

Fun of Modeling Electricity Markets using Equilibrium Models

Yihsu Chen, Ph.D.

**School of Engineering
University of California, Santa Cruz**

April, 2016



Menu

- **All about me**
- **Background of electricity markets**
- **Basic models**
- **Results & discussions**
- **Other research**

All About Me: Yihsu Chen

B.S., Environmental Science and Engineering, Tunghai University, Taiwan, 1991-1995

M.S., Harvard University, School of Public Health, 1997-1999

Ph.D., Environmental Engineering, The Johns Hopkins University, 2000-2006

Assistant Professor, School of Social Sciences, Humanities and Arts, School of Engineering, University of California, Merced, 2006-2012

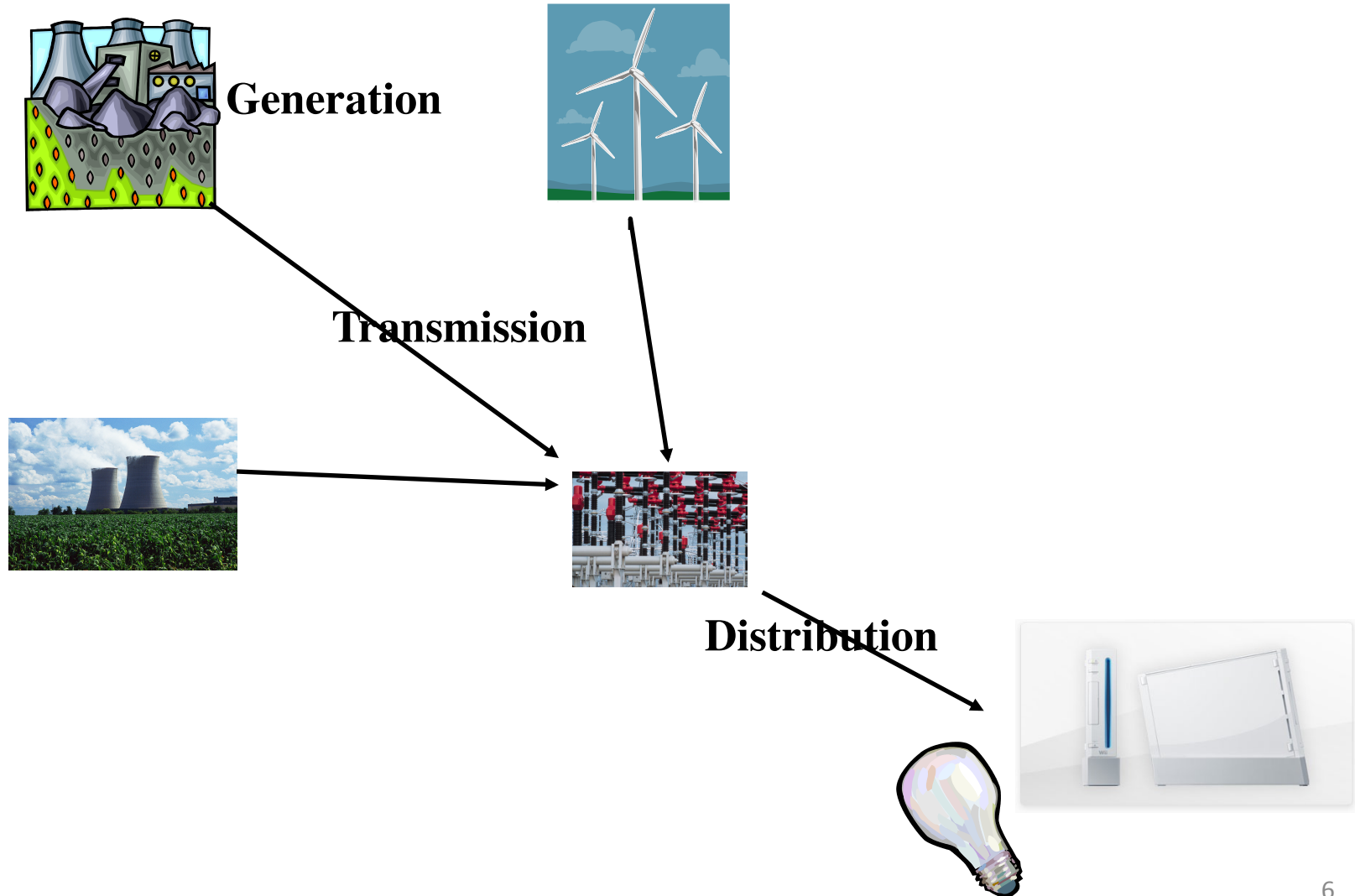
Associate Professor, School of Social Sciences, Humanities and Arts, School of Engineering, University of California, Merced, 2012-2015

Associate Professor, Technology Management, Baskin School of Engineering, University of California, Santa Cruz, 2015–

