Nailing the Peak: City-Scale, Building-Specific Load Factor and Contribution to a Utility's Hour of Critical Generation

Presented at:

IBPSA Building Simulation Conference

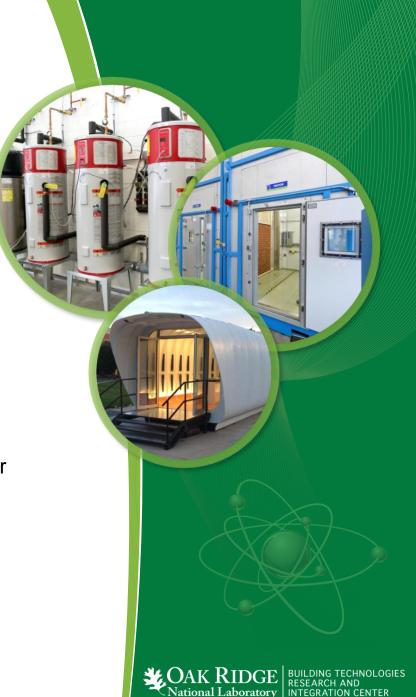
Rome, Italy

Presented by:

Joshua New, Ph.D., C.E.M., PMP, CMVP, CSM Building Technologies Research & Integration Center Subprogram Manager, Software Tools & Models Oak Ridge National Laboratory

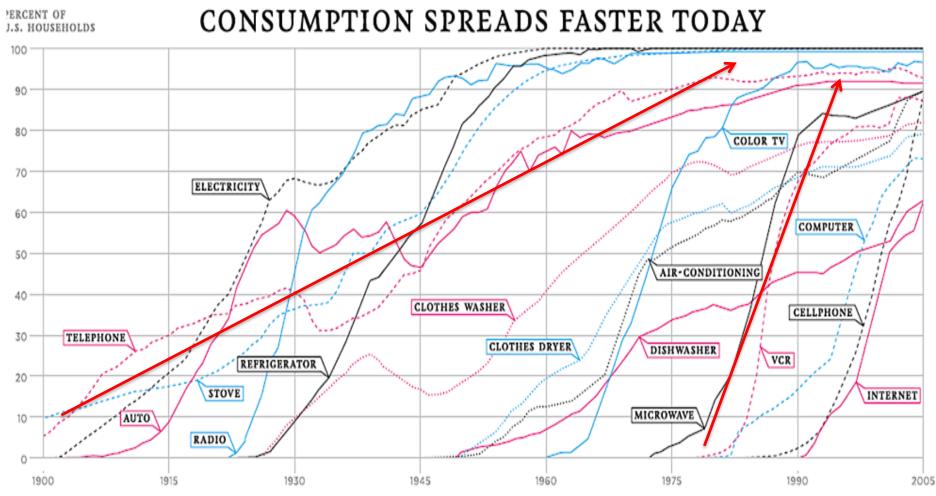
September 2, 2019

ORNL is managed by UT-Battelle for the US Department of Energy





Technology Adoption Rates Accelerate



Nicholas Felton



Wireless Broadband IoT Age Is Upon Us



Papal Conclave 2005

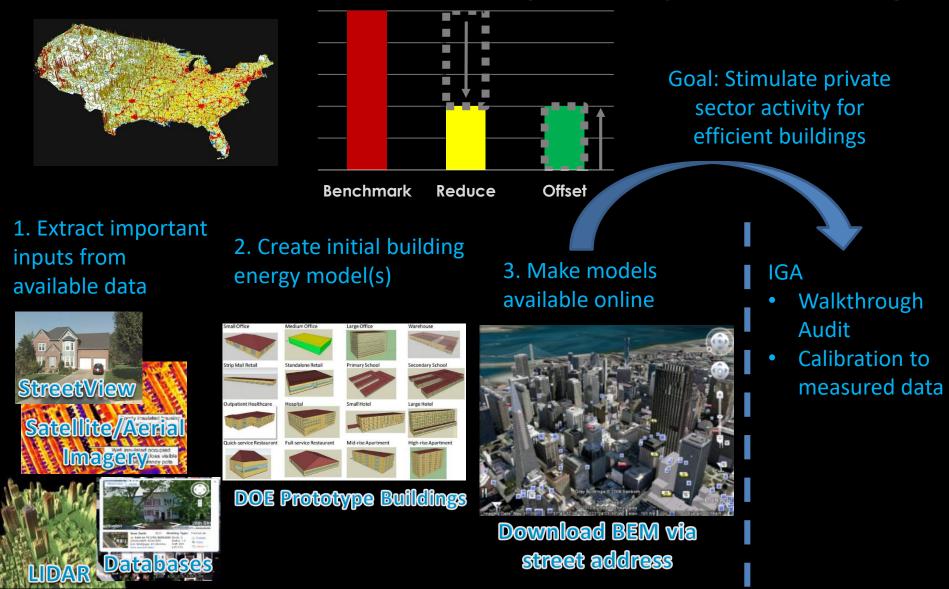


Gigabit Speed Wireless Broadband Coming Soon in 2018-2019



Papal Conclave 2013

Model America 2020 – BEM for every U.S. Building



What matters and how much?

