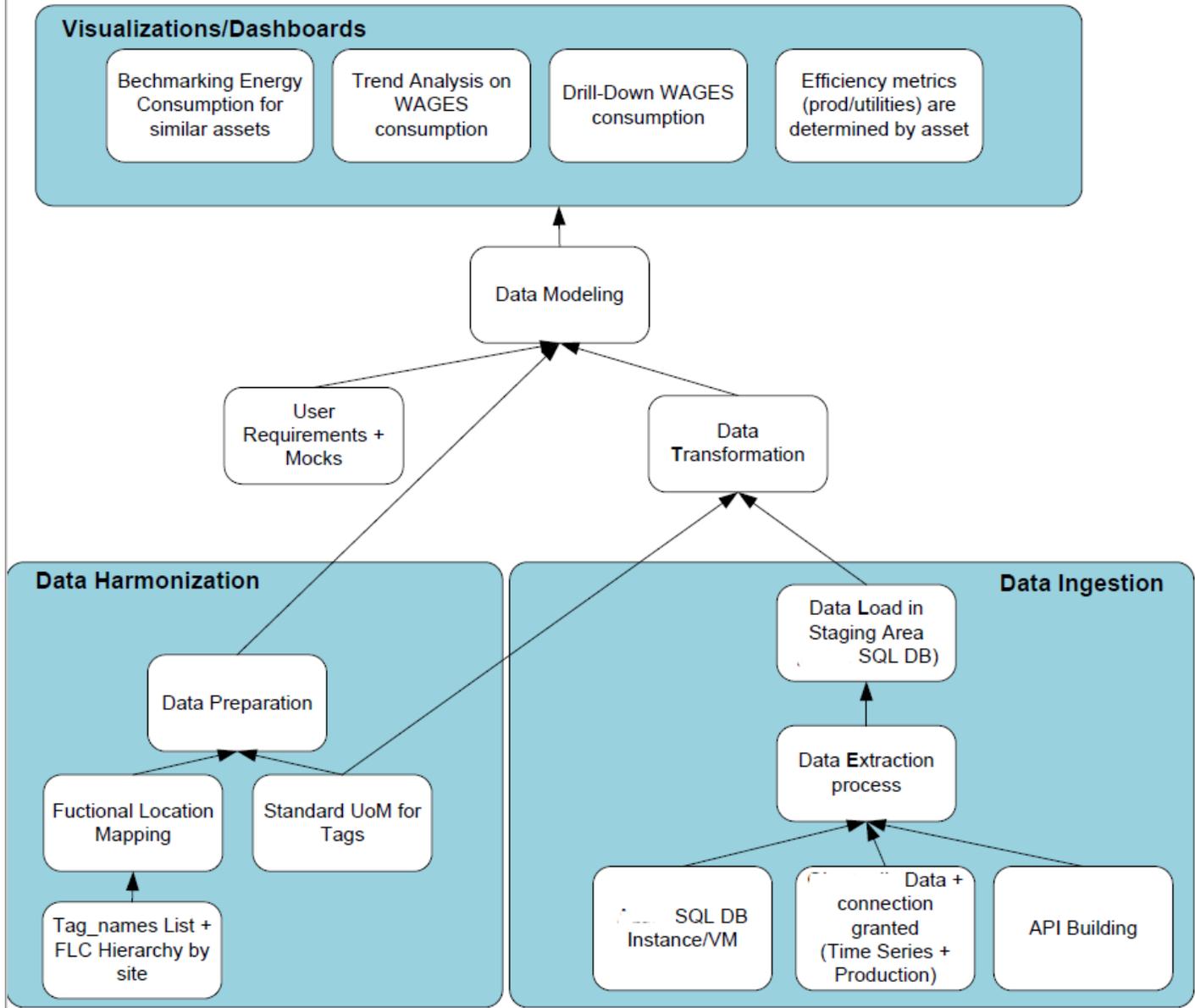
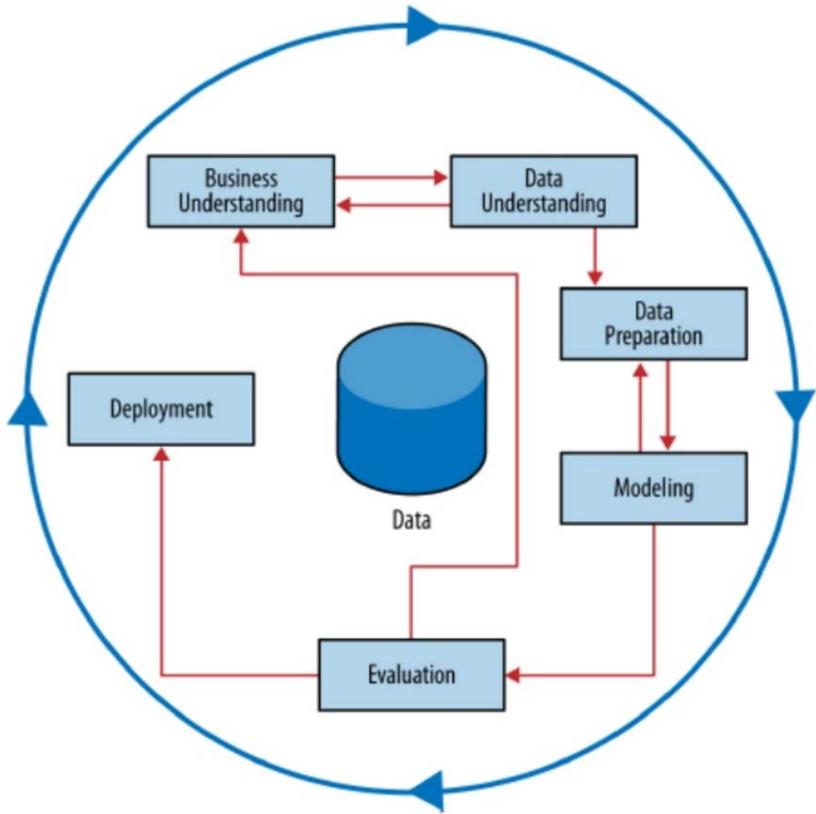
SUPPLIERS: There is big money in energy data. Everyone is knocking on our door.