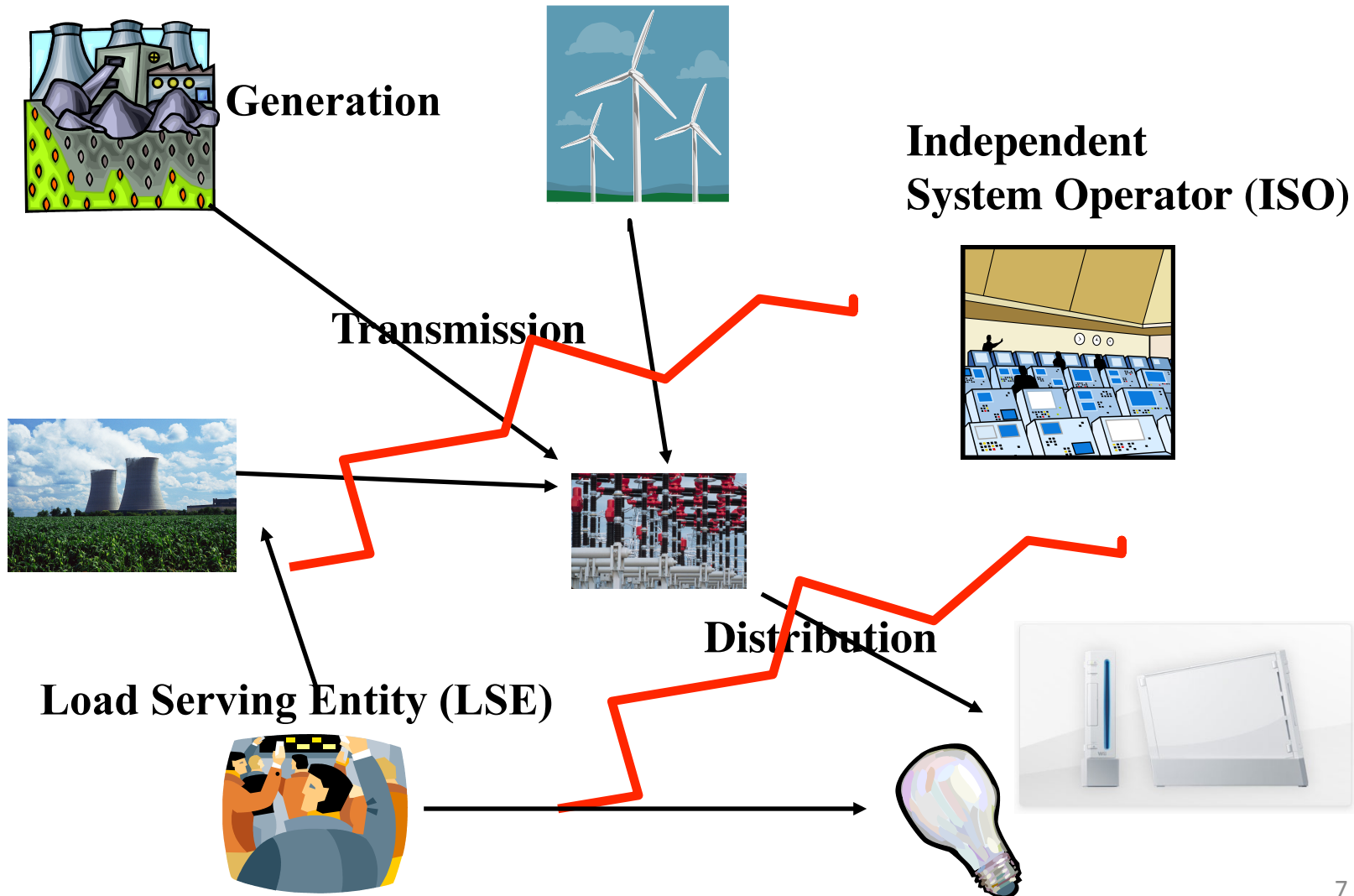
Background: Electricity vs Ice Cream

Electricity	Ice Cream
Homogeneous good (except green power)	Differential product (e.g., flavors: peach, strawberry, etc)
Real-time balance	Inventory (refrigerator, storage, etc)
No real-time pricing	You know prices at spot
Low short-run demand elasticity	High elasticity due to product substitution
Power follows Kirchhoff's laws	Transported by trucks and follows lights
Ancillary services (e.g., reactive power, spinning reserves)	Only need spoon and refrigerator

Background: Vertical Integrated Power Sector



Background: Restructure or Deregulation



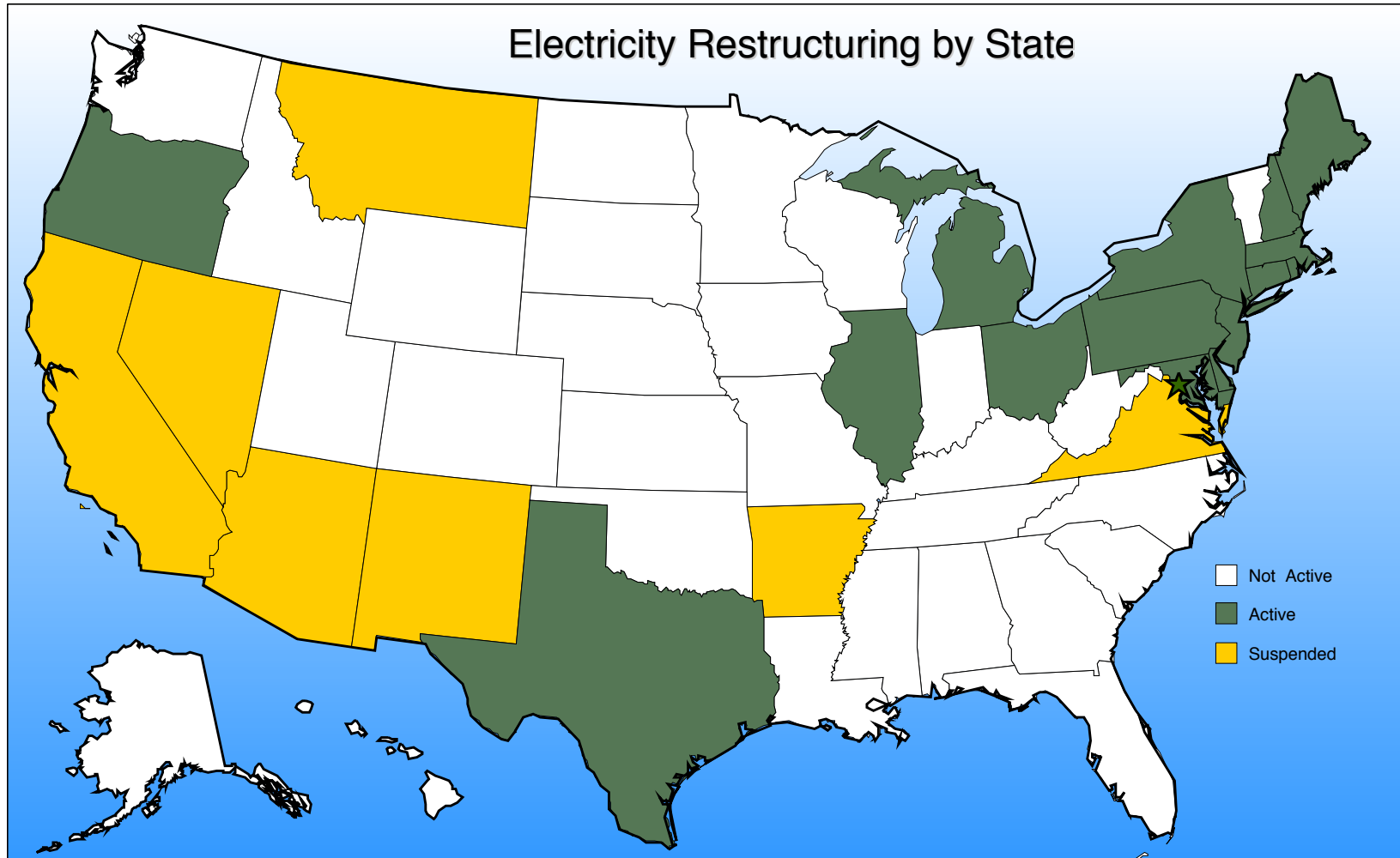
Background: Why Deregulation?