- Sensitivity analysis for all building types
 - 80% of commercial buildings 16 climate zones, 16 building types, averaging 5.75 vintages
 - 281-4,617 building descriptors (e.g. thermostat, insulation level) were modified
 - Fractional Factorial (FrF2) resolution IV statistical design of experiments
- Summarize 768 lists of impactful variables
 - 254,544 annual simulations were completed on the nation's fastest supercomputer (Titan)
 - 216 Excel spreadsheets were created listing the energy and demand impacts of each building property
- Quantify Most Important Building Parameters
 - Top 10 annual <u>energy (kWh)</u> and <u>demand/peak-shaving (kW)</u> variables for each of the 16 building
 - Publication in-review with supplemental Excel spreadsheets for each bldg. type, location, and vintage for 47-470 variables each.

	Small	Outpatien	Large Office	Medium Office	Hospital	Warehous	Small Hotel	Large
	Office	t						hotel
Inputs	458	3483	1072	760	1955	333	1823	887
	Stri p	Retail	Quick Service	Full Service	Mid Rise	High Rise	Secondary	Primary
	Mall		Restaurant	Restaurant	Apt	Apt	School	School
Inputs	800	438	281	286	1464	4617	1621	1051



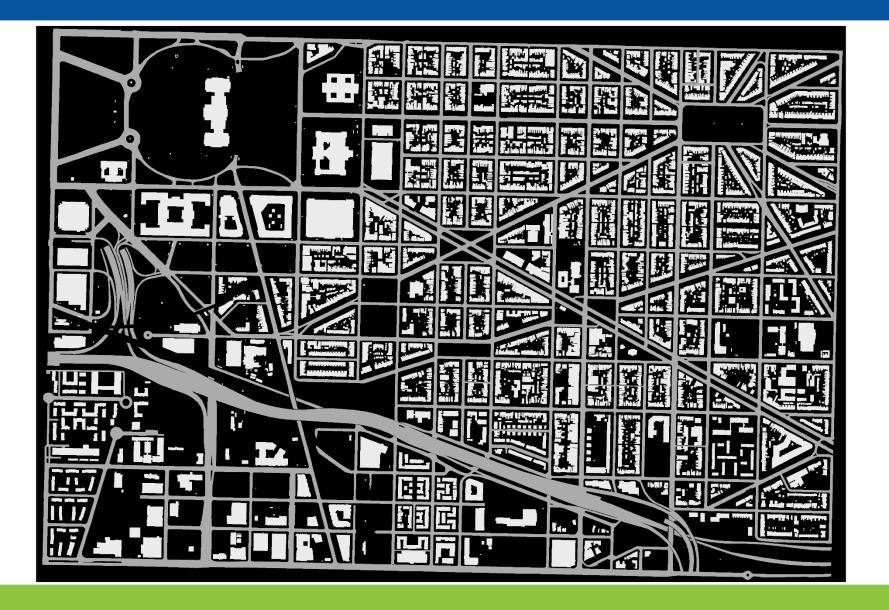
Data Sources

- Database and image sources for urban model generation
 - Satellite and airborne imagery
 - Cartographic data
 - Ground level images

- Elevation data
- Building information databases
- 3D building model databases

	Short Title
Summary	Satellite imagery, including panchromatic and multispectral images
Data type	Image
Company	
Website	
Temporal resolution	Cities - 3-11 times per week
Spatial resolution	0.3 m
Measure accuracy	
Cost	\$11 per sq. km
Format	GeoTiff
Mapping to building input variables	Building footprints
Mapping to area properties	Vegetated areas, road surface, buildings, parking lots
Mapping to material properties	Road pavement materials (e.g., concrete, asphalt), parking lots (e.g., gravel, soil)
Coverage of US	Over 10 million km ² of coverage of the contiguous US
Orientation	Aerial
Existing internal software	N/A
Existing expertise	Remote sensing data analysis tool
Restrictions	N/A
Comments	

