

# Price Integration in Competitive Markets with Capacitated Transportation Networks

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# Are Gasoline Prices Integrated?

Prices are integrated if they respond to the same price shocks

- Typically answered using co-integration tests.
  - Does  $p_1 = Ap_2 + B$ ?
- NA Fuel Markets “fairly” Co-integrated

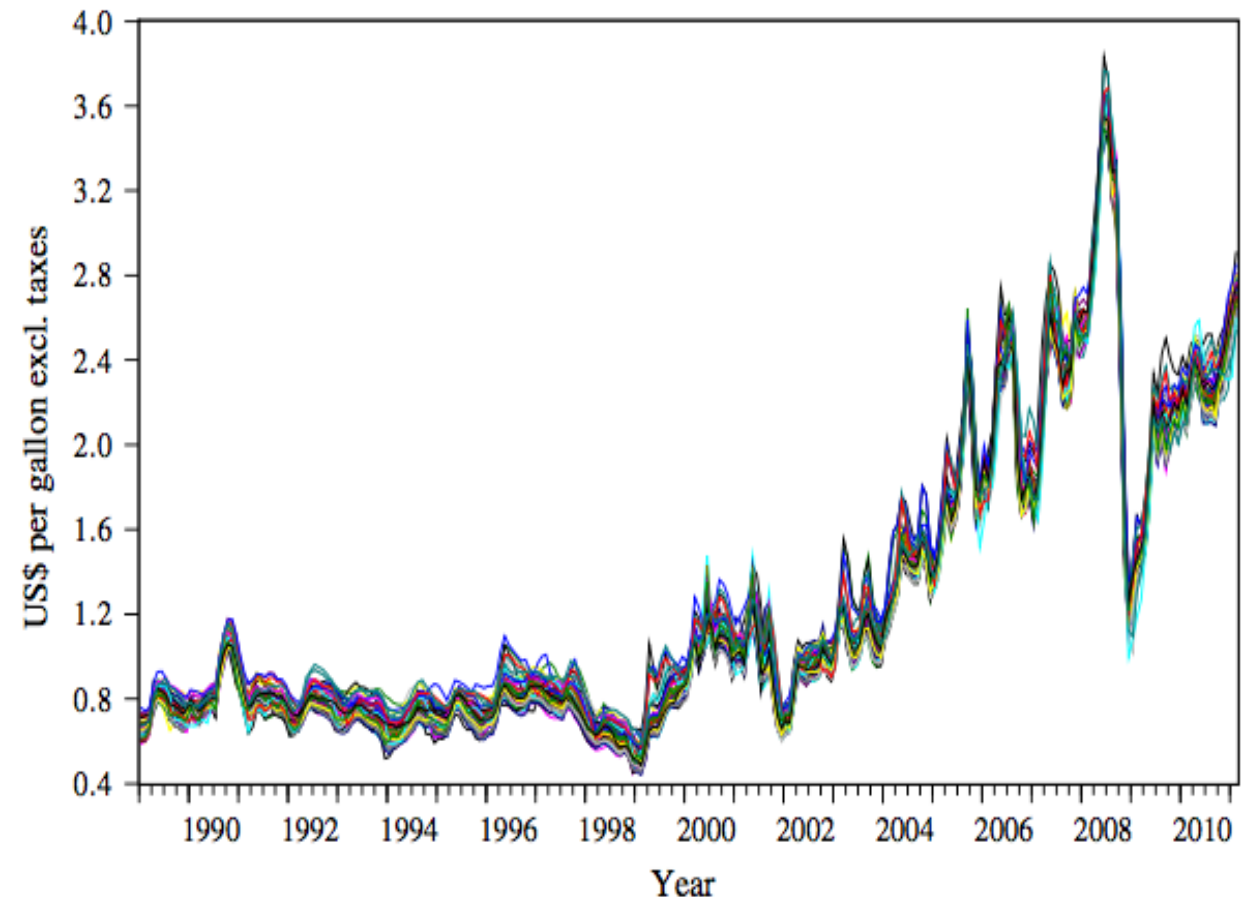
Holmes et al. (2013), Paul et al. 2001

But do co-integration tests capture the whole picture?

Statistical co-integration tests

- Aggregate tests of stationarity over long periods of time
- Based on pair-wise tests
- **Not able to detect transient dis-integration**
- **Not able to detect submarkets**  
(McNew Fackler 97)