- **Provide correct signal for long-term investment in generation capacity and transmission**
- **Improve economic efficiency**
 - **Productive efficiency: production frontier (lower costs), innovation**
 - **Allocative efficiency: quantities and prices**

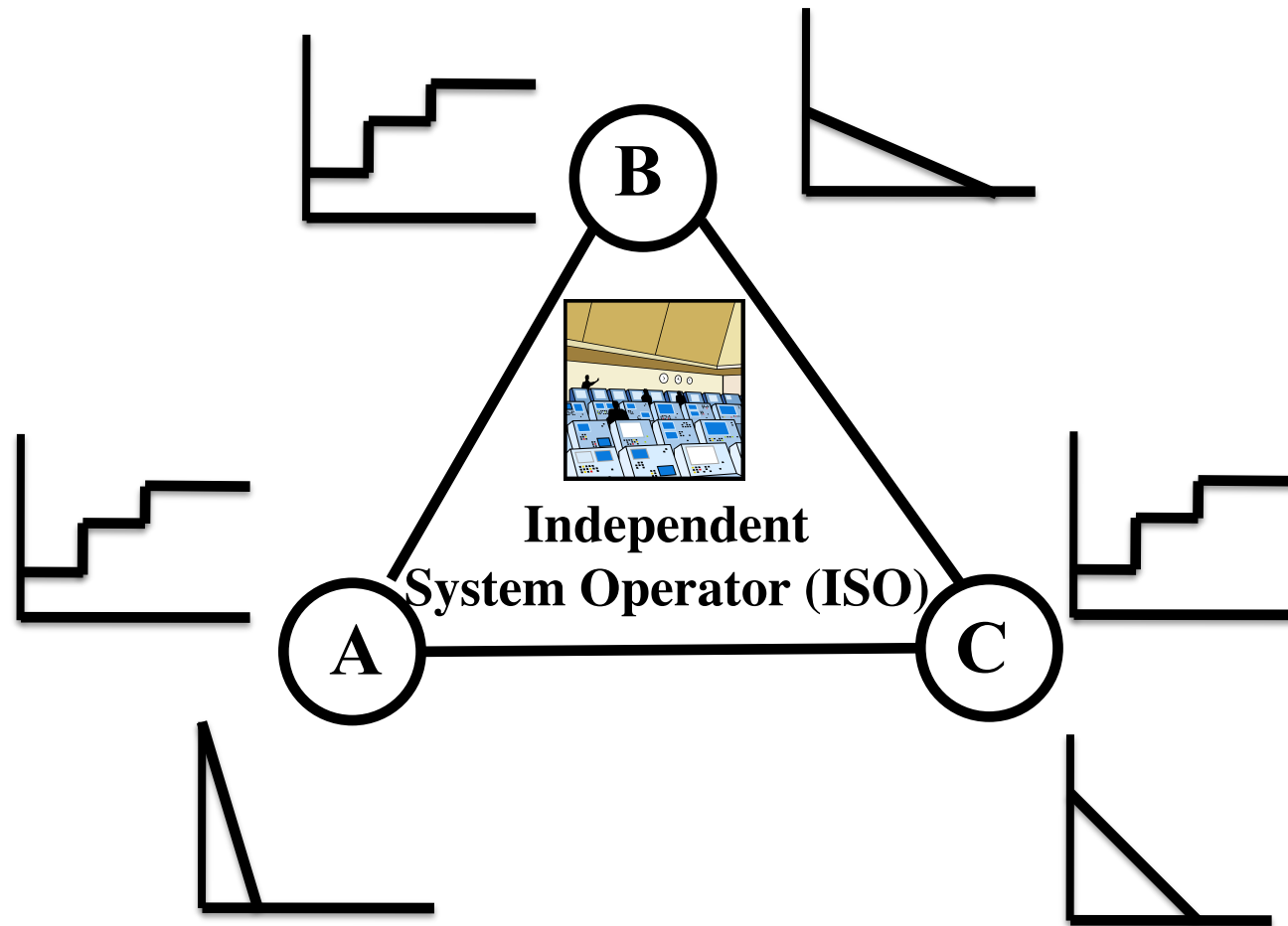
Contemporary Federal and State leadership