Energy & Water Analytics Proof of Concept

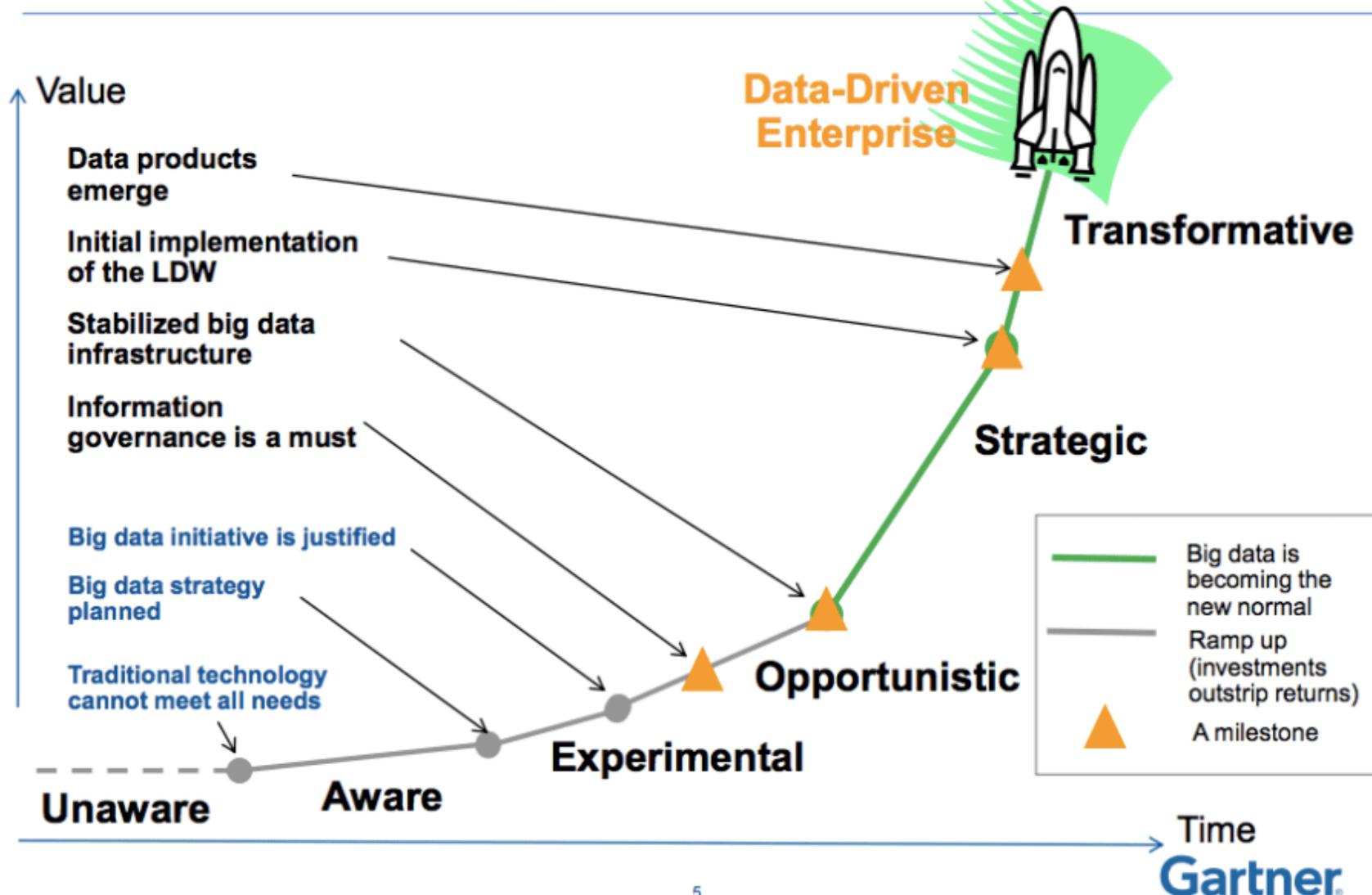


Energy & Water Analytics Proof of Concept

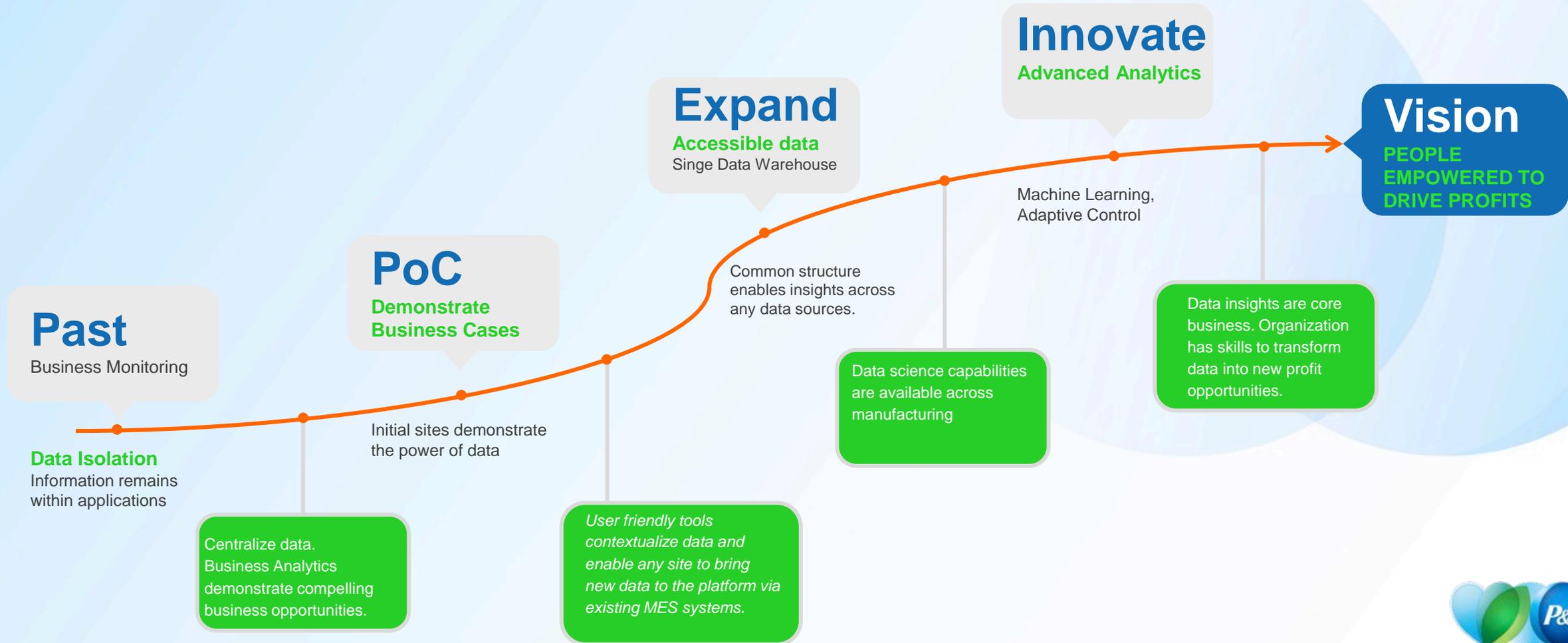




The Road Map: Typical Stages and Milestones of Big Data Adoption



DATA IS ACCESSIBLE | COMMON STRUCTURE | UNLEASH ORGANIZATION TO CREATE INSIGHTS



Recommendations