*M.J. Holmes et al. / Energy Economics 36 (2013) 503–510*



# Co-integration and the Law of One Price

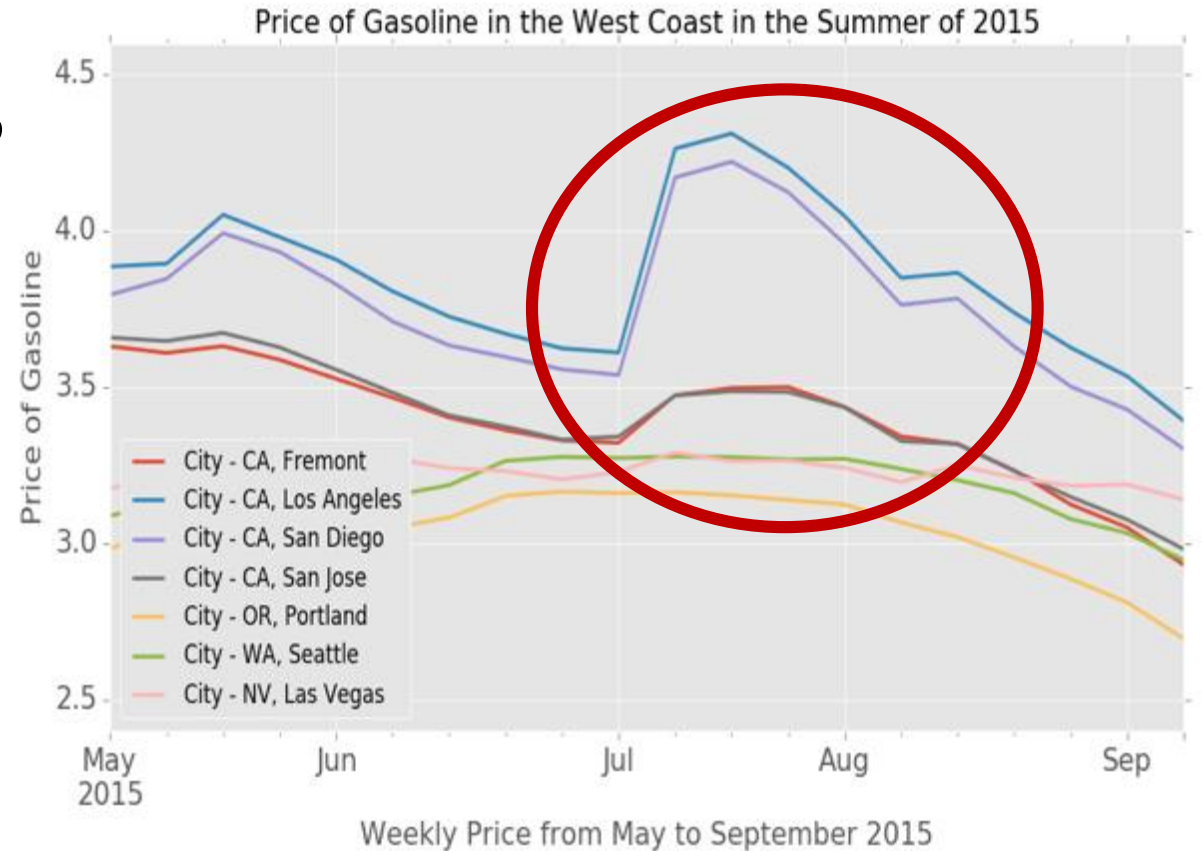
Zooming in: **transient failures of LOOP**

Arbitrage would fix this if:

- Same product
- Transparent pricing
- **Cheap and fast transportation**

Hypothesis:

- **Transient sub-markets can arise due to capacitated transportation**



# Research questions

## Theoretical:

- What are the implications of capacitated transportation to price integration?

## Empirical:

- Can these results motivate algorithms to identify highly integrated sub-markets?
- Are transient submarkets identifiable in the North American gasoline market?

# Aside: Overview of approach

1. Concerned with markets where there is spatial variation in prices
  - Can we use pricing data to isolate the effect of capacity constraints on spatial variation?
2. Assume an optimal form to the market mechanism
  - Electricity markets: market facilitator (ISO) finds the “welfare maximizing” allocation
  - Gasoline (and other) markets

Assumption of competitive markets: equilibrium == solution of optimal allocation problem
3. Determine data assumptions
  - Treat prices and allocation as partial dual and primal solutions
  - Allow demand and supply to vary over time
  - Fix the transportation network
4. Use optimality conditions to make inferences
  - Inverse optimization approach (Birge, Hortaçsu, Pavlin 2017)

# Model: A Spatially Separated Market

Consider an idealized model of a spatially separated market for a single commodity

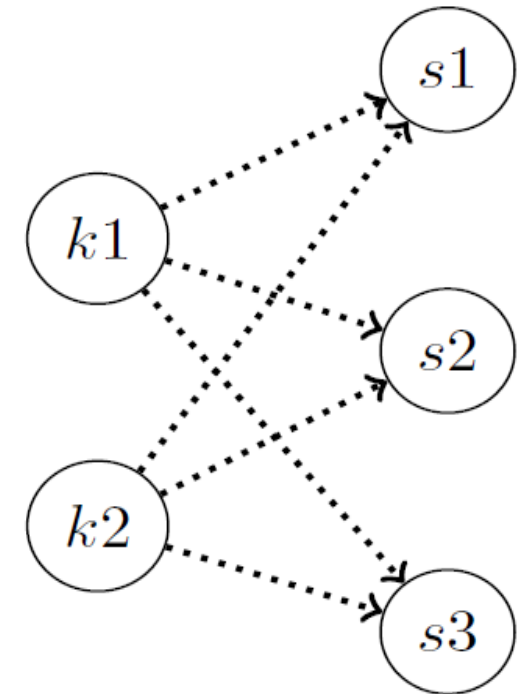
Uncapacitated markets: Samuelson 52, Takayama and Judge 64, McNew and Fackler 97

Capacitated energy market models Cremer et al. 03, Secomandi 2010, Gabriel et al. 05

- A market consists of a set of **production nodes**  $K$  and **demand nodes**  $S$ 
  - Production node  $k \in K$  produces  $b_k$  units at cost  $W_k(b_k)$
  - Demand node  $s \in S$  consumes  $b_s$  units receiving welfare  $W_s(b_s)$
- Nodes are connected by a set of **capacitated transportation links**  $E$ 
  - Flow over  $(i, j) \in E$  is  $f_{ij} \leq u_{ij}$  at cost  $f_{ij}c_{ij}$

Market dynamics:

- **Demand/supply** functions may **change** every period
- **Network** structure remains **stable**
- At each period **prices and allocation** correspond to a **spatial price equilibrium**



# Single Period: Spatial Price Equilibrium Model

SPE extends a competitive pricing equilibrium

- With diminishing marginal returns i.e. concave  $W_s(b_s)$  convex  $W_k(b_k)$
- **Allocation** optimizes the welfare maximization problem

$$\max_{f,b} \sum_s W_s(b_s) - \sum_k W_k(b_k) - \sum_{(i,j)} c_{ij} f_{ij}$$

S.T.

$$\sum_{(i,s)} f_{is} - b_s - \sum_{(s,j)} f_{sj} = 0 \quad \forall s \in S$$

$$\sum_{(i,k)} f_{ik} + b_k - \sum_{(k,j)} f_{k,j} = 0 \quad \forall k \in K$$

$$0 \leq f_{ij} \leq u_{ij}$$

$$b \geq 0$$

- Commodity **prices**  $\lambda_s, \lambda_k$  are locational and correspond to dual variables of flow balance constraints<sub>7</sub>

# Optimality/Equilibrium Conditions

## Nodal prices

- $\lambda_s, \lambda_k$

$v_{ij}$  = shadow price of flow u.b.