- **FERC Order 888, 889 (1996) and 2000 (Regional Transmission Organization, RTO) and Standard Market Design (2002), etc.**
- **CA and MA lead in 1995**
 - **most States were following until California 2000-2001 crisis**

Background: A snapshot



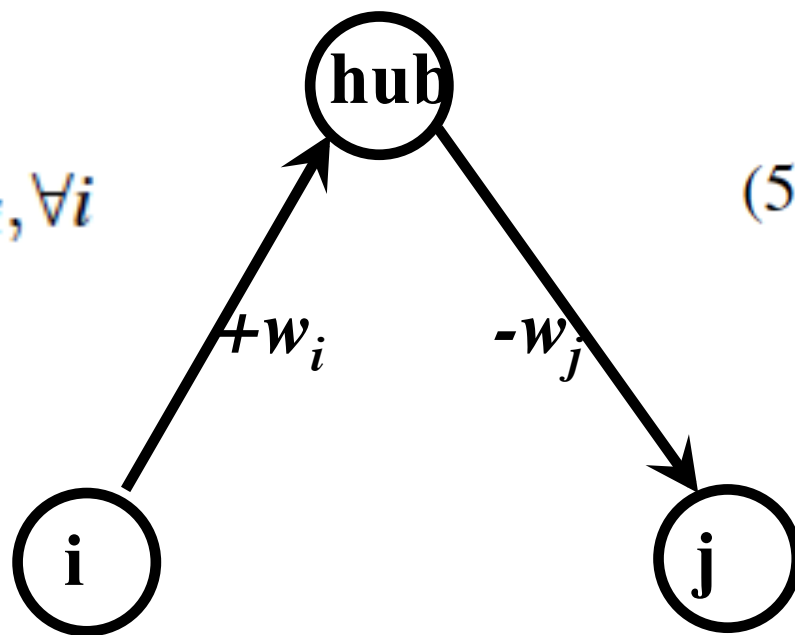
Background: Electricity Markets



Models: Producers & Market Clearing Conditions

$$\begin{aligned}
 & \underset{x_{fih}, s_{fi}}{\text{maximize}} \sum_j (p_j^E - w_j) s_{fj} - \sum_{i, h \in H(i, f)} (C_{fih} - w_i) x_{fih} \\
 & \text{subject to } x_{fih} \leq X_{fih}, (\rho_{fih}) \\
 & \quad \sum_j s_{fj} = \sum_{j, h \in H(j, f)} x_{fjh}, (\theta_f) \\
 & \quad x_{fih}, s_{fj} \geq 0.
 \end{aligned} \tag{4}$$

$$y_i = \sum_{f, h \in H(i, f)} x_{fih} - \sum_f s_{fi} - a_i, \forall i \tag{5}$$

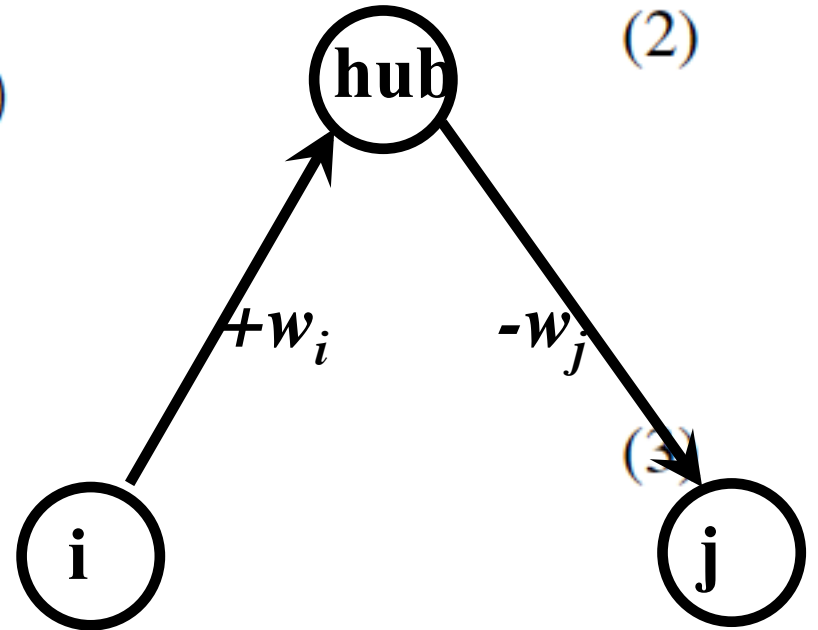


Models: Consumers & ISO & Arbitrager

$$p_j^E = P_j^0 - \frac{P_j^0}{Q_j^0} (\sum_f s_{fj} + a_j), \forall j \quad (1)$$

$$\begin{aligned} & \underset{y_i}{\text{maximize}} \sum_i w_i y_i \\ & \text{subject to } \sum_i PTDF_{ki} y_i \leq T_k, (\lambda_k) \end{aligned} \quad (2)$$

$$\begin{aligned} & \underset{a_i}{\text{maximize}} \sum_i (p_i^E - w_i) a_i \\ & \text{subject to } \sum_i a_i = 0 \end{aligned}$$



Solving a duopoly (2 firms) game

Consider two firms with constant marginal cost c facing a linear demand function: $p(Q)=a-Q$, where $Q=q_i + q_j$. For firm i , we have