Manual Segmentation of DC



Automatic Road Extraction



Automatic Building Footprint Extraction

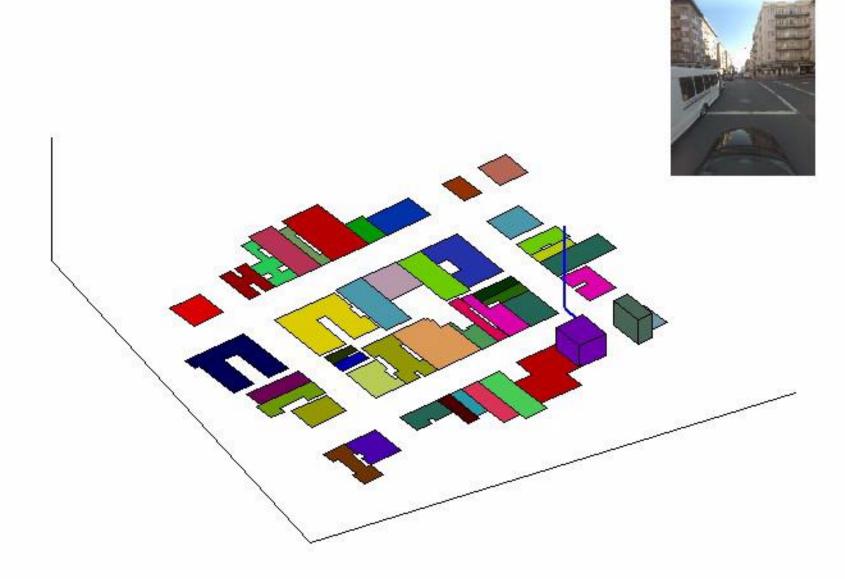
Algorithm: Deep Learning extended and using GPUs for fast building footprint and area extraction over large geographical areas.



Multi-company Competition Precision/Recall – 30/35; Current Precision/Recall – 60+/60+

Processing Street-Level Imagery (Jiangye Yuan)

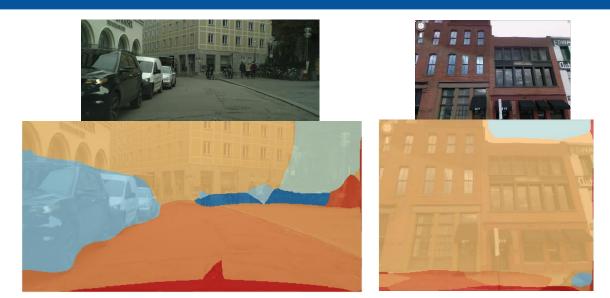
3D Building Model Generation



Street-level imagery (Lexie Yang)

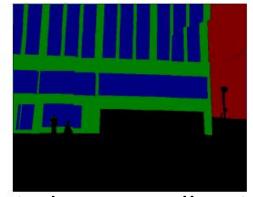
Façade Type

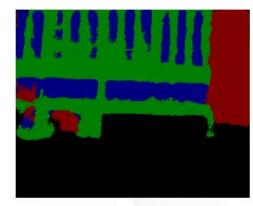




Windows (blue) Façade (green) Street/open (black) Other building (red)







Window-to-wall ratio

Input image

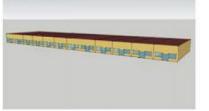
Model output

Prototype Buildings



Strip Mall Retail

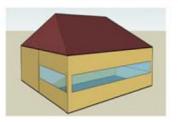
Small Office



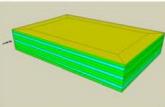
Outpatient Healthcare



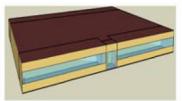
Quick-service Restaurant



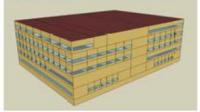
Medium Office



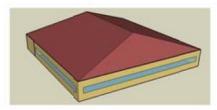
Standalone Retail



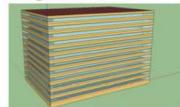
Hospital



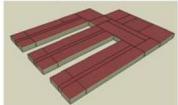
Full-service Restaurant



Large Office



Primary School



Small Hotel



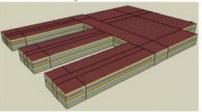
Mid-rise Apartment



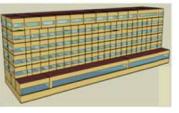
Warehouse



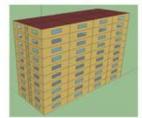
Secondary School



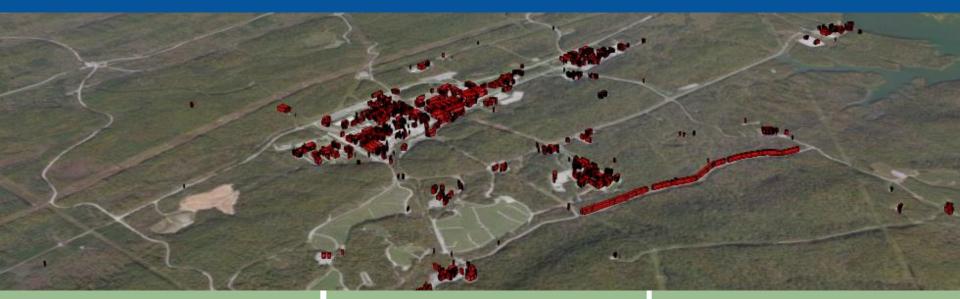
Large Hotel

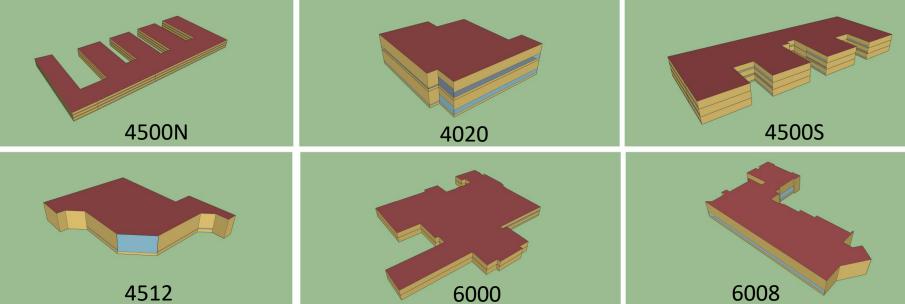


High-rise Apartment



Oak Ridge National Laboratory





Oak Ridge National Laboratory (interactive)