## Price differences

Driven by transportation costs  
and a surcharge for congestion

$$\begin{aligned}\lambda_s &= W'_s(b_s) + \alpha_s & \forall s \in S \\ \lambda_k &= W'_k(b_k) - \alpha_k & \forall k \in K \\ \lambda_j - \lambda_i &= c_{ij} - w_{ij} + v_{ij} & \forall (i, j) \in E \\ \alpha_n &\geq 0 & \forall n \in N \\ w_{ij}, v_{ij} &\geq 0 & \forall (i, j) \in E\end{aligned}$$



# Analysis with costly transportation: The Neutral Band

**A Neutral band** is a range within which the **difference** between a pair of prices may deviate as a result of changes in demand and supply

- Characterized for neighbouring market locations resulting from transaction costs (e.g. Goodwin, Piggot 2001)
- Small range of neutral band  $\cong$  higher price integration

## **Theorem** – *network neutral band*

In the absence of active capacity constraints, the **network neutral band** between a pair of consumers  $s, r$  operating in a costly transportation network is defined by:

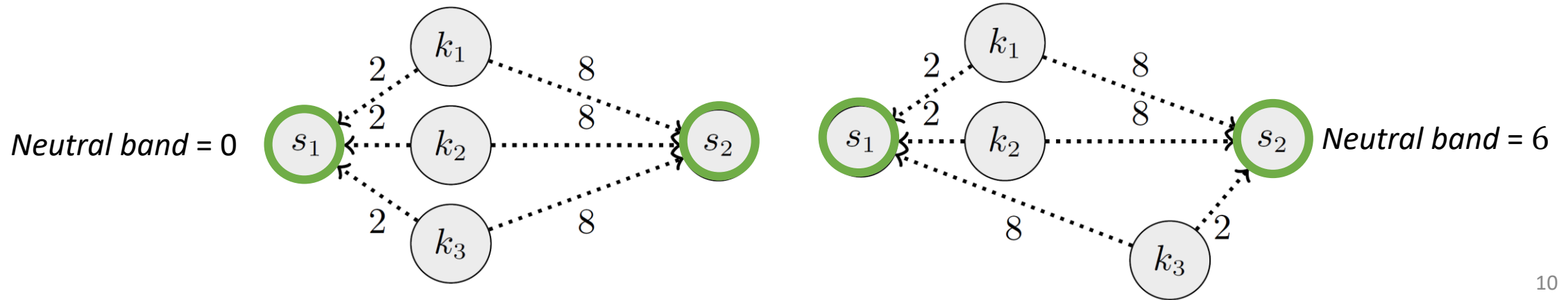
$$\min\{p_{ks}^* - p_{kr}^* | k \in K\} \leq \lambda_s - \lambda_r \leq \max\{p_{ks}^* - p_{kr}^* | k \in K\}$$

Where  $p_{ks}^*$  is the shortest path between supplier  $k$  and consumer  $s$

# Analysis with Costly Transportation: Example of Network Neutral Bands

## Network neutral band

$$\min\{p_{ks}^* - p_{kr}^* | k \in K\} \leq \lambda_s - \lambda_r \leq \max\{p_{ks}^* - p_{kr}^* | k \in K\}$$



# Analysis with Costly Transportation and Capacity Constraints: The congestion surcharge

**The congestion surcharge**  $w_s$  is the amount by which the equilibrium cost exceeds the uncapacitated delivery cost

$$w_s = \max\{\lambda_s - \lambda_k - p_{ks}^* | k \in K\}$$

**Theorem** – *neutral band shifted by congestion surcharge*

The price difference between a pair of consumers  $r$  and  $s$  is bounded by:

$$\min\{p_{ks}^* - p_{kr}^* | k \in K\} + w_s - w_r \leq \lambda_s - \lambda_r \leq \max\{p_{ks}^* - p_{kr}^* | k \in K\} + w_s - w_r$$

Where  $p_{ks}^*$  is the shortest path between supplier  $k$  and consumer  $s$

# Analysis with Costly Transportation and Capacity Constraints: A bottleneck link may not evenly affect all downstream consumers

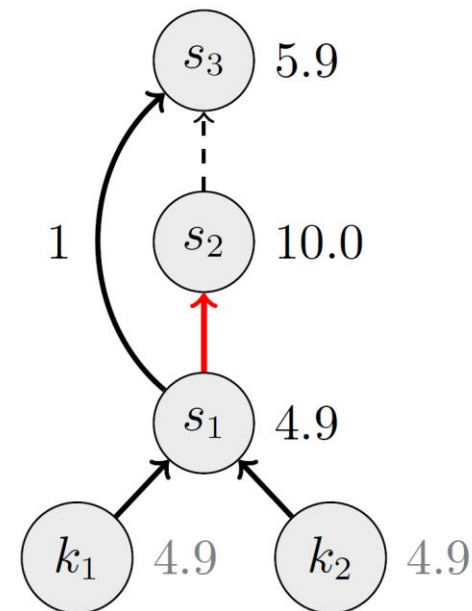
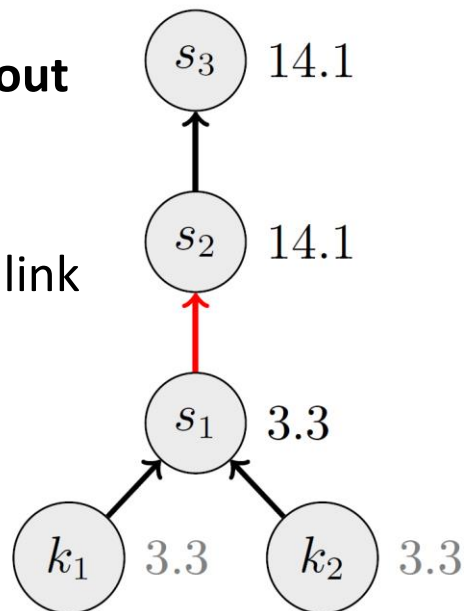
Congestion surcharge  $w_s$  from a single congested link  $e$