- Engage and Staff IT Leadership on project
- Start with clear user & business requirements
- Seek help from experts in Data Science and Infrastructure



P&G

THANK YOU



A photograph of an industrial facility with two men in the foreground. One man in a light blue shirt is holding a tablet, and the other in a white shirt is pointing at it. They are standing next to a large piece of industrial machinery. The word "DIGITAL" is overlaid in large green letters across the center of the image.

DIGITAL

Metrics and Big Data

Large Facility Energy Analysis

Presented by: Anand Varahala

Schneider Energy Action – North America

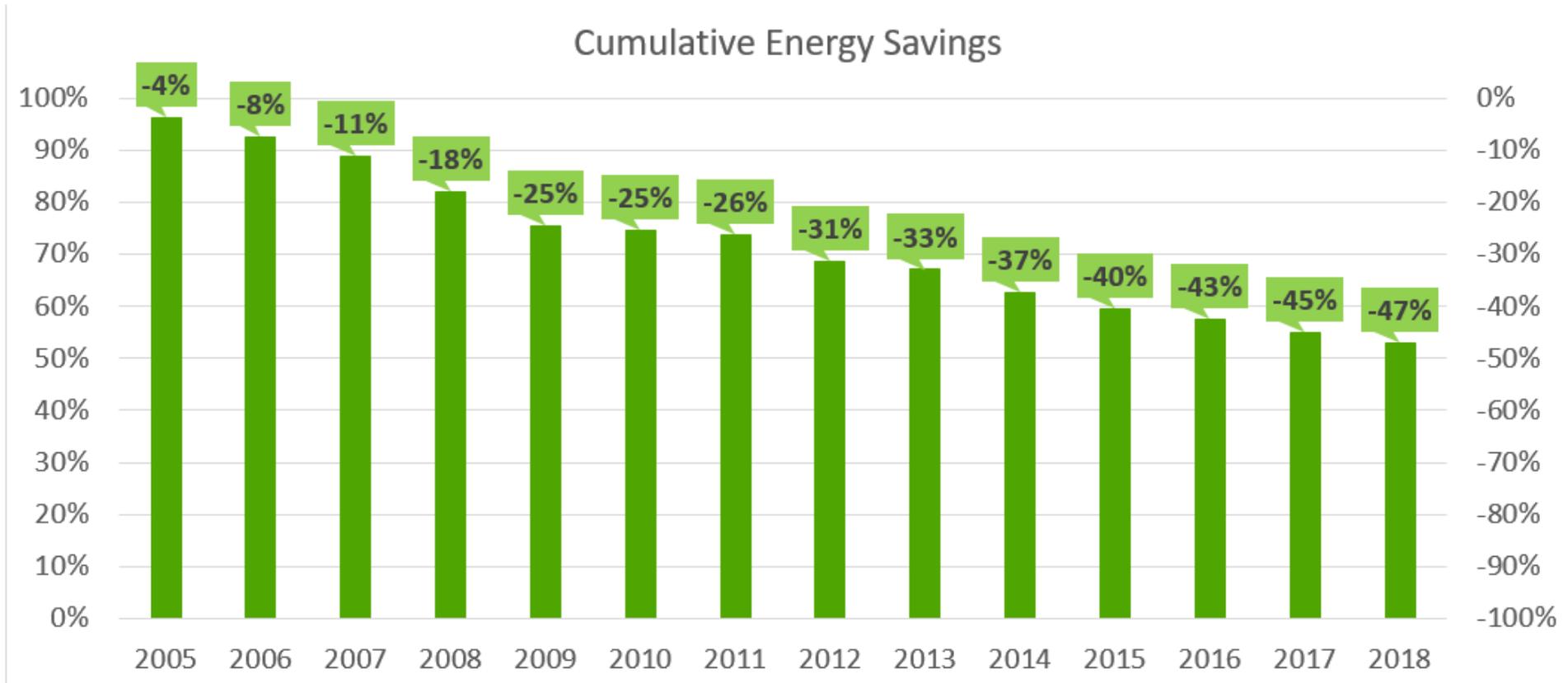
A little history...

