# **Natural gas analytics**

Analytics for leak detection forecasting and workforce optimization in natural gas

Spring 2023



POWERING WHAT'S NEXT



# **Speaker introductions**



#### **Devon Grodkiewicz**

Technology Analyst E Source Data Science



**Mark Knox** Retired Director, Forecasting Southern Company Gas



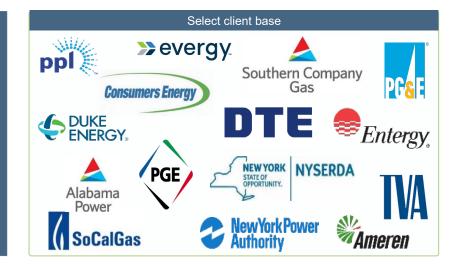


#### **E Source Data Science**

Data scientists with deep AI expertise *and* utility domain knowledge

50+ sophisticated AI models that unlock utility data, savings, and insights

Serving 30+ utilities, each with proven results



\$5 million-\$10 million ... annual asset maintenance savings

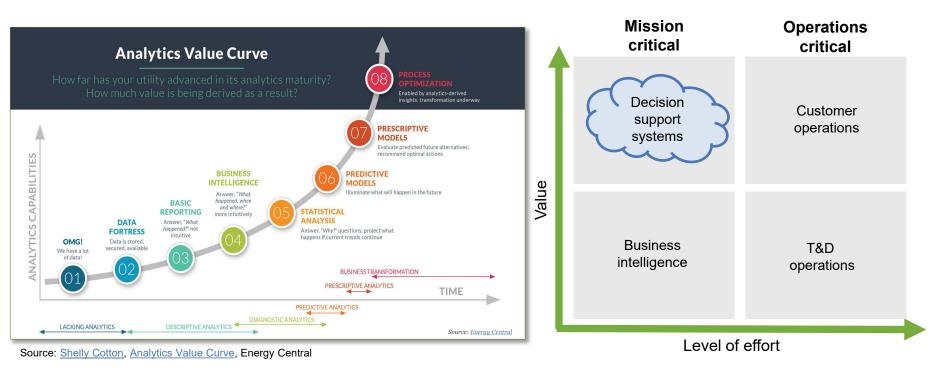
75% + ... reduction in utility protective device expense 84% ... accuracy on structural vulnerabilities 48% ... reduction in deviation of actual-toscheduled hours



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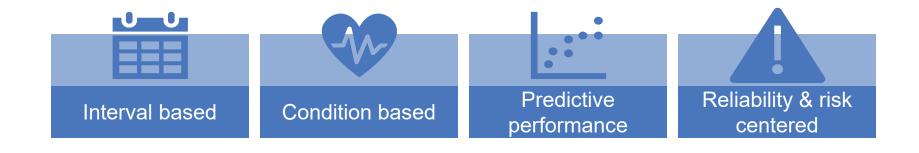
(10%–20% reduction)

## The data science tech paradigm shift Data-driven decision-making





# **Evolution of maintenance with technology**



Progression of change



#### Benefits and challenges of predictive analytics



#### Improve operations and decision-making





Optimize service offerings

Increase profits

## Challenges

Unclear goals

Incomplete/inaccurate data

Unpredictability of human behavior

Requires consistent data uploads

#### **Predictive analytics: Focusing on what matters**

Predictive analytics provide insight and support in achieving operational goals and strategic initiatives?

#### Sustainability

Safety

#### **Customer experience**

#### System reliability



## Leak pressure: From the IRA to regulators



# Lost gas alone is worth an estimated \$2 billion each year.

**E**Source



#### Southern Company Gas

#### **E Source client case study**

Southern Company Gas serves approximately 4.2 million natural gas utility customers through its regulated distribution companies in four states and more than 600,000 retail customers through its companies that market natural gas.

- Across the Southern Company Gas service territory, leaks are discovered or encountered every day.
- To proactively repair services, field workers would use a gas detector and manually walk the lines to identify areas of concern.
- This process was very labor-intensive. Staffing and budgeting for the detection and repair of leaks is typically done by manual estimation.
- To meet forecasting, budgeting, and planning requirements, staffing decisions often must be made a year in advance.





