

Refinery Optimization

Recent Advances in Planning and Blending Operations

March 19, 2013

Dimitrios Varvarezos Senior Director, R&D

Overview

- **Introduction to refinery optimization**
- **Optimal crude selection and refinery planning**
- **Product blending optimization**
- **Conclusions**

Introduction to Refinery Optimization

- **Historical perspective**
- **Planning, 70s** $-$ LP \rightarrow SLP \rightarrow MINLP
- **Advance process control, 80s** \mid P
- **Real-time optimization, 90s** $-$ NLP \rightarrow SOP
- **Blending and scheduling, 90s**
	- $-$ NLP \rightarrow MINLP
- **Supply chain, 90s**
	- MILP

Petroleum Supply Chain Scope

Refining Process

Definition and Motivation

- Refinery Planning is a Work Process that involves multi-period, nonlinear optimization, including discrete decisions
- Goal is to decide the optimal set of crudes to be purchased
	- Crude valuation
	- Demand affects price and availability
	- Optimization horizon 2 6 months
- This process is enhanced by numerous ad-hoc scenario evaluations
	- Price scenarios
	- Unit availability scenarios
- **There is no comprehensive framework that will:**
	- Analyze the nonlinear solution space for decision variables
	- Consider multiple objectives
	- Ensure a global solution in terms of math programming

New Planning Process Highlights

- A new comprehensive way to integrate, expand, and analyze the optimization solution of a refinery planning model in the context of a complex business decision making process
- **Introduce rigorous and automated analysis that identify optimal** feedstock ranges
- Incorporate goal programming into the daily planning work process
- **Provide a complete global optimization framework that includes** stochastic and deterministic elements

Optimal Solution Range Analysis

- Explores the non-linear surface of the optimal planning solution
- Identifies a range of values for key decision variables such as feed stocks and products
- **Provides two indices per variable**
- Minimize / Maximize feedstock based on nominal optimality relaxation (typically 0.1%)

Solution Range Analysis

Flexibility and Utility Indices

Flexibility and Utility index for seven different crudes in a typical US refinery

Solution Range Benefits

- Improved profitability (Trading)
	- Enhanced crude valuation
- **Improved flexibility (Operations)**
	- Scheduling insight
- **Better understanding of the solution space**
- Better risk management capability

Goal Programming

- **Ordered multi-objective optimization**
- **User defined secondary and tertiary objectives**
- **User defined relaxations**
- **Final economic optimization restores Lagrange multipliers**

Global Optimization

- Three component strategy
- Powerful specialized algorithm incorporated into the PIMS Non-Linear Programming (XNLP) technology to avoid local optima often caused by inactive pools
- **A** highly efficient, parallel-processing, statistical multi-start algorithm consistently determines the globally optimal solution
- For certain classes of model non-linearity, a convex relaxation algorithm is able to prove global optimality of the solution found

Stochastic Global Optimization

Convex Relaxation Framework

- Convexification of bi-linear and tri-linear terms
- Additional linear constraints are generated by applying a Reformulation-Linearization-Technique (RLT) technique to selected constraints of the original pooling problem
- **Selective de-activation of RLT constraints**
- Bound reduction algorithm

Deterministic Global Optimization

Key Points and Conclusions

- Created a powerful framework for supporting and augmenting the refinery planning optimization process
- Solution analysis for both feed stocks and products greatly enhances the current processes of crude trading and acquisition
- Global optimization framework using a combination of stochastic and deterministic techniques ensures best solution even for large-size complex problems

Introduction to Blending Operations

- The goal of product blending operations in a refinery is to
	- Meet all the shipments on schedule and on specification
	- Operate within the tank inventory constraints
	- Perform optimally in terms of overall cost and profitability
- Aspen MBO is an event based, multi-period, blending optimization system
	- Rich nonlinear blending property prediction correlations
	- State-of-the-art optimization
	- All major oil companies use it as well as many others
- The proprietary mathematical model optimizes
	- Blend recipe
	- Blend volume
	- Transfers
	- Shipments
	- Receipts

Standard (Inline) Blending

- **Step 1: Schedule a blend**
- Step 2: Components sent simultaneously through a header
- **Step 3: Mix product tank and analyze results**

Problem Definition

- **Rundown blending describes a situation where at least one** component comes directly from a process unit without an intermediate storage tank
- **Problem definition**
	- Traditional blending systems are not designed to optimize blending operations without tanks for all components
	- Most refiners in North America and Europe operate with sufficient intermediate storage
- Business case
	- Many refineries in APAC and Europe (eastern) have operations without component tanks
		- Most customers in those regions use excel with trial-and-error to get a feasible solution
	- Even in NA and Europe, diesel and fuel oil are blended without intermediate tanks
		- Recent strict sulfur regulations make blending a challenge

Rundown Blending

- **Step 1: Schedule a blend**
- **Step 2: Operating unit lined up to product tank** Static components fed simultaneously
- **Step 3: Mix product tank and analyze results**

Rundown Blending Optimization Scope

NE.

áspen)

Rundown Blending Scope

NE.

aspen

Model Structure

- Key features of the new model
	- Mixed-integer nonlinear programming problem (MINLP)
	- Global event-based continuous-time formulation
	- Solution of the model determines
		- **Optimal recipe (static and rundown components)**
		- **Precise start and stop date and time**
		- **Rundown component sequencing for each period**
		- **Split ratio for components with multiple dispositions**
- **New formulation explicitly accounts for blend event** sequencing and start and stop times using binary variables
	- Underlying problem is very large, nonlinear, and involves many discrete decisions
	- Provisional patent granted
- Framework for future expansion to optimal refinery scheduling

Example 1 - Hawaiian Refinery

- Actual customer blending operation for diesel blending
	- Horizon of 16 days
	- 6 blends created for low sulfur diesel
	- 2 rundown components and a static component
	- 3 product tanks
	- 5 fixed shipments
- Use Aspen Refinery Multi-Blend Optimizer (MBO) to model the refinery and optimize the blending operations

Example 1 - Hawaiian Refinery Initial

Example 1 - Hawaiian Refinery Optimal

Example 2 - Gulf Coast Refinery

- **The new model has also extended the mathematical** formulation to handle splitting of the rundown streams to have multiple destinations
- For this example, we have 2 rundown streams split into 3 streams each, where only 1 of those 3 streams has storage
	- 4 total rundown streams used in blending
	- Time horizon is 14 days
	- 15 blend events for 3 products with 16 shipments
- Formulation minimizes the changes in the split ratio over the entire campaign

Example 2 – Gulf Coast Refinery Initial

JF.

aspen

Example 2 – Gulf Coast Refinery Optimal

Key Points and Conclusions

- **The new formulation provides an optimal solution to** complex rundown refinery scenarios - a challenging operational problem
- **The mathematical model is automatically and dynamically** constructed
- **The approach has been validated on several commercial** problems
- The proposed solution
	- Provides for stream containment
	- Ensures that all products meet their specifications
	- Minimizes the use of slop tanks
	- Minimizes operational upsets
	- Minimizes the incidence of product giveaway
	- Maximizes the refinery operating margin

Conclusions and Discussion

- Refinery operations have a long tradition of using state-ofart optimization
- **Refining optimization has come a long way but there are** remaining challenges
- **Deterministic global optimization**
- **Large-scale optimization under uncertainty**
- **Schedule optimization**
- **Refinery-wide real-time optimization**