- $w_s$  is less than or equal to the shadow price  $v_e$
- $w_s \in [\min\{v_e, \min(\delta_{ks}(e))\}, \min\{v_e, \max(\delta_{ks}(e))\}]$ 
  - Rerouting cost  $\delta_{ks}(e)$  is the additional cost of routing from  $k$  to  $s$  without link  $e$

**Equilibrium prices with and without an additional link**

Red link is at capacity

Links are costless except the new link



# Analysis with Costly Transportation and Capacity Constraints: Price Decomposition

## Theorem

Assume stable market structure, set of sample market realizations  $T$ , price  $\lambda_n^t$  of node  $n$  may be decomposed as follows:

$$\lambda_n^t = \eta^t + \rho_n + \epsilon_n^t + w_n^t,$$

- Supply cost **in green**
  - $\eta^t$  — node invariant “market price”
  - $\rho_n$  — time invariant “transportation cost”
- Neutral band variation **in blue**
  - $\epsilon_n^t \in [-h, h]$  — variation within the uncongested neutral band
- Congestion surcharge **in red**

# Analysis of Equilibrium With Capacity Constraints: Implications of Price Decomposition

## Implications for Co-Integration

Without congestion:

- Prices are integrated if  $h = 0$ , otherwise they may vary within a neutral band

With congestion:

- Variation in congestion surcharge may decrease co-integration

## Implication for Market Structure

Congestion-Induced Submarkets

- A Congestion-Induced Submarket  $S_e$  is a set of nodes with a non-zero congestion component for a particular link  $e$

## Implications for Identification of Price Components

- Difficult to disentangle  $\epsilon_n^t$  and  $w_n^t$
- Overcome by assuming  $w_n^t = 0$  most of the time

# Bringing it to data:

## Surcharge Estimation Model (SEM)

- Assume stable market structure  $(N, E, c, u)$  over samples  $T$
- Demand and supply vary  $(W_k, W_s)$

- *Utilize price decomposition*
- Search for congestion surcharge:
- Minimizing neutral band
- Implemented as MIP
- Parameters:
  - Number of submarkets
  - Maximum time submarkets are active (parameter  $\beta$ )

$$\begin{aligned} & \underset{\eta^t, \rho_s, \epsilon_s^t, w_s^t, \alpha_s}{\text{minimize}} && \sum_{s \in \mathcal{S}} \alpha_s \\ & \text{subject to} && \lambda_s^t = \eta^t + \rho_s + \epsilon_s^t + w_s^t, \quad \forall s \in \mathcal{S}, t \in \mathcal{T}, \\ & && |\epsilon_s^t| \leq \alpha_s, \quad \forall s \in \mathcal{S}, t \in \mathcal{T}, \\ & && w_s^t \in \mathcal{W}(\theta)^+, \quad \forall s \in \mathcal{S}, t \in \mathcal{T}. \end{aligned}$$

# Bringing it to data: Colonial Pipeline Disruption Case Study

- Colonial pipeline serves gasoline from Gulf to Eastern US
- Disruption on September 9<sup>th</sup> 2016
- Data
  - Jan-Dec 2016
  - Average daily prices at 18 locations