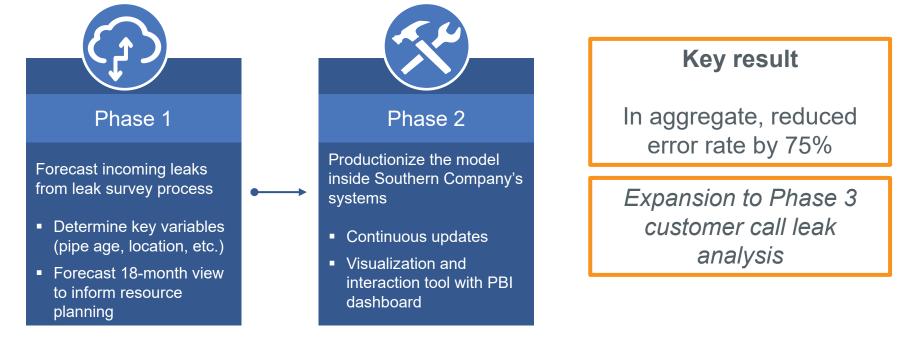
#### Enhanced effectiveness of resource and budget planning

If the team at Southern Company Gas reduced overestimated manpower for leak detection by 5%, annual cost optimization opportunities could be expected to reach \$500,000+ per year. Reducing underestimated manpower by 5% could optimize \$300,000+ per year.



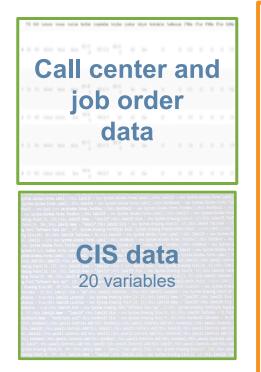
# **Survey leak forecasting: Overview**

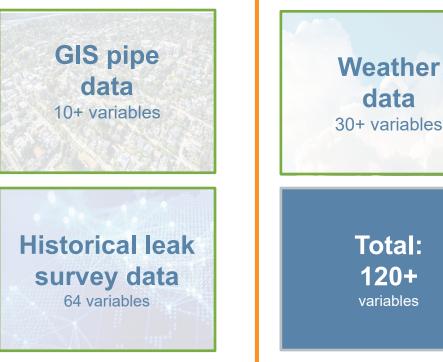
**Null hypothesis:** Predicting future leak survey results relying on historical leak survey data will produce a "just as good" result as relying on predictive analytics





## **Data inputs**





30+ variables Total: 120+ variables



#### **Predictors considered**

Data variable		
Region	Survey completion time	Correct grid number
Company	Survey status	Odd riser location
Meter location	Leak grade (1, 2, 3)	Comments
Service center	Leak type (meter, service, main)	Ad-hoc status
Grid #	Leak location code	Map page / grid
Survey interval	Leak details	Customer conversation
Premise address	Riser data (type, size)	Date-time data of survey
Appt #	Curb box	Numeric data*
City	Map correction reason	Grid survey schedule data
State	Correction comments	Compliance data
Survey type	GIS pipe data by service center	

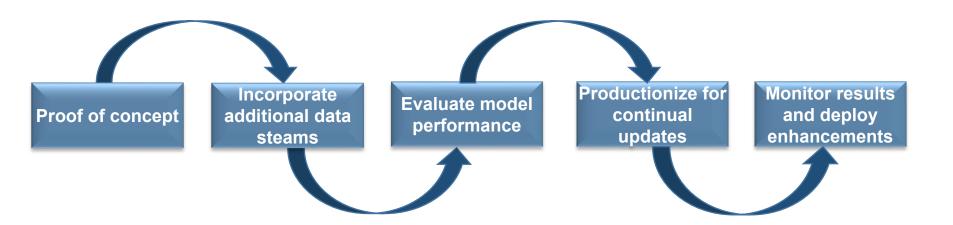


### **Major predictors utilized**

	Data variable	
Region	Survey completion time	Correct grid number
Company	Survey status	Odd riser location
Meter location	Leak grade (1, 2, 3) -1 not utilized	Comments
Service center	Leak type (meter, service, main)	Ad-hoc status
Grid #	Leak location code	Map page / grid
Survey interval	Leak details	Customer conversation
Premise address	Riser data (type, size)	Date-time data of survey
Appt #	Curb box	Numeric data*
City	Map correction reason	Grid survey schedule data
State	Correction comments	Compliance data
Survey type	GIS pipe data by service center	*Numeric data corresponds to corrosion type, repair needs, soil subsiding data, bollard data, 30+ numerical leak data points.



