Mixed Integer Non-linear Optimization for Crude Oil Scheduling

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Enterprise-wide Optimization Project

Introduction

- Purchasing and Scheduling of crude oil in refineries involves <u>billions of dollars</u> every year
- Simulation tools are used by refinery schedulers to determine feasible schedules
 - □ Interfaced with "Black-box" optimization tools
 - □ Suffer from poor run-times

GOAL:

Derive benefits from use of optimization to obtain crude movement schedules in refinery front-end

Motivation

Economic goals

- Significant reduction in current costs of scheduling and inventory management is practically possible
- <u>Reduced costs</u> in crude-movement scheduling
- * Meet product quality targets and demands more accurately

Technical goals

- * Improve optimization capabilities inside refinery simulators
- *Predicting yields*, crude qualities and tank inventory levels over the whole horizon
- * Integration with *Marine Transportation* models

Problem Outline REFINERY Schematic Crude Tanks Distillation Columns Products Products

Optimization

Given:

- 1. Scheduling Horizon
- 2. Tank inventory (min, max,

initial levels)

- 3. Available crude types and their properties
- 4. Product property specifications and demands
- 5. Bounds on crude and product flows

Determine:

- 1. When to order crudes
- 2. How much of each crude to order
- 3. Operating flows of crude between tanks
- 4. Charges to Pipestills
- 5. How much of each product to produce

Simulation

Schedulers consider multiple scenarios and perform simulations





Model Development

 <u>Rigorous multiperiod formulation</u> to model the important aspects of scheduling (*Discretized time domain*)



Model Constraints



Inventory balances

$$I_{b,t-1}^{tot} + \sum_{a \in A_b} V_{abt}^{tot} = I_{bt}^{tot} + \sum_{c \in C_b} V_{bct}^{tot} \quad \forall t \in T, \forall b \in B$$
$$I_{jb,t-1}^{comp} + \sum_{a \in A_b} V_{jabt} = I_{jbt}^{comp} + \sum_{c \in C_b} V_{jbct} \quad \forall j \in J, \forall t \in T, \forall b \in B$$

$$I_{jbt}^{comp} = f_{jbt} I_{bt}^{tot} \forall j \in J, \forall t \in T, \forall b \in B$$

Non-linear equations
containing Bilinearities

u Inflow and Outflow restrictions

 $w_{abt} + w_{bct} \le 1 \quad \forall a \in A_b, \forall c \in C_b, \forall b \in B, \forall t \in T$

– Flow balances

Model Constraints (Contd ...)

Crude Arrival Constraints

Crude unloading contraints

- ^o <u>Only one tank</u> can be fed by <u>one vessel</u> at any instance of time
- ^o <u>Only one crude type</u> can be unloaded at a point in time

Distillation Column Constraints



 Inlet flow and property restrictions for Distillation Column

D Allocation constraints

• <u>At most two charging tanks</u> can charge a pipestill at any time

- Yield relations to calculate product flows and properties
- Product flow and property balances
- Product boiling points calculations

Optimization Model

Scheduling problem modeled as a

Multiperiod Mixed Integer Non-linear Program (MINLP)

- Discrete variables used to determine which flows should exist and when
- □ Model is highly <u>non-linear</u> and <u>non-convex</u>

Problem Parameters

- □ 11-day horizon, 7 Crude Tanks, 41 Crudes
- **2** Pipestills, **13** Cuts, **6** Final Products

Model statistics

	Base case	Extreme case
# Time periods	3	20
# Properties	3	8
# Binary variables	315	2,100
# Continuous variables	15,522	106,560
# Constraints	18,153	124,540

* Need <u>specialized algorithm</u> to solve such large-scale models

Solution Technique

<u>Bi-level heuristic algorithm proposed</u> to solve MINLP formulation



Alternate Solution Method

<u>Rolling Horizon Approach</u> to temporally decompose model and solve it

Solve single time period and fix inventories in tanks for next time period



- **Easier to solve** temporally decomposed sub-problems
- □ Results of this method are similar to full space optimization
- **Lower likelihood** of finding solutions

General Remarks

- Locally optimal schedules obtained for problems where 3 properties are targeted in the products over the entire scheduling horizon
- Cost objective problems easier to solve than problems with product property deviation objectives
- ✤ Solution times vary between <u>500 CPU secs 5000 CPU secs</u>
- * Scaling of data and using good starting points in the optimization are very important \rightarrow lead to solutions in much reduced solution times

Conclusions

Summary of work

- Development of a rigorous <u>Mixed Integer Non-linear</u> formulation to model crude oil movement in a refinery "front-end"
- Heuristic algorithms to optimize non-linear model to get (locally) optimal schedule
- □ Method to obtain a <u>bound on the best possible cost</u> objective

Potential future work

- □ <u>Target more properties</u> in product specifications
- Enhance existing model by including downstream processing and crude vessel routing
- Robust non-linear solver development for solving large-scale refinery scheduling MINLPs

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