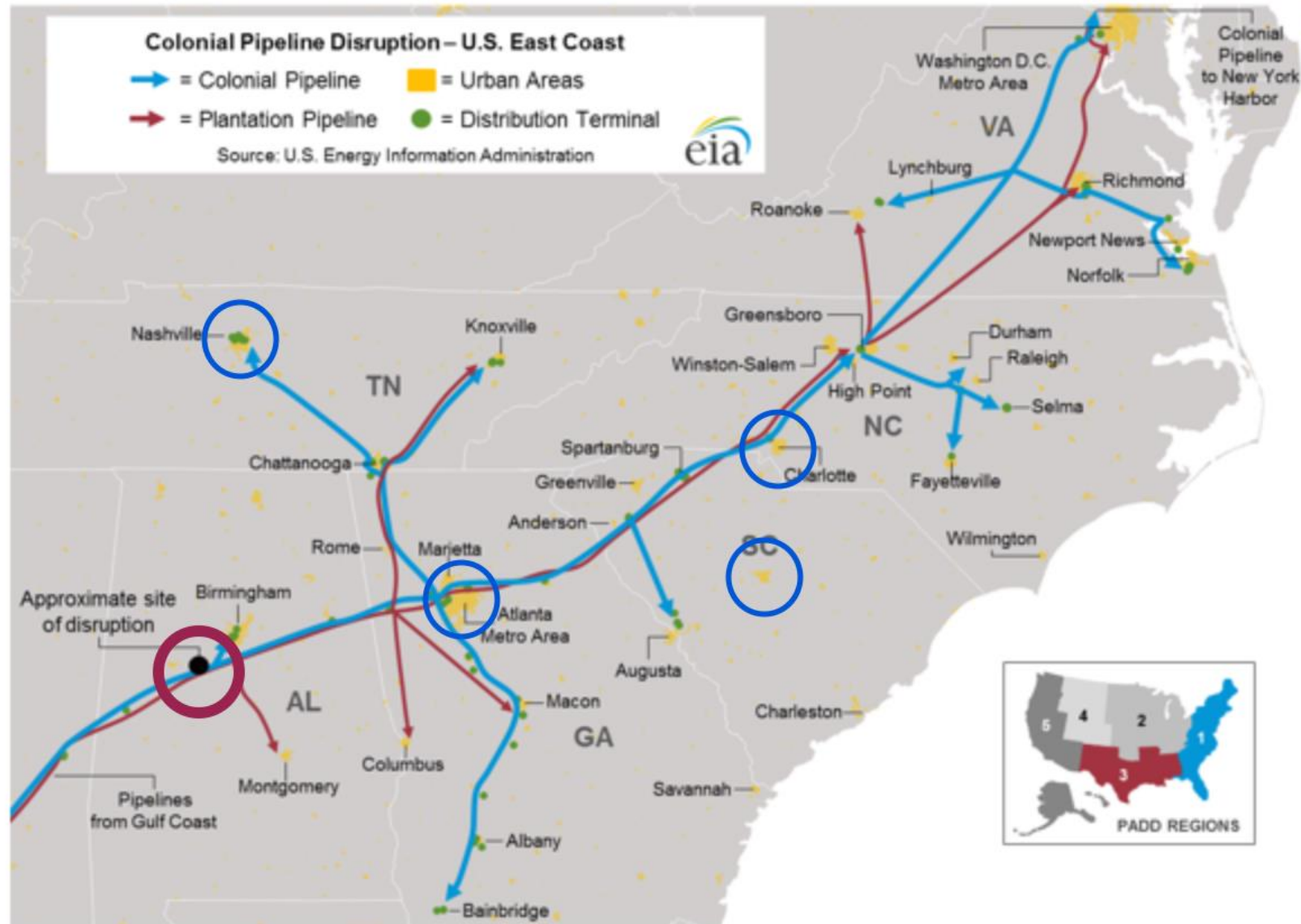


## Colonial Pipeline Fire Severs Fuel Artery in Alabama

A pipeline that supplies much of the East Coast with gasoline, diesel and jet fuel caught fire Monday after work crew equipment struck the line. Photo: Facebook/April Ruth Everett



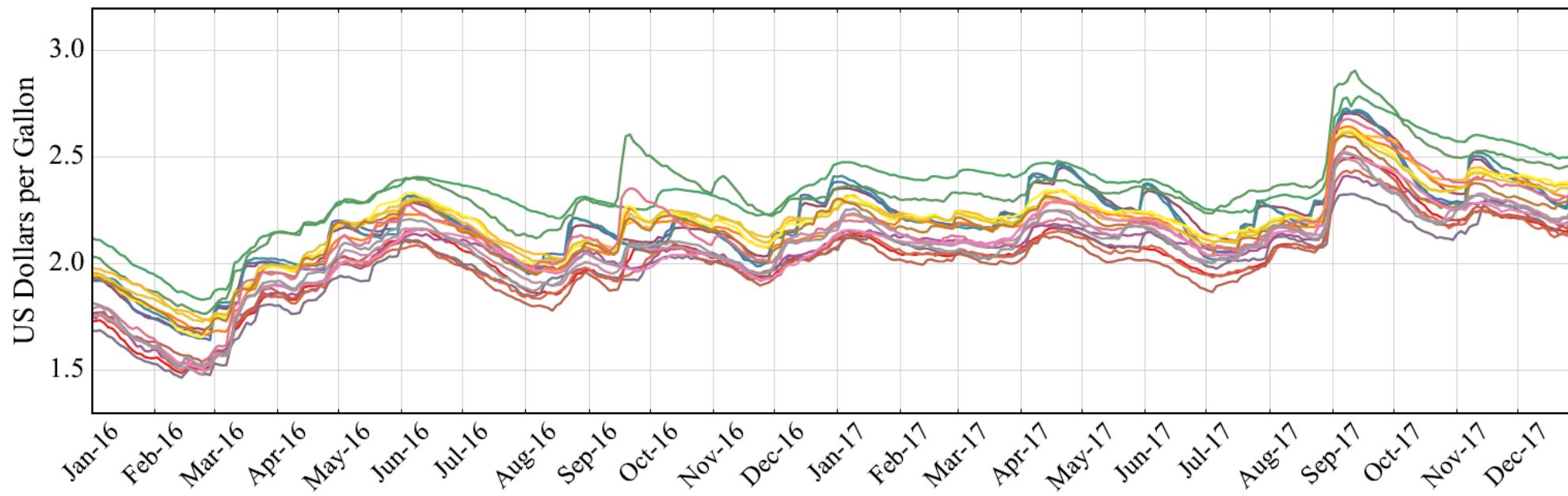
# Colonial Pipeline Disruption Case Study



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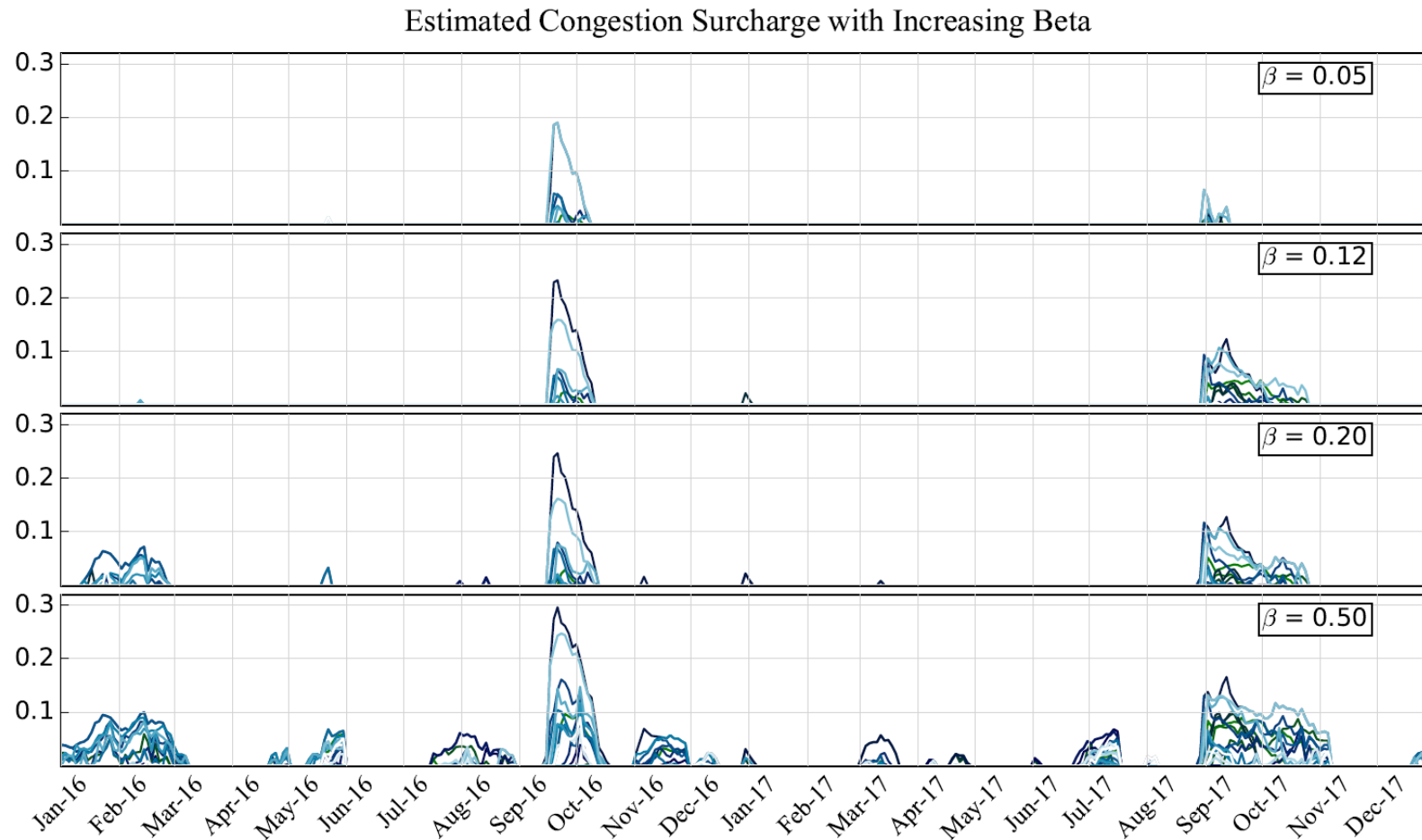
## Prices in market footprint