2012

Years

4500N Name: Central Research & Administration North Year Built: 1952 Number of People: 450 Gross Square Footage: 363,980 Number of Floors: 3

Energy Usage (for visualization purposes only, data is inaccurate): 0.9





, URBAN DYNAMICS AK RIDGE NATIONAL LABORATORY

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The University of Tennessee (2 days)



Use Case - Scenarios

- Preliminary building-specific estimates of energy, demand, and cost savings totaling \$11-\$35 million per year based on 9 scenarios prioritized by EPB.
- 1. Peak Rate Structure
 - 1. Scenario #1a, Peak contributions for each building
 - 2. Scenario #1b, Cost difference, in terms of dollars per year, for all building

2. Demand Side Management

- 1. Scenario #2a, Monthly peak demand savings, annual energy savings, and dollar savings based on rate structure for all buildings.
- 2. Scenario #2b, Location-specific deferral of infrastructure cost savings potential

3. Emissions

1. Scenario #3a, Emissions footprints for each building

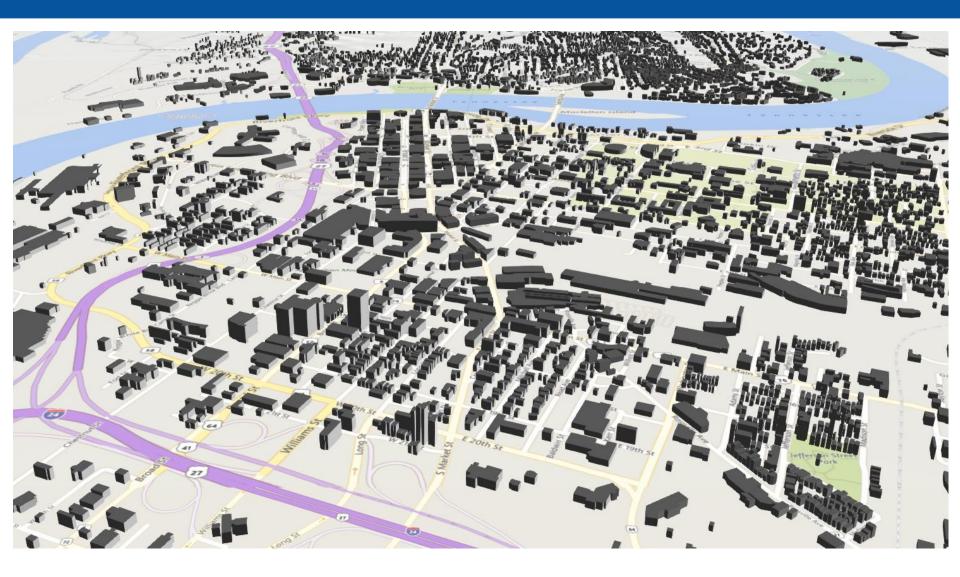
4. Energy Efficiency

- 1. Scenario #4a, Optimal retrofit list of independent ECMs
- 2. Scenario #4b, Optimal retrofit package of dependent ECMs

5. Customer Education

- 1. Scenario #5a, Percentile ranking of each building's EUI by building type and vintage
- 2. Scenario #5b, Monthly peak demand savings, annual energy savings, and dollar savings based on rate structure for all buildings compared to AMY weather file scenario.

Chattanooga, TN (100,000+ buildings)



The AutoBEM technology "axe"

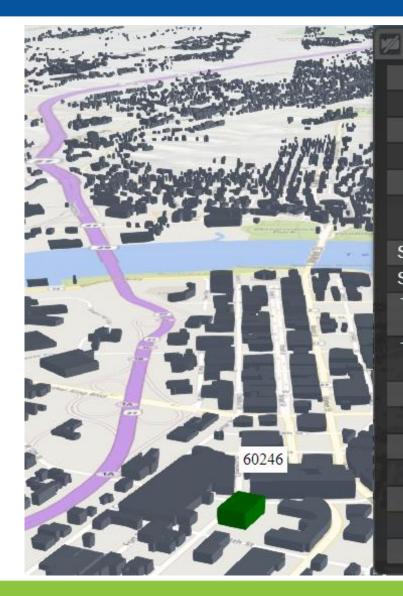
135,481 building models have been created and matched to EPB's PremiseID Limitations: limited building types, not calibrated, will improve quarterly QA/QC: will show how close our simulations are to 15-min data

2.3 million EnergyPlus building energy models using AutoBEM technology, Titan, cloud, and local servers to produce and analyze 13 TB of simulation data.

