

Self Service Analytics for Processing of Hydrocarbons

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Burnaby Refinery - who are we?

- Burnaby Refinery:
 - Located near Vancouver, BC, Canada
 - We provide 25% of British Columbia's gasoline and diesel needs
- 80+ years of history:
 - 1935: Refinery built and operated by Standard
 Oil of California (present-day Chevron)
 - 1950s -> 11,000 bpd
 - 1970s -> 35,000 bpd
 - 2017: Assets sold to **Parkland Corporation** -Canadian independent fuel retail company
 - Today -> 55,000 bpd
 - In Progress -> Expanding range of renewable biofeedstocks (incl. tallow, canola) to lower GHG emissions and carbon footprint



5 cases studies on self-service analytics – empowering SMEs



How we use and benefit from Seeq:

- Used by Engineers (Technical Dept.) to support Operations with daily monitoring, incident investigation etc.
- Helps engineers save time, eliminate low-value work (e.g. data cleaning) and reduce analysis friction

Case Study 1: Conditional Filtering of Large Process Datasets



 Filtering large process datasets in spreadsheets is a very time-consuming and tedious process



SOLUTION

• Used Seeq's Value Search and Chain View to easily apply the right filters and hide irrelevant data



RESULTS

- Completed analysis in Seeq within an hour, compared to >40 hours in previous failed attempts using spreadsheets
- Time saved used for performing highervalue work (instead of data cleaning) and improving quality of investigations



Process background: anomalies in flow measurements

• During HAZOP, found trip SP set too low:

- Engineering team identified risks with the Main Air Blower (MAB) low flow trip setpoint being set too low and recommended a higher setpoint
- Tried to raise setpoint, but...
 - Received pushback from Operations:
 - Observed random dips in the flow measurements
 - Concerned that a higher setpoint may cause spurious trips in the unit
- Would we trip the unit? Need data:
 - Due diligence: Find out how often these flow anomalies happen, and why before making further changes



Challenges with filtering in spreadsheets

- How much data do we need?
 - Dips seem to happen sporadically (every few months)...
 - Dips may only last a few seconds...
 - Want to review 5-10 years of high-resolution data
- What are the challenges in spreadsheets?
 - SLOW, HUGE FILES.
 - Formulas and filters on millions of rows
 - PI SYSTEM EXPERTISE. Can we just write queries?
 - No. Investigation team was only familiar with Excel, not comfortable with PI queries, SQL or Python scripts.
- Initial spreadsheet attempt took >40 hours:
 - Played with historian settings, but didn't trust the data
 - Is PI averaging the values? Do I need higher resolution?
 - Hit Excel row limit, had to split into multiple files



Solution: conditional filtering in Seeq

- Successful attempt in Seeq took just 1 hour
 - Easily defined Composite
 Condition and adjusted Investigation
 Range as needed.
 - Avoided the need to tweak historian retrieval settings and run IF-ELSE formulas in Excel and filters over millions of rows and multiple files
- Technically, could've been faster!
 - Team just started our Seeq trial and was still learning how to use the tool



Business results for conditional filtering in Seeq

• Conclusions

- Rapidly and correctly identified all time periods with anomalies in Seeq
- Seeq analysis provided confidence to proceed with setpoint changes.
- **Business Results & Implications**
- Spurious trip would cost ~\$1mil per day of downtime.
- Not proceeding with change would've resulted in unacceptable safety risks.



Source: Siang, L. C., Elnawawi, S., & Steele, D. (2022). Self-Service Analytics and the Processing of Hydrocarbons. *Digital Chemical Engineering*, 100021.

Case Study 2: Inferential Performance Assessment



- Inferential models are used for estimating process variables in the absence of online measurements.
- Need to monitor model performance, but difficult to align timestamps for prediction vs. lab



SOLUTION

• Use Seeq *Capsules* to easily re-align timestamps for all samples and calculate prediction error.



RESULTS

- Avoid the need to manually align sample timestamps.
- Scale calculations to multiple inferentials plantwide using *Asset Trees*.

Source: Curreri, F., Fiumara, G., & Xibilia, M. G. (2020). Input selection methods for soft sensor design: A survey. *Future Internet*, 12(6), 97.

Calculation background 1: residuals method

- How do we measure model performance?
 - Simplest method: take the absolute difference between predictions and lab (a.k.a. *error* or *residuals*).



Source: Siang, L. C., Elnawawi, S., & Steele, D. (2022). Self-Service Analytics and the Processing of Hydrocarbons. *Digital Chemical Engineering*, 100021.

Calculation background 2: aligning prediction vs. lab

- Sounds easy. Why is it difficult?
 - Data Cleaning! Samples collected at the prediction time <u>(red dot)</u>, but lab results obtained hours later <u>(blue dot)</u>.
- Must realign timestamps for inferential predictions vs. lab results
 - Must shift blue dot to align with red dot
 - Shift varies for each sample, depending on how long the test takes, how busy the lab is, etc.
 - It is not just a simple constant time shift.
- Easy to manually shift one sample, but what if you have many, many samples?



Source: Siang, L. C., Elnawawi, S., & Steele, D. (2022). Self-Service Analytics and the Processing of Hydrocarbons. *Digital Chemical Engineering*, 100021.