- Inception in 2005 with 26 facilities
- Strategic partnership between Global Supply Chain and Energy & Sustainability Services
- **65** managed facilities across Canada, Mexico, and USA
- 5 GSC Clusters, 4 Business Operations represented



Program Performance

Cumulative energy savings (electricity & natural gas) normalized for weather and production



How do we measure performance?

KPI – Energy Model

BU / Cluster	Jan-18	Feb-18	Mar-18	Apr-18	May-18	Jun-18	'18 vs '17 Baseline		Elec	Gas
GSC NA - CTO/ETO Total	-3%	-4%	-2%	-2%	-4%	-15%	-5%		-8%	-2%
GSC NA - Electronics Total	-11%	-19%	-13%	-16%	-20%	-22%	-17%		-15%	-27%
GSC NA - ETO/CTO Total	-14%	-1%	-2%	-4%	-7%	-6%	-6%		0%	-10%
GSC NA - Logistics Total	6%	17%	13%	6%	-5%	0%	7%		-4%	24%
GSC NA - MTS/MTO Total	1%	-1%	6%	4%	3%	0%	2%		0%	6%
Industry Total	0%	-15%	-3%	-9%	-6%	-5%	-6%		-3%	-18%
Buildings & IT Total	11%	9%	6%	-5%	-6%	-14%	1%		-6%	16%
Operations NA Total	-16%	-5%	-5%	-9%	-7%	-4%	-8%		-6%	-15%
Grand Total	-5%	-2%	0%	-2%	-4%	-7%	-3%		-4%	-3%

Benefits

- Linear regression analysis
- Unique model for each site
- Utility invoice data (> 200 invoices per month)
- Weather data (cooling and heating degree days)
- Production and activity data
- Whole facility energy performance

Challenges

- Calendarization
- Time intensive invoice data processing
- Delay in reporting (What happened 6 weeks ago?)
- Timely identification of issues