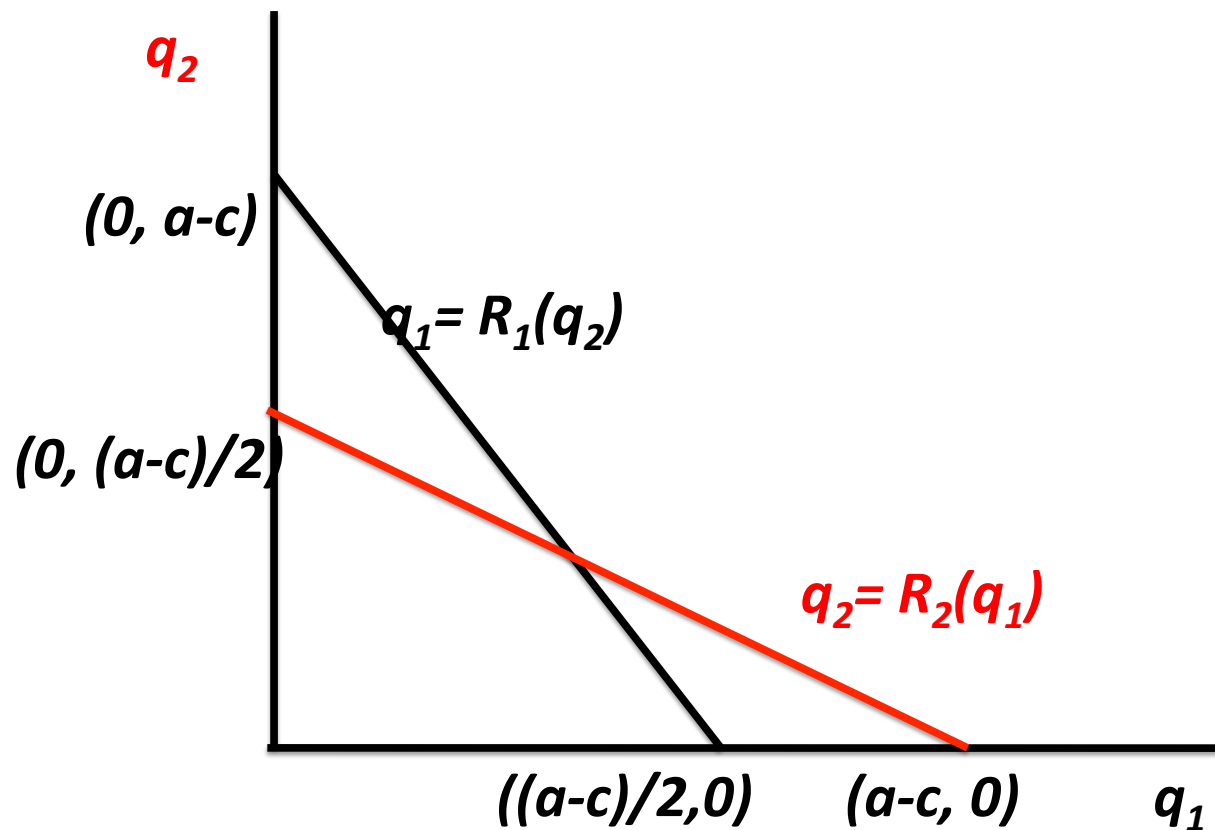
$$\text{Max } \pi_i(q_i, q_j) = p_i(Q)q_i - q_i \times c_i.$$

A Nash Equilibrium is defined by a pair of strategy (q_i^*, q_j^*) such that for each i , we have $\pi_i(q_i^*, q_j^*) \geq \pi_i(q_i, q_j^*)$

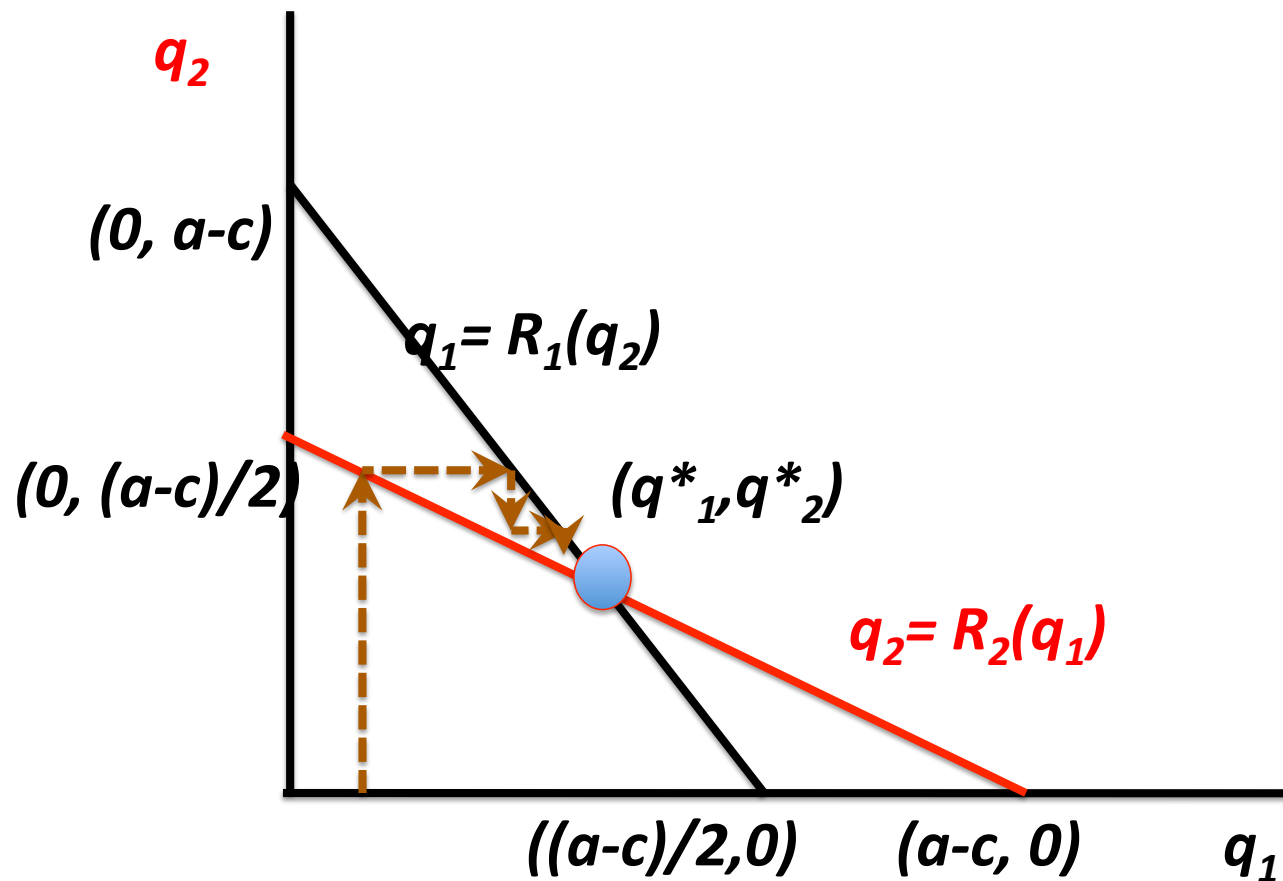
To find each i 's optimal strategy, we solve for $\frac{\partial \pi_i(q_i^*, q_j^*)}{\partial q_i} = 0$