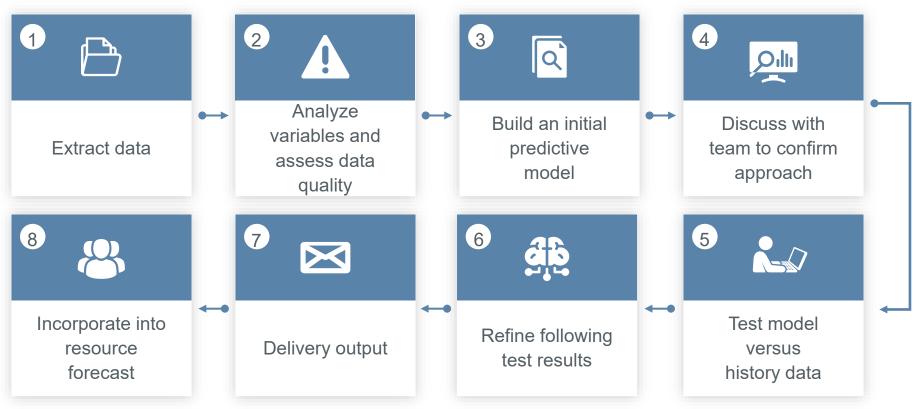
## Data science approach



Progressively move toward a more fully automated Al-driven tool that predicts leaks in real time, staffs accordingly, and re-routes a smaller, more efficient group of field workers to most efficiently manage the end-to-end leak management process.



### Gas leak survey prediction modeling process



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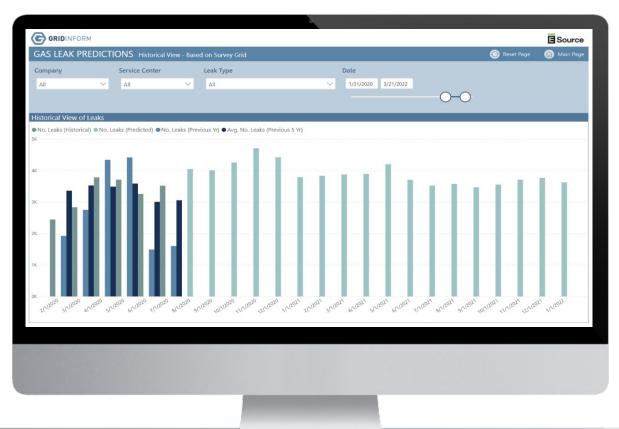
#### Gas leak predictions: Year-to-date view



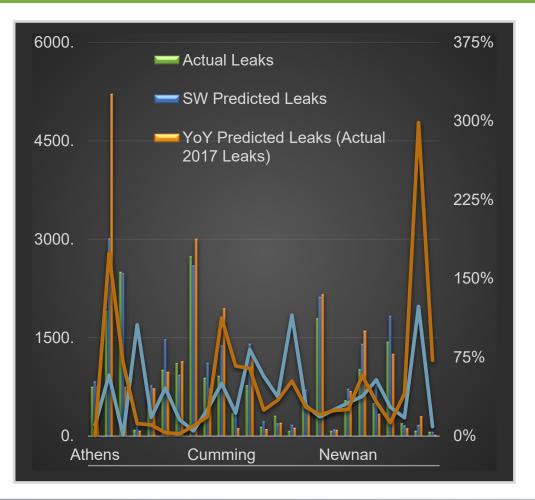
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## **Gas leak predictions: Historical view**







# Results: Prediction accuracy

- ✓ Predictive analytics were within 5.8% of the total number of leaks detected in 2018
- ✓ The model did better than year-over-year predictions for annual leaks detected
- ✓ At an individual service center level, the model cut predicted error in half



# Predictive analytics' effectiveness in achieving desired outcomes



Improved directional accuracy of Southern Company Gas' forecast enables more effective allocation of budget funds across the enterprise due to enhanced planning of field overtime and/or sourcing activities