The Incident, the Problem, and the Opportunity

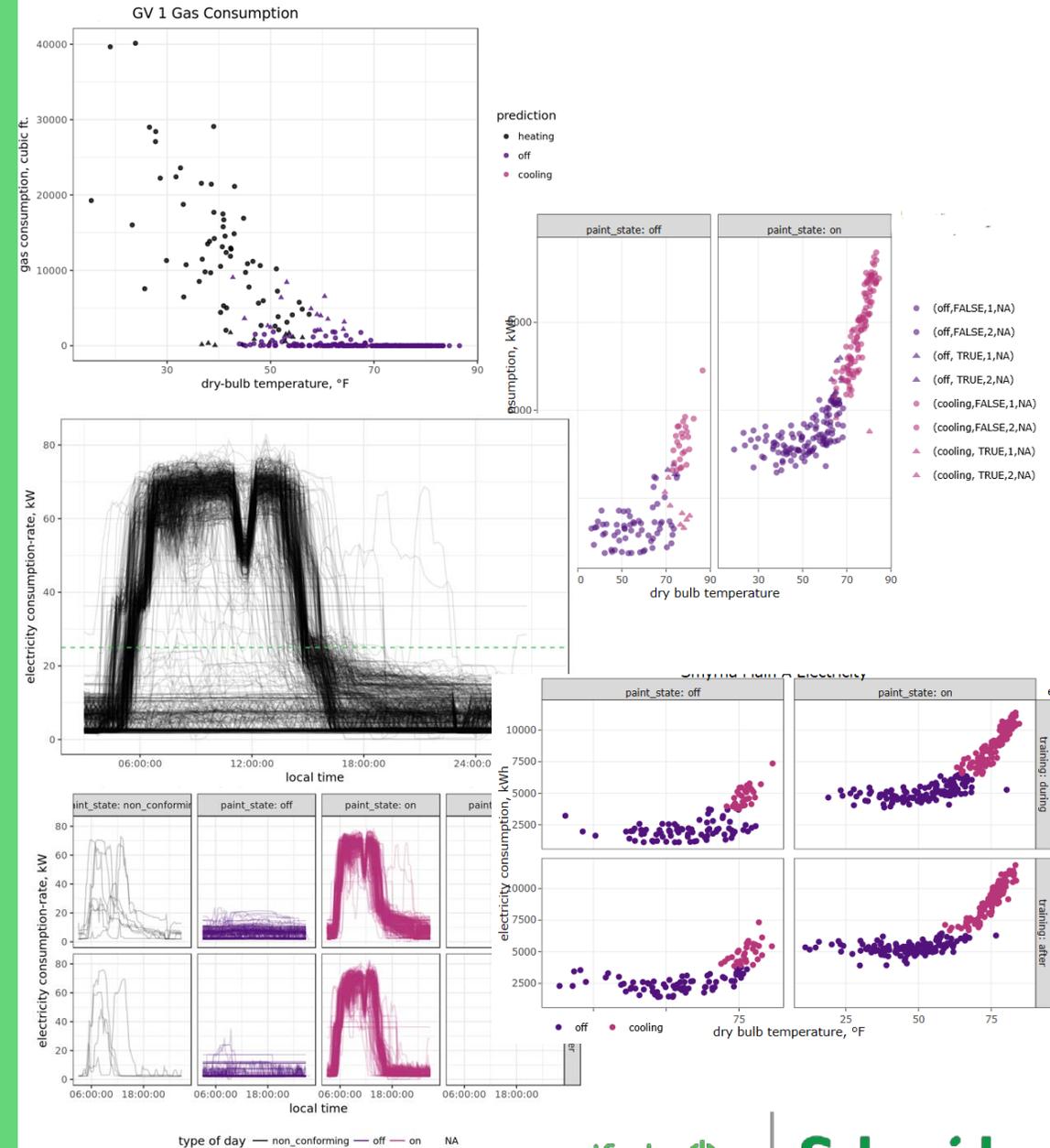
- Q1 2016
- Significant increase in gas consumption
- > 55k USD impact
- 2 months to identify deviation
- 3 months to confirm trend
- Reactive response

- Monthly frequency of the Energy Model reports
- Dependence on utility invoices
- Time/labor intensive process
- 4 to 6 week delay in determining deviations

- Existing network of sub-meters
- Central monitoring software
- Access to weather data through web-services (API)
- In-house expertise
- Strong collaboration with Analytics and Artificial Intelligence team

The Concept

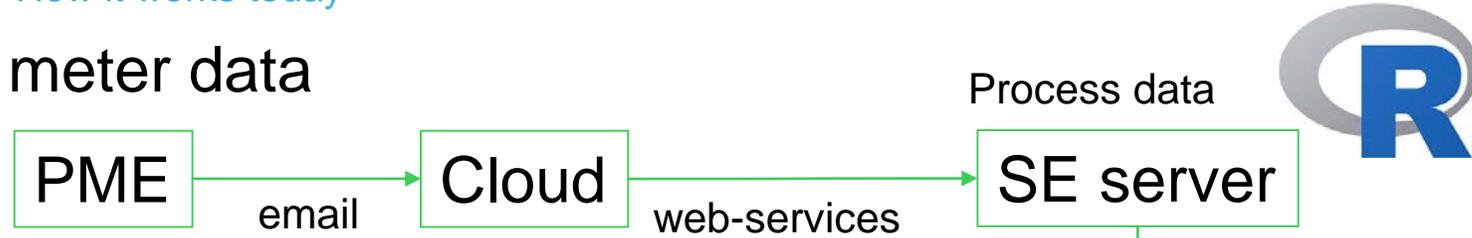
- A daily 'energy model'
- Take advantage of interval data from existing meters
- Develop production/activity related 'features' by analyzing the 15-minute interval electricity and gas consumption data
- Create visualization that is easy to read, interpret, and, understand