Calculation background 3: indicator function to capsules

- How do we know the time shift?
 - Use an indicator function configured in DCS: gives us the ΔT between the sampling time and lab results
 - Signal = 0 initially. Step change from 0 to 1 indicates sample taken
 - Signal = 1 while waiting for results.
 Step change from 1 to 0 when results are obtained
- Solution:
 - Step 1 Convert indicator function signal to Seeq capsules
 - Step 2 Use the capsules to align the prediction vs. lab values.



Source: Siang, L. C., Elnawawi, S., & Steele, D. (2022). Self-Service Analytics and the Processing of Hydrocarbons. *Digital Chemical Engineering*, 100021.

Solution for inferential calculations in Seeq



- Step 1 Convert lab and prediction signals to scalars bounded by indicator function capsules
 - First value of \$pred in \$A is when the sample is taken:
 - Last value of \$lab in \$A is when lab results are returned:

\$pred.toScalars(\$A).first()
\$lab.toScalars(\$A).last()

Solution for inferential calculations in Seeq



- Step 2 Difference in prediction vs. lab is easy to calculate as a Seeq formula
 - Previous formulas created 2 new, cleaned signals. Predictions and lab results are now aligned.
 - Ready for residual calculation.

Solution for inferential calculations in Seeq



• Step 3: Putting it all together - Residuals calculation is a simple one-liner in Seeq

<pre>abs(\$lab.toScalars(\$A).last() -</pre>	<pre>\$pred.toScalars(\$A).first())</pre>
Lab	Prediction

Results for inferential calculations in Seeq



• Results

- Avoid manually aligning timestamps in Excel by leveraging *capsule properties*
- Use the calculations as a template. Easily scale to other inferentials by defining Asset Trees (Groups)
- Conclusion: Simple tasks can be tedious without the right tools! Users often don't need/want machine learning/advanced algorithms (depending on the problem), but just the ability to quickly access, filter and clean our data. Seeq makes it easy to do all these -> empowering engineers to quickly get the results they need.

Case Study 3: Incident Investigation



- Incident investigation for flooded boiler
- Fuel gas system has 16+ inputs Excel and other data historians are time consuming to pull data for analysis.



SOLUTION

 Use Seeq to pull large amounts of data quickly and use Value Search to identify parameters outside of normal operating range.



- Identified culprit fuel gas producer and improved start-up procedures based on results.
- Incident investigation took 1 day instead of 1 week of analysis due to Seeq.



Results of incident investigation

• Use Seeq to pull data for 16+ inputs

- Easily trend everything rather than pulling data then trending in Excel
- Quickly add limits and other visual cues for presentation to stakeholders
- Identified the culprit stream and events leading up to incident using Value Search
- Conclusion and Business Results
 - Identified culprit fuel gas producer and improved start-up procedures based on results.
 - Incident investigation took 1 day instead of 1 week of analysis due to Seeq.



Case Study 4: Flowmeter Correlation



- Steam flowmeter for compressor turbine broken for 2+ years.
- Equipment is a large 400# steam consumer and required for steam balance / energy saving initiative.



SOLUTION

• Use Seeq XY Plot and Model & Predict to build a correlation between the turbine governor valve and steam flow using data from 2016-2018.



 Correlation used for steam reduction test run and daily monitoring of approximate steam consumption using the live governor position.



Figure 1 Turbine Speed Control

Source: Ismail, M.M. (2012). Adaptation of PID controller using AI technique for speed control of isolated steam turbine. *Japan-Egypt Conference on Electronics, Communications and Computers*

Results of X Y correlation

Use Seeq to build correlation

- Quickly pull years of data to build a good correlation.
- Seeq automatically color codes by year, ensuring any drift or step changes in correlation are captured.
- Conclusion and Business Results
 - Plot created in ~5 minutes.
 - Relationship had an r2 = 0.973.
 - Rather than generating coefficients in excel, submitting an IT ticket to PI for a calculation tag, and then monitoring on PI/XHQ, easily create a live calculation value in Seeq energy tracking dashboard.
 - Results in 1 day rather than weeks and save an IT resource and PI Tag slot.



Case Study 5: Flare Baseline Flow



- Flaring above design purge requirements wastes valuable energy and increases
 GHG emissions.
- Hard to tell what the target is without a proper baseline.



SOLUTION

- Use Seeq's Value Search to find times when all equipment is on and at high rates.
- Use Seeq's *Signal from Condition* to find the average baseline flow.



- Analyzed >10 years' worth of data in short amount of time
- Identified and mitigated \$500,000 worth of excess flaring in 2022



Results of flare analysis

- Establish Minimum Actual Rate
 - Minimum theoretical rates set by purge requirements of 1 ft/s.
 - Minimum actual rates will be higher always some extra flow from passing PSVs, etc. that cannot be fixed until the turnaround
 - Minimum actual may change depending on frontend rates and which units are online
 - Value Search and Signal from Condition easily identify these periods, rather than manually searching/conditioning in Excel
- Conclusion and Business Results
 - Determined baseline for flare flow anything above this requires investigation and mitigation
 - Identified and mitigated \$600,000 /year worth of excess flaring in 2022



Conclusion

- Seeq is:
 - Reducing the amount of time to analyze data
 - Increasing the quality of the conclusions drawn
 - Expanding opportunities to troubleshoot and optimize
- Where are we going?
 - Increase Seeq literacy in the Technical group
 - Build more tools for continuous improvement
 - Use as a link from PI System data to other interfaces (eg. Power BI) because of ease of calculation edits

Seeq excel watching me use google sheets





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