- Retail gasoline prices at 20 cities in Southeastern US



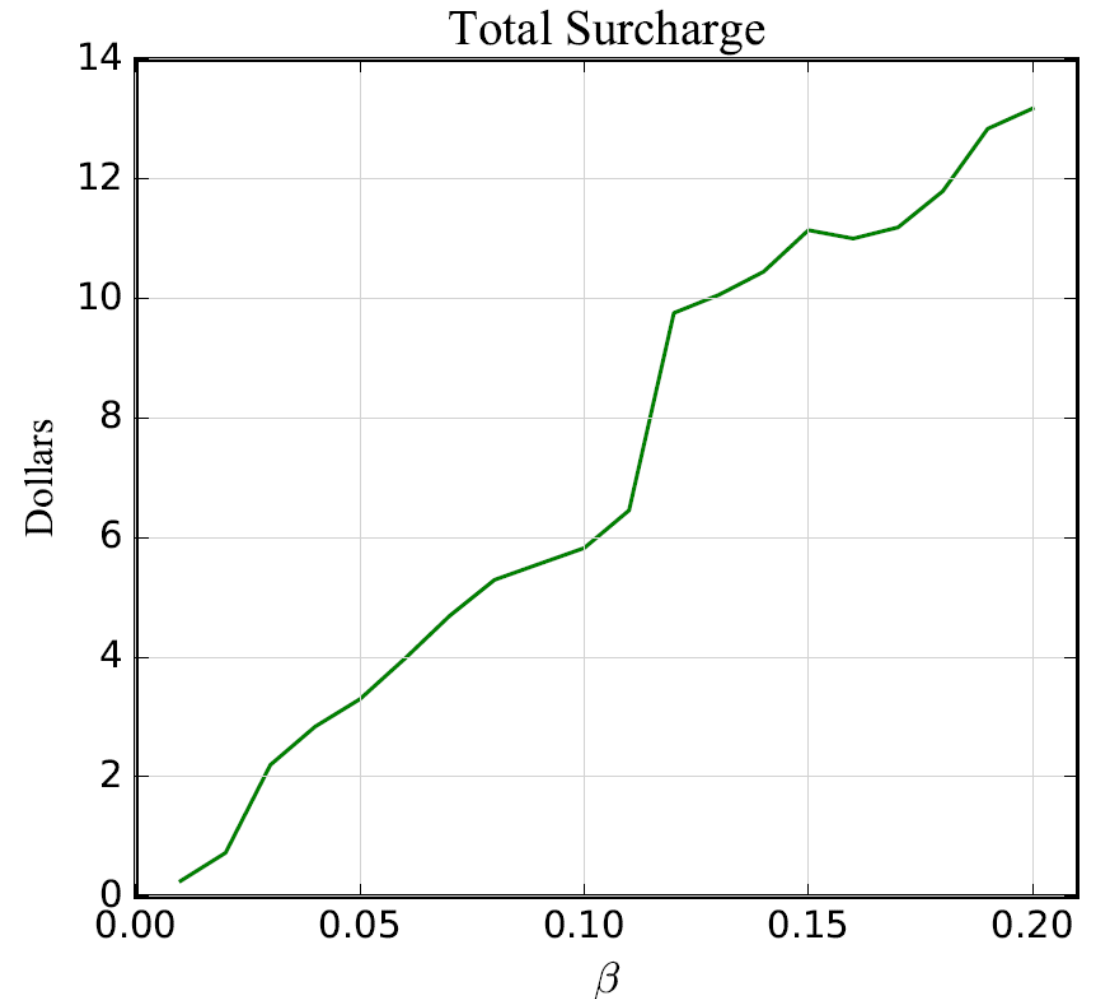
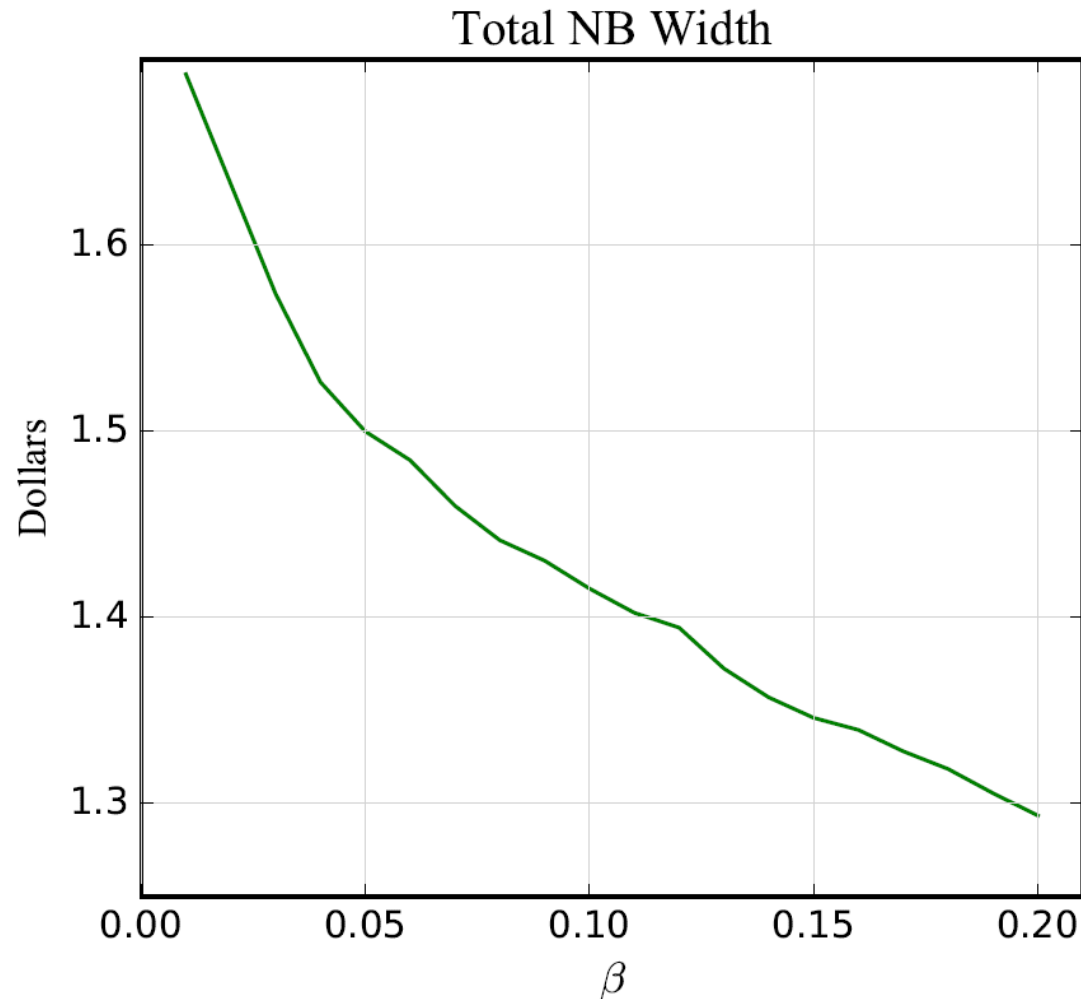
# Colonial Pipeline Disruption Case Study:

## Consistent estimates of periods with high surcharge



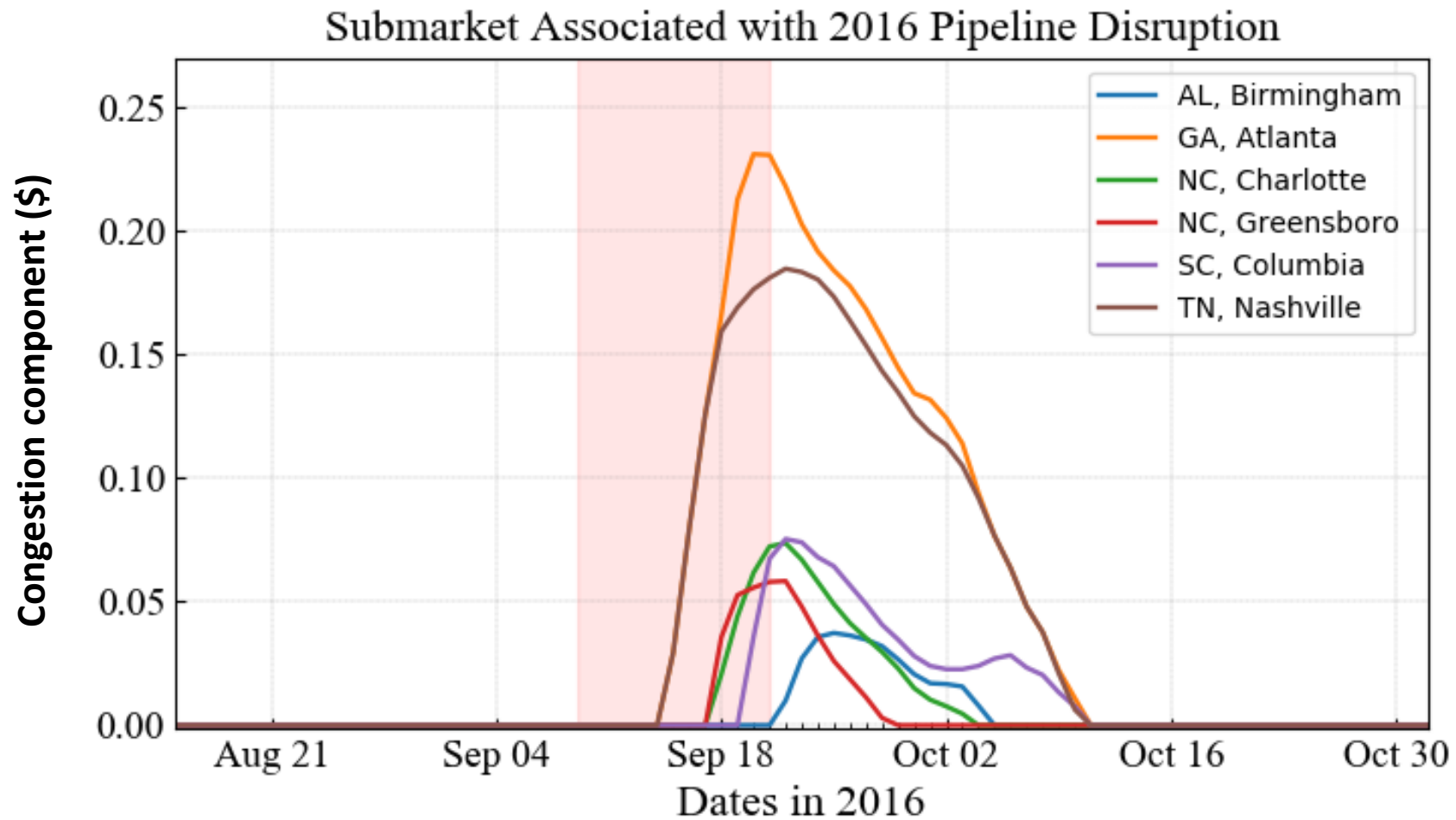
# Colonial Pipeline Disruption Case Study:

## Difficult at present to determine “correct” $\beta$



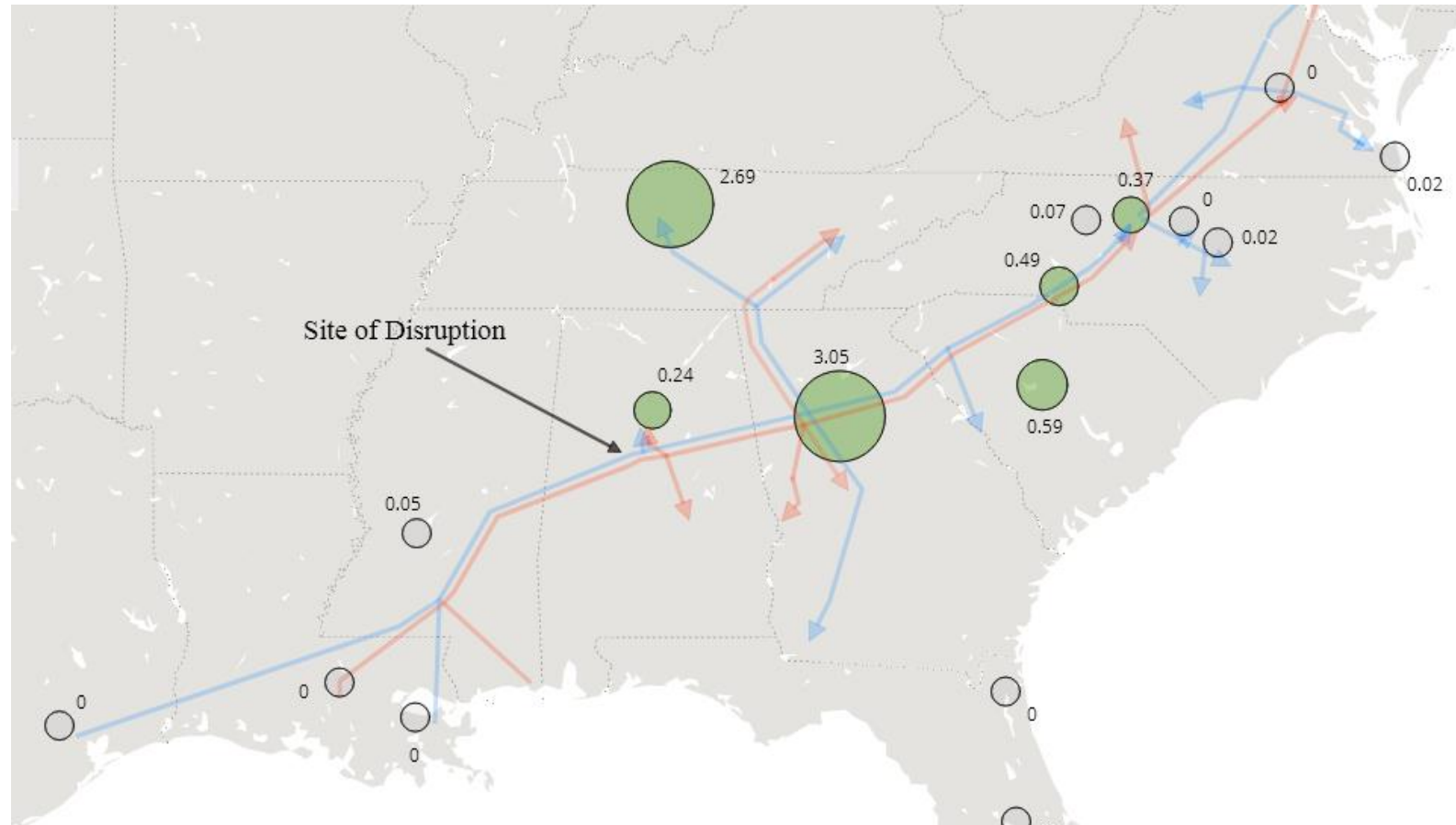
# Colonial Pipeline Disruption Case Study

## Congestion component of submarket



# Colonial Pipeline Disruption Case Study

## Pipeline map and discovered submarket



Green nodes are in the colonial disruption submarket

Size of node = sum of daily congestion components

Colonial pipeline in blue

Plantation pipeline in red

# Conclusions

- Capacitated spatial price equilibrium model of energy markets
  - Characterize neutral band and congestion surcharge on network
  - Price decomposition
- Implications for market price integration
  - $\epsilon_n^t$ : Neutral band bounded by transportation network structure
  - $w_s$ : Congestion can lead to transient separation of regional prices
- SEM Algorithm:
  - Identifying nodes consistent with pipeline disruption
  - Identifying magnitude of congestion related price changes
- Ongoing:
  - Identify appropriate parameter levels to differentiate between congestion and neutral band variation
  - perform larger scale empirical investigations