Architecture

How it works today

meter data



Process data

Azure

web-services

Azure Data Factory

web-services

weather data



SE server

Evaluate models
Build web pages

GitHub Enterprise

Serve static web-pages

Daily Energy Model

Data Acquisition



Analytics



Visualization

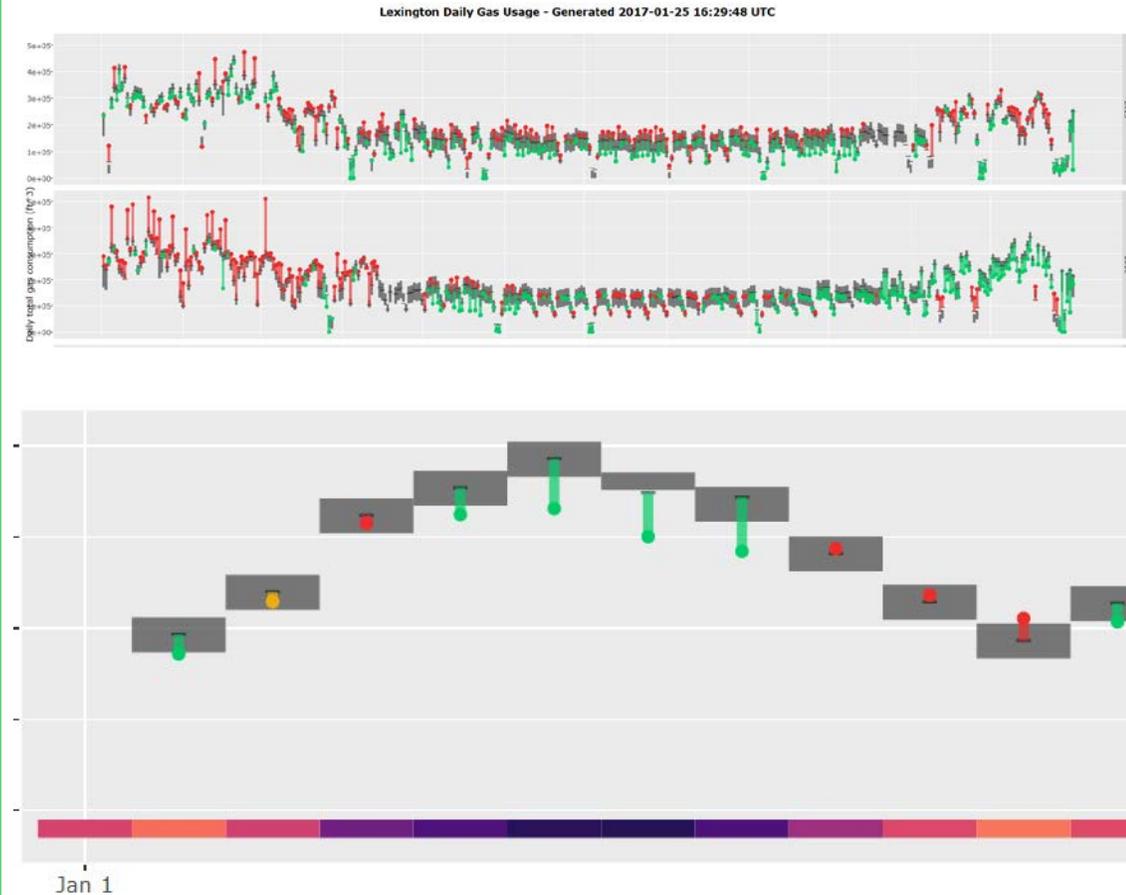


Detect and correct problems before they become expensive:

- In Q1 of 2016, spent 55 kUSD more than expected on gas, but did not know for over two months
- We developed a daily, normalized energy-usage model and visualization of the facility's energy performance.
- The facilities management team uses this tool daily to assess performance and to avoid subsequent problems.
- Allows on-site operators and SE facilities management to detect and correct problems before they go out of control

Daily Energy Model

- Shows performance on a daily basis so actions can be taken more quickly
- One model for gas and one model for electricity
- Predicted daily range in gray, with color-coded performance (details by hovering) to align with goals
- Outside temperature scale color-coded along bottom (details by hovering)



Future

- Robustness – streamlined
- Industrialization – mass producibility
- EcoStruxure architecture – near real time production data
- Full integration into existing software

Life Is On



Schneider
 Electric