$$q_i(q_j) = \frac{1}{2}(a - q_j - c) \text{ and } q_j(q_i) = \frac{1}{2}(a - q_i - c)$$

Solving a duopoly (2 firms) game



Solving a duopoly (2 firms) game



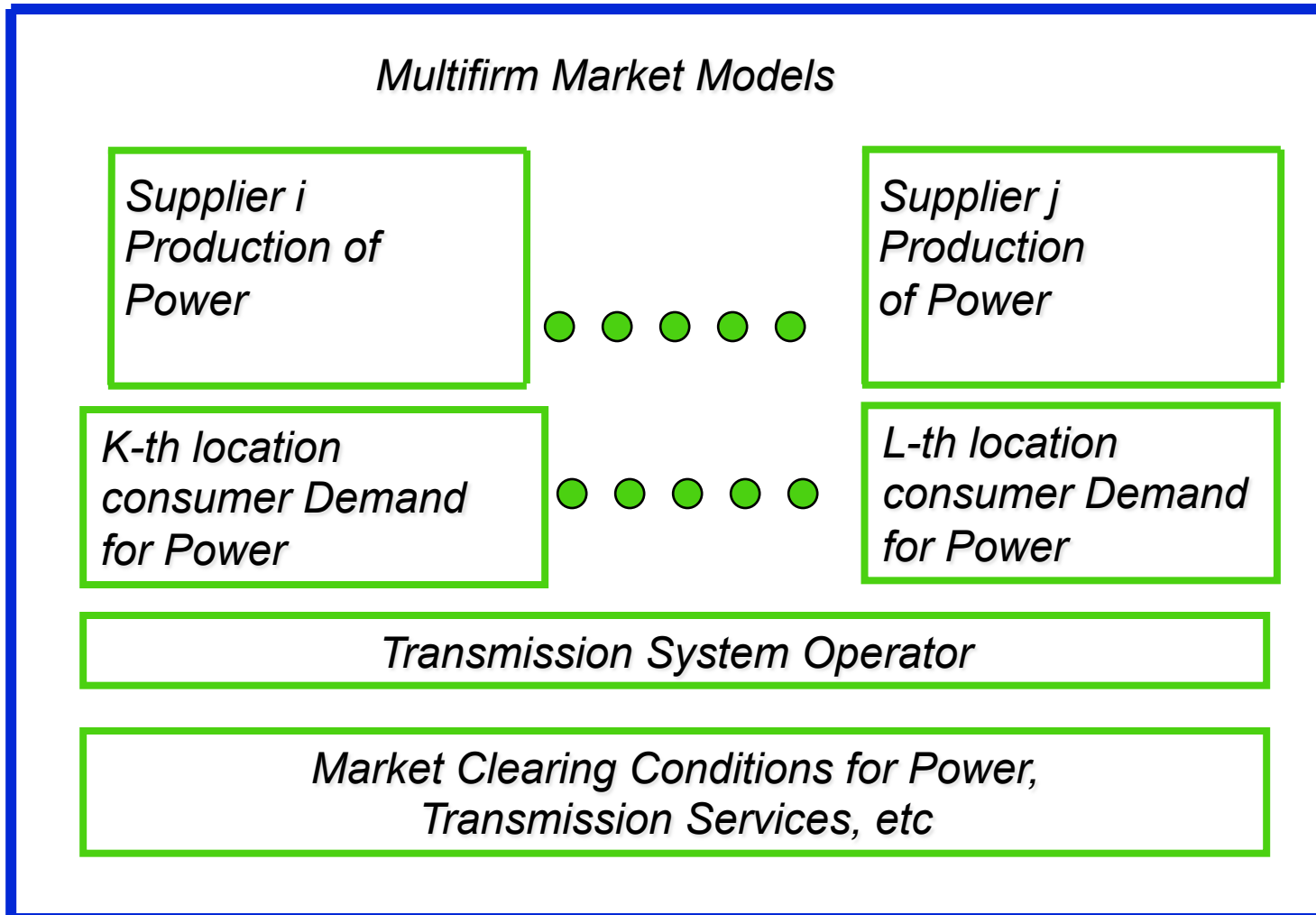
Solving a duopoly (2 firms) game

Find cell such that π_A is highest in column (Firm A maximizes its profit given y_B) and π_B is highest in row (Firm B maximizes its profit given y_A). In the below table, ***Bold italics*** represents Firm A's best response to y_B , while **Bold** represents Firm B's best response to y_A . The format of the table is:

$y_A \setminus y_B:$	B's y
A's y	π_A
	π_B

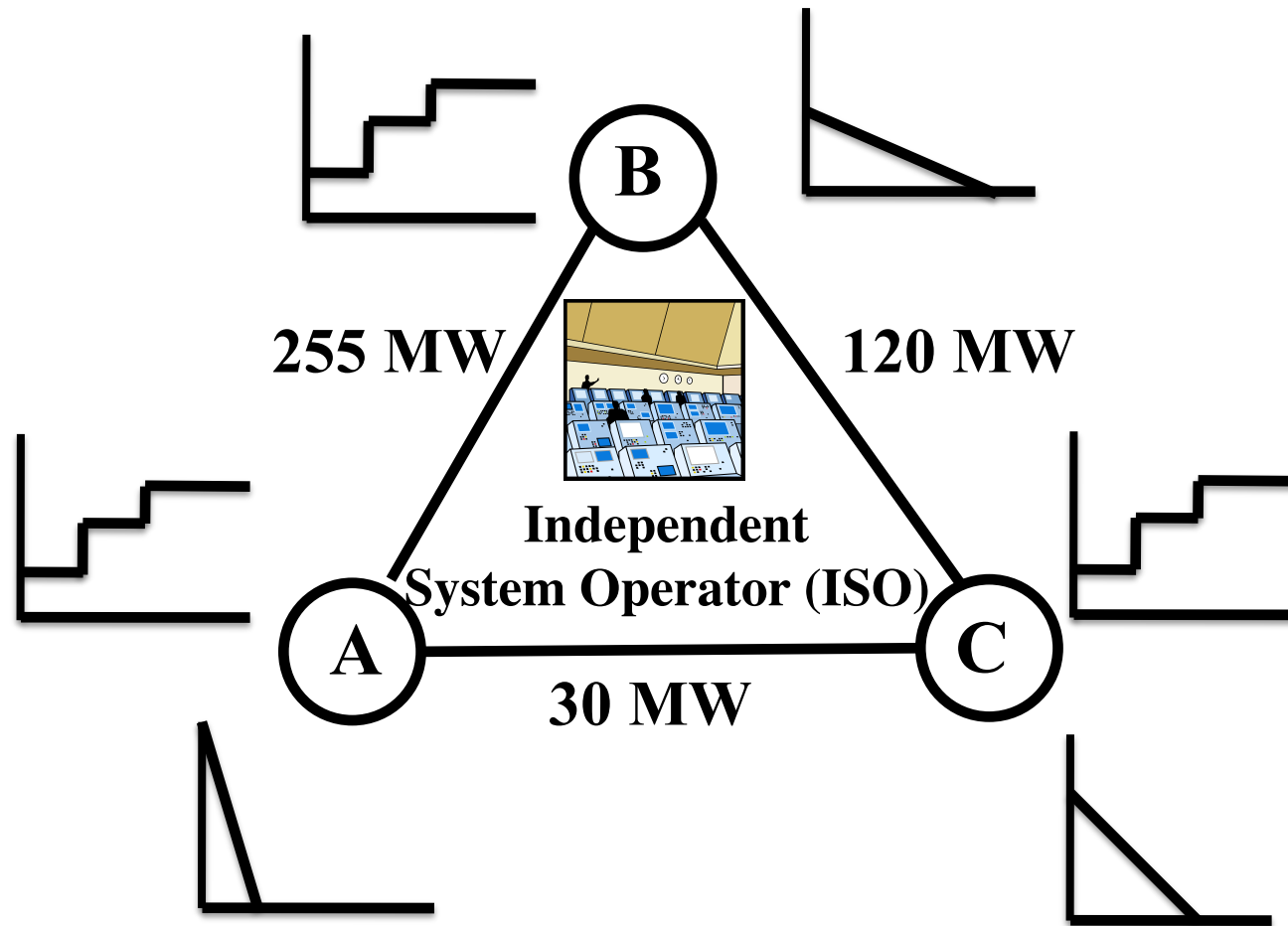
$y_A \setminus y_B:$	30	32	34	36	38	40	42	44	46	48	50
30	1655 1650	1688 1696	1500 1734	1560 1764	1530 1786	1500 1800	1470 1806	1440 1804	1410 1794	1380 1776	1350 1750
32	1696 1620	1664 1664	1632 1700	1600 1728	1568 1748	1536 1760	1504 1764	1472 1760	1440 1748	1408 1728	1376 1700
34	1734 1590	1700 1632	1666 1666	1632 1692	1598 1710	1564 1720	1530 1722	1496 1716	1462 1702	1428 1680	1394 1650
36	1764 1560	1728 1600	1692 1632	1656 1656	1620 1672	1584 1680	1548 1680	1512 1672	1476 1656	1440 1632	1404 1600
38	1786 1530	1748 1568	1710 1598	1672 1620	1634 1634	1596 1640	1558 1640	1520 1628	1482 1610	1444 1584	1406 1550
40	1800 1500	1760 1536	1720 1564	1680 1584	1640 1596	1600 1596	1560 1554	1520 1512	1480 1496	1440 1472	1400 1450
42	1806 1470	1764 1504	1722 1530	1680 1548	1638 1558	1596 1560	1554 1554	1512 1540	1470 1518	1428 1488	1386 1450
44	1804 1440	1760 1472	1716 1496	1672 1512	1628 1520	1584 1520	1540 1512	1496 1496	1452 1472	1408 1440	1364 1400
46	1794 1410	1748 1440	1702 1462	1656 1476	1610 1482	1564 1480	1518 1470	1472 1452	1426 1426	1380 1392	1334 1350
48	1776 1380	1728 1408	1680 1428	1632 1440	1584 1444	1536 1440	1488 1428	1440 1408	1392 1380	1344 1344	1296 1300
50	1750 1350	1700 1376	1650 1394	1600 1404	1550 1406	1500 1400	1450 1386	1400 1364	1350 1334	1300 1296	1250 1250