The predictive algorithm was used as a key input in generating 2021 and subsequent annual leak repair forecasts



Testing identified marginal error improvements in overall forecast for leak repairs of 15% to 20% (based on leak repair types)



Estimated budget allocation benefits of improved leak repair projections could be more than \$0.2 million (based on marginal error improvements)



Allowed for effective determination and validation of proactive leak repair strategies, while supporting other key enterprise-wide goals and strategies

**H**Source



# Exploring the possibilities with predictive analytics

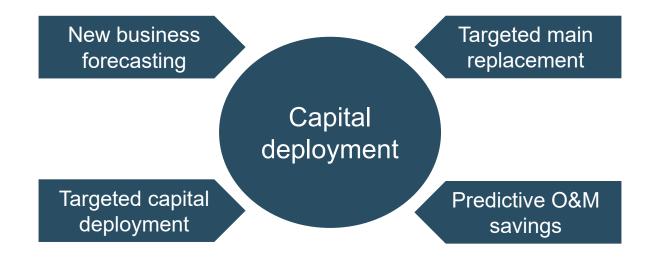






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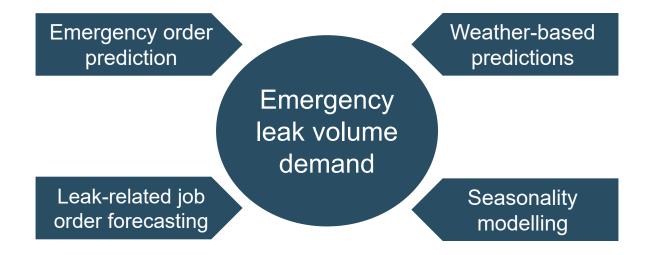






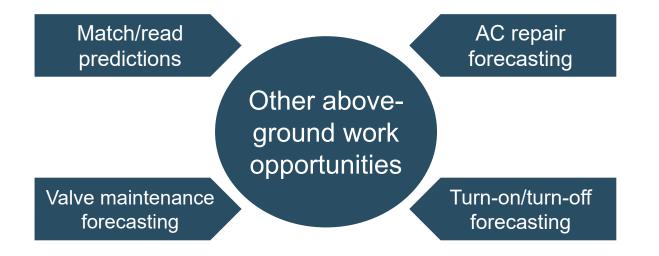
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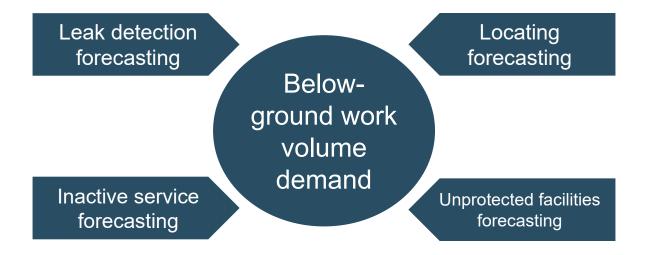




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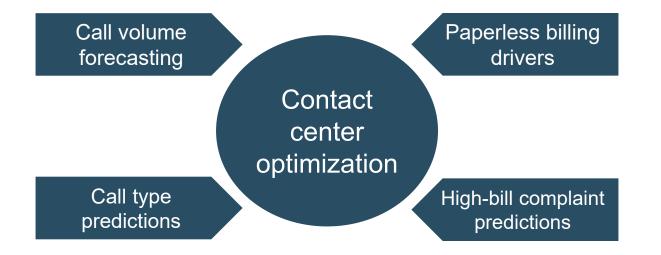
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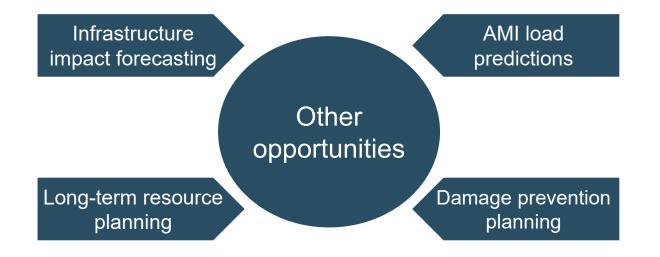




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