- 1. Generate baseline building OpenStudio (1.5-3h Amazon, 30h internal)
- 2. Run ECM measures OS Measure (30 mins AWS, 2h internal), Custom (1m AWS, 5m intl.)
- 3. Copy data to Titan 1 min (1.2GB tar.gz)
- 4. Submit to Titan 0-2 hours in queue
- 5. EnergyPlus simulation time 30-45 mins (5mins/sim = 1.4 years to simulate EPB on 1 core)
- 6. Data transfer 40 mins (160GB tar.gz)
- 7. Uncompress 10-15 mins
- 8. Reformat data 20-30 mins
- 9. Analysis 5-10 mins

Time for creation, annual simulation, and analyzing "all" EPB buildings 6.5 hours (6.1h –36.5h)

Virtual EPB (provided by ORNL) shows the value of technology with interactive dynamic results



6024	46 ×	•
ID	60246	
DOE Building Type	SmallOffice	
Num Floors	3	
Percentile	87.70 %	
Estimated wholesale vs rei	tail cost \$ 9797.07	
CO2 emissions	222052.32 Ibs/year	
Smart Thermostat - 4F cost	t savings \$ 1316.61	
Smart Thermostat - 8F cost	t savings \$ 2325.84	
TMY->AMY Smart Thermos cost savings	stat - 4F \$ 204.99	
TMY->AMY Smart Thermos cost savings	stat - 8F \$ 103.41	
HVAC Efficiency ECI	M \$ 1291.79	
Gas HVAC ECM	\$ 4276.69	
Gas Water Heater EC	CM \$ 725.58	ģ
Heat Pump Water Heater	r ECM \$ 476.95	
Insulation ECM	\$ 736.27	
Infiltration ECM	\$ 1577.50	
Lighting ECM	\$ 2898.95	

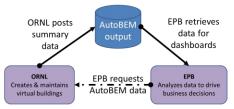
Comparison to real data

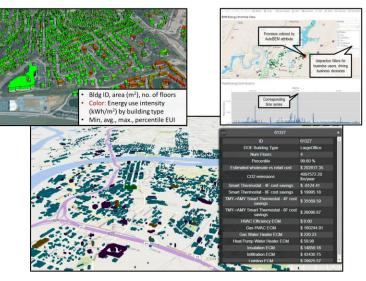
- Empirical Validation
 - 15-minute whole-building electrical for 178,368 buildings
 - More accurate than BEM created by a human¹

Operational Use of BEM Simulations

Use Cases

- Peak rate structure
- Demand-side mgmt
- Emissions
- Energy efficiency
- Customer education





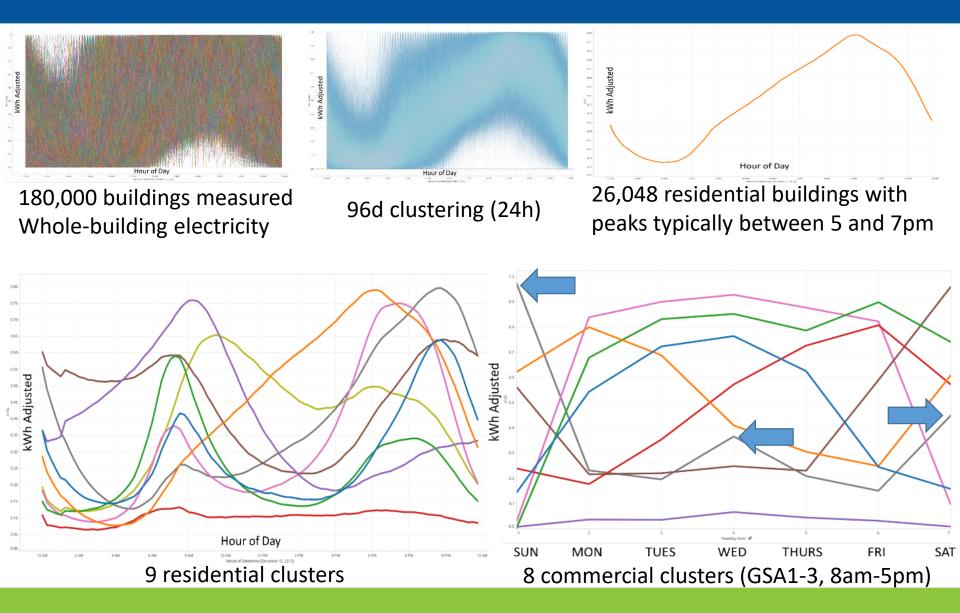
Measures

- Lighting, HVAC COP, infiltration, insulation
- Smart thermostats
- Water heaters
- PV/solar
- EV charging
- Future weather
- Dual-fuel HVAC
- Microgrids

Result: \$11–35 million/year in potential savings identified via simulation-informed data and valuation for *energy, demand, emissions, and cost impact* to EPB and each customer for each building under five use cases covering nine monetization scenarios