Model Structure



Equilibrium calculation: Solve n conditions for n unknowns

Electricity Markets

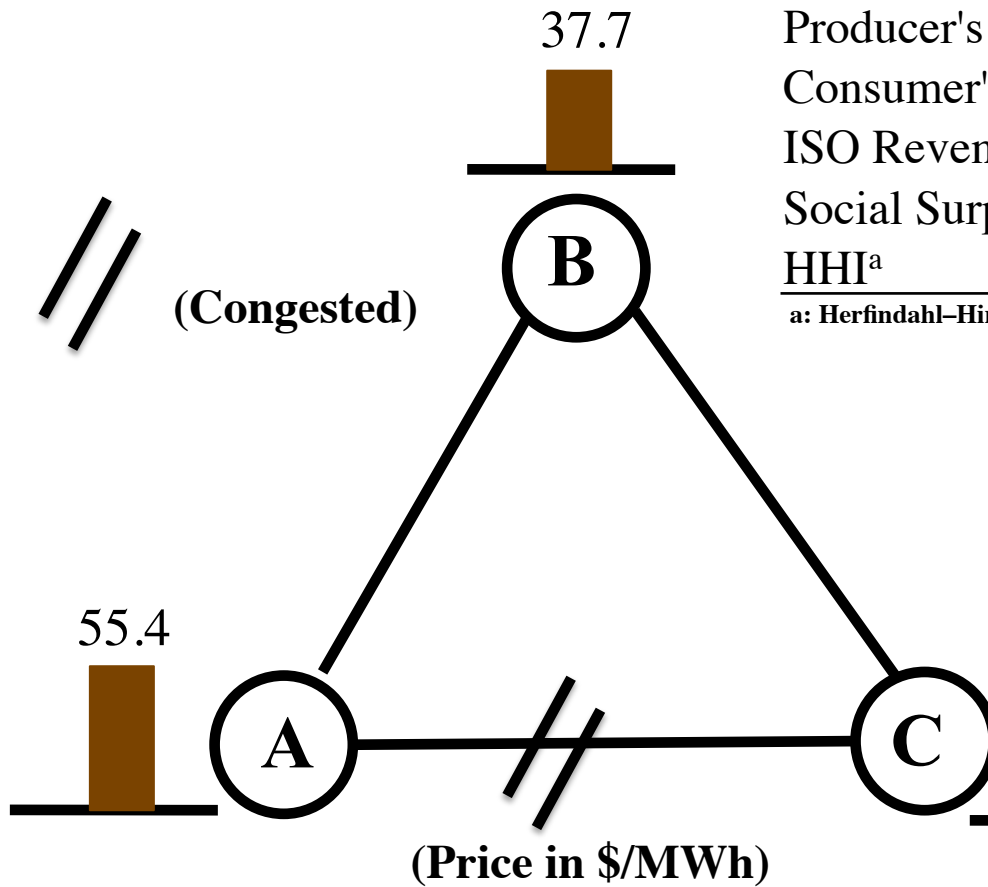


(10 Facilities, 3 producers)

Case Analysis

- **Case 1: Perfect Competition**
- **Case 2: Oligopoly Competition (Cournot)**
- **Case 3: Duopoly Competition (Merger)**
- **Case 4: Monopoly Competition**
- **Case 5: Transmission Investment (A-C: 30 → 430 MW)**

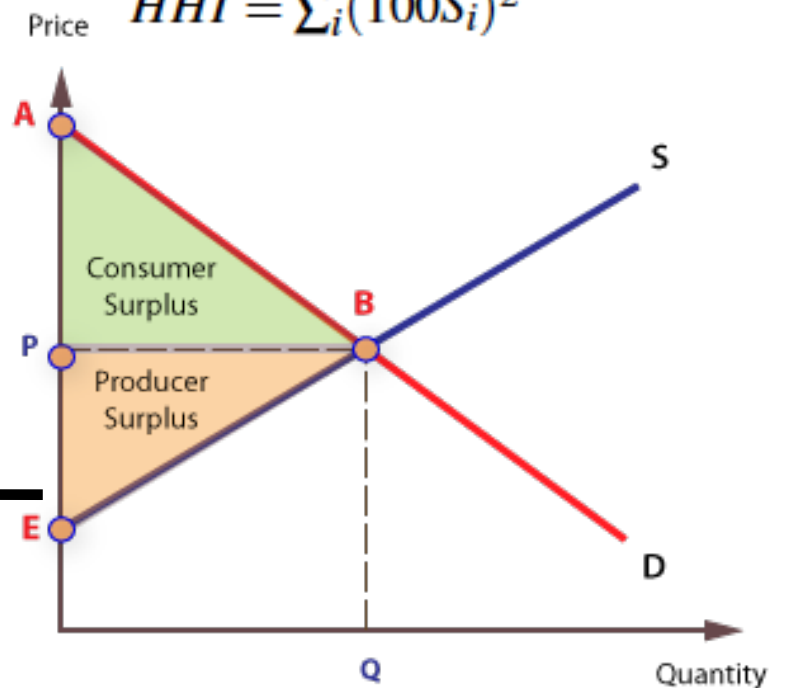
Results: Competitive Market