¹Garrison, Eric, New, Joshua R., and Adams, Mark (2019). "Accuracy of a Crude Approach to Urban Multi-Scale Building Energy Models Compared to 15-min Electricity Use." Best PhD Student Paper award. In *Proceedings of the ASHRAE Winter Conference*, Atlanta, GA, Jan. 12-16, 2019. [PDF] [PPT]

Clustering of (real) 15-min electrical data



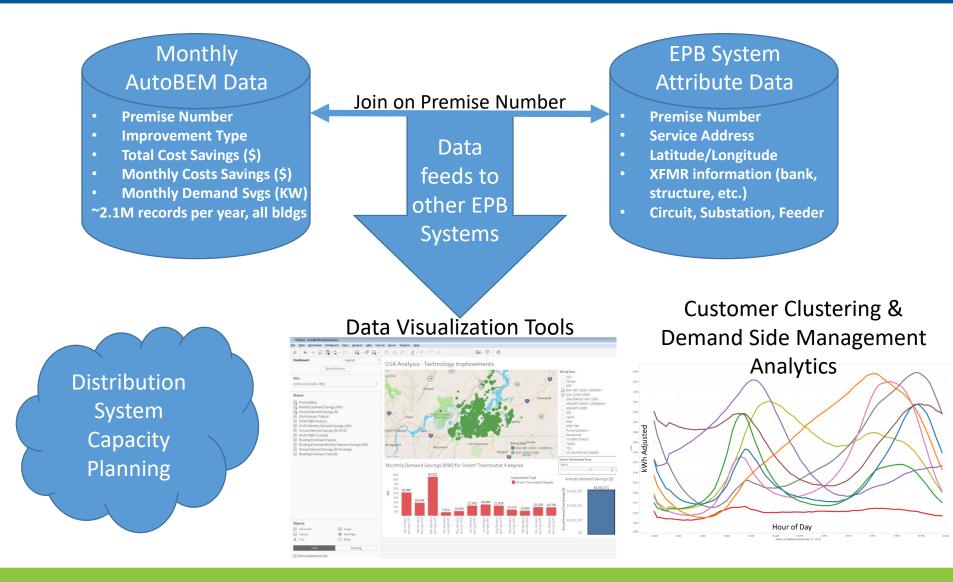
Load Factor summary

- Utility load factor LoadFactor = Total(kWh) / (kWpeak * numHours)
 - Close to 0, more opportunity for energy storage

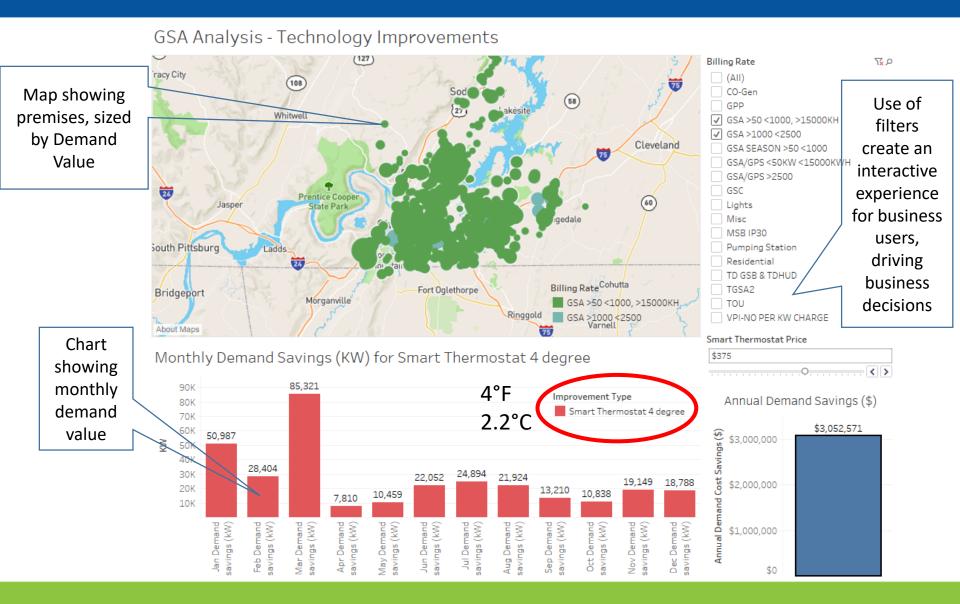
	Vintage	Num Bldgs	% of all Bldgs	Avg. Load Factor
Re	2006	16217	9.1%	0.170
Residential	2009	6357	3.6%	0.177
tial	2012	149247	84.0%	0.163
Commercial	Pre-1980	670	0.4%	0.405
	1980-2004	1064	0.6%	0.296
	90.1-2004	1478	0.8%	0.255
	90.1-2007	268	0.2%	0.338
	90.1-2010	1224	0.7%	0.208
	90.1-2013	1808	1.0%	0.256

Building Type	Num Bldgs	% of all Bldgs	Avg. Load Factor
IECC Residential	171821	96.35%	0.164
Warehouse	799	0.45%	0.166
MidriseApartment	851	0.48%	0.261
SmallHotel	1557	0.87%	0.261
HighriseApartment	2068	1.16%	0.263
LargeHotel	408	0.23%	0.365
QuickServiceRest.	318	0.18%	0.380
Hospital	319	0.18%	0.399
Outpatient	59	0.03%	0.501

Database status (new): AutoBEM Monthly Data joined with EPB System Data



EPB's operational Business Intel. Dashboard



Discussion

Joshua New, Ph.D., CEM, PMP, CMVP, CSM Subprogram manager for Software Tools & Models Building Technologies Research and Integration Center (BTRIC) Oak Ridge National Laboratory

newjr@ornl.gov

Bill Copeland

Director, Business Intelligence Electric Power Board of Chattanooga, TN HPC Tools for Modeling and Simulation Capturing building energy consumption

CAK RIDGE National Laboratory

THE UNIVERSITY OF

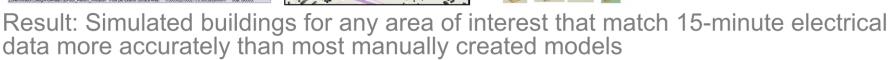
Data source and software overview

Automatic Detection and Building Energy Model Creation (AutoBEM)

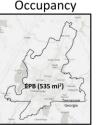
Data Sources

- Imagery (satellite, aerial)
- Street-level imagery
- Cartographic layers
 - Elevation, GIS
- Tax assessors
- Ranking of descriptors EE and Demand impacts (281–4,617 per building type)





Da	ta comparison matrix	Occupancy
	Short Title	
Summary	Satellite imagery, including panchromatic and multispectral images	and in 1/a and a the
Data type	Image	12
Company	in the second seco	
Website		and the first
Temporal resolution	Cities - 3-11 times per week	
Spatial resolution	0.3 m	
Measure accuracy		312 (
Cost	\$11 per sq. km	and a second and the
Format	GeoTiff	
Mapping to building input variables	Building footprints	mit in mon
Mapping to area properties	Vegetated areas, road surface, buildings, parking lots	£PB (535 mi ²) - /
Mapping to material properties	Road pavement materials (e.g., concrete, asphalt), parking lots (e.g., gravel, soil)	EPB (535 mi-)
Coverage of US	Over 10 million km ² of coverage of the contiguous US	+
Orientation	Aerial	Tennessee
Existing internal software	N/A	Georgia
Existing expertise	Remote sensing data analysis tool	1.5
Restrictions	N/A	- Long
St	reet-level data	Building footprints
Brick V Façade	Vindow-to-wall ratio	
		Stall Office Large Office Wanthave



Software Tools

- Occupancy (every 90m)
- Aerial best footprints
- Street height, type, WWR
- LiDAR geometry
- GIS database API
- Building type
- Model generator
- Fastest buildings simulator
- Web-based visual analytics

Virtual NYC – interactive results

12 FAILHING	1921	Bldg_12210017046				
	ECM	Annual Electricity/Savings	Jan Demand/Savings	Feb Demand/Sa		
a strange of the second s	Baseline	41282530.14 kWh	8463.14 kW	8426.13 kW		
	Change Elec Base COP	0.09 kWh	155.27 kW	3232.78 kW		
	Change Lighting Power Density	2796916.23 kWh	975.65 kW	3597.39 kW		
	Change Roof Insulation	688072.96 kWh	267.15 kW	3348.45 kW		
	Change to Elec Water Heater	-24140.75 kWh	152.50 kW	3230.01 kW		
	Change to Gas Water Heater		155.27 kW	3232.78 kW		
	Change Space Infiltration	411236.67 kWh	176.59 kW	3436.11 kW		
And the second the sec	Smart Thermostat	14573.47 kWh	155.27 kW	4155.11 kW		
of the second se	the state	Sanor	NOSONH	10724		

1

Virtual NYC – interactive results

Savings across 152 buildings

E=e	energy (MWh), D=demand (k	W), [min,avg,max]
1.	Smart thermostat 2.2C (4F) E=[-72, 1.4, 525]	pre-condition D=[-938, 918, 13907]
2.	Natural gas water heater (8 E=[0, 0, 0]	0% efficient) D=[0, 772, 13907]
3.	Heat pump water heater (C E=[-184, -16.4, -2]	C OP 2.2) D=[-30, 768, 13853]
4.	HVAC Efficiency (COP_H 3.55 E=[0, 0,0]	and COP _c 3.3) D=[0, 772, 13908]
5.	Lighting Efficiency (0.85 W/ E=[77, 784, 6757]	′ft²) D=[23, 999, 14410]
6.	Infiltration (reduce 25%) E=[40, 774, 4648]	D=[-0.8, 840, 14020]
7.	Insulation (R16.12 to R28.5 E=[12, 204, 1600]	7) D=[1.9, 817, 13928]

building_id	Elec_savings (kWh)	Jan (kW)	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep	Oct	Nov	Dec
1221000024	7 -25559.37	5852.94	4309.38	5366.97	6592.65	4262.77	870.72	457.84	431.46	2406.69	6587.84	6630.63	6242.57
1221000046	9 -721.30	25.08	89.07	156.48	6.73	26.60	168.46	150.48	4.51	7.44	6.22	6.39	12.52
1221000051	8 -2701.65	5 7.76	1762.35	3394.86	22.85	550.36	2560.03	2263.61	2697.12	709.85	75.10	74.33	0.00
1221000099	4 -8481.05	1055.41	905.72	990.42	1474.41	848.47	196.65	194.14	3.66	548.06	1276.96	1252.69	887.56
1221000115	6 -5736.16	5 1196.35	1000.11	1101.04	1673.10	954.47	222.70	215.54	216.98	671.15	1469.96	1418.45	973.81
1221000119	7 546.94	¥ 77.19	407.63	1004.80	8.67	143.51	888.66	830.73	901.23	227.86	8.38	69.38	0.00
1221000125	2 -42452.78	3 7440.43	5315.70	6722.96	8265.02	6113.35	511.76	625.95	1.15	4142.70	8232.69	7282.54	6572.38
1221000149	0 -905.68	8 8.48	39.99	177.69	8.06	30.01	16.00	155.19	158.46	81.12	3.35	0.64	3.88
1221000200	1 -16751.35	2353.22	1774.32	2166.91	3481.47	2139.86	229.42	419.27	3.45	1610.13	3238.90	2776.87	1857.80
1221000203	1 -1226.89	1.96	145.84	444.87	-0.66	9.45	-496.24	63.75	116.22	-111.25	10.52	49.65	0.00
1221000204	7 -692.42	22.52	99.76	370.75	9.39	64.11	347.40	349.46	382.89	179.84	11.41	3.25	0.00
1221000215	0 30.35	60.50	255.99	806.49	3.48	66.73	812.78	689.28	5.88	147.68	3.96	77.14	0.00
1221000262	9 -3701.09	882.04	839.96	797.77	1059.83	717.52	153.35	140.93	1.94	482.84	970.80	1050.46	741.97
1221000320	0 -28557.63	2563.00	2406.68	2510.02	2962.48	2202.89	357.32	419.56	410.43	1519.44	2886.95	3028.13	2298.47
1221000329	2 -1583.53	4.30	1311.22	2826.54	9.47	448.83	2200.25	1916.39	2278.23	611.96	56.75	85.90	0.00
1221000330	2 -5519.82	119.63	2140.16	4235.07	33.48	608.39	2922.95	2597.64	2876.36	866.10	103.19	9.61	0.00
1221000331	4 -5708.34	4.91	2444.18	3982.02	0.71	264.42	1733.69	315.91	49.63	393.20	12.85	158.36	0.00
1221000331	7 -3372.72	111.52	510.91	1173.94	9.19	255.62	1086.71	950.19	1091.03	408.28	31.76	9.90	0.00
1221000333	3 -1604.88	3 1.96	84.76	221.98	5.28	6.44	21.88	212.90	20.55	82.06	2.34	5.21	0.15
1221000334	6 -2131.77	5.41	1474.50	3019.75	11.03	474.57	2247.52	1996.23	2414.45	577.55	42.52	74.95	0.00
1221000335	0 -891.15	54.48	119.26	163.76	6.54	16.33	187.35	170.99	3.96	9.51	5.52	7.50	9.59
1221000335	4 -17125.53	4898.93	4556.63	4387.42	5461.75	3512.42	38.73	682.57	661.52	2552.29	5508.01	5809.67	4086.09
1221000336	1 -1214.74	9.77	58.91	210.65	0.63	6.87	-5.49	191.78	204.65	97.66	2.19	1.29	0.45
1221000337	9 759.09	63.94	308.85	797.63	4.71	97.05	791.79	701.29	669.47	6.47	4.73	63.31	0.00
1221000338	3 2626.09	118.49	681.87	1367.29	25.09	224.15	1207.86	1043.81	2.29	341.70	64.43	95.22	0.00
1221000366	1 -42.89	7.47	88.91	223.08	6.93	15.57	232.71	212.10	5.81	1.05	10.39	4.12	0.09
1221000379	1 -676.58	3 1.83	108.79	273.89	4.62	-0.71	266.11	11.88	10.69	51.54	12.46	30.17	0.21
1221000411	5 2116.15	22.16	265.01	744.60	3.08	96.42	759.76	510.54	57.95	57.73	6.83	49.82	0.00
1221000420	5 1070.38	8 82.79	545.31	1222.73	3.25	160.14	1072.99	873.94	14.49	48.06	8.56	99.29	0.00
1221000422	3 380.10	38.15	95.93	354.41	15.05	76.17	337.95	319.08	361.81	192.99	36.14	1.04	0.00
1221000440	6 -2142.06	5 123.81	557.33	1385.46	15.95	278.12	1199.00	1104.96	1282.00	403.57	12.01	11.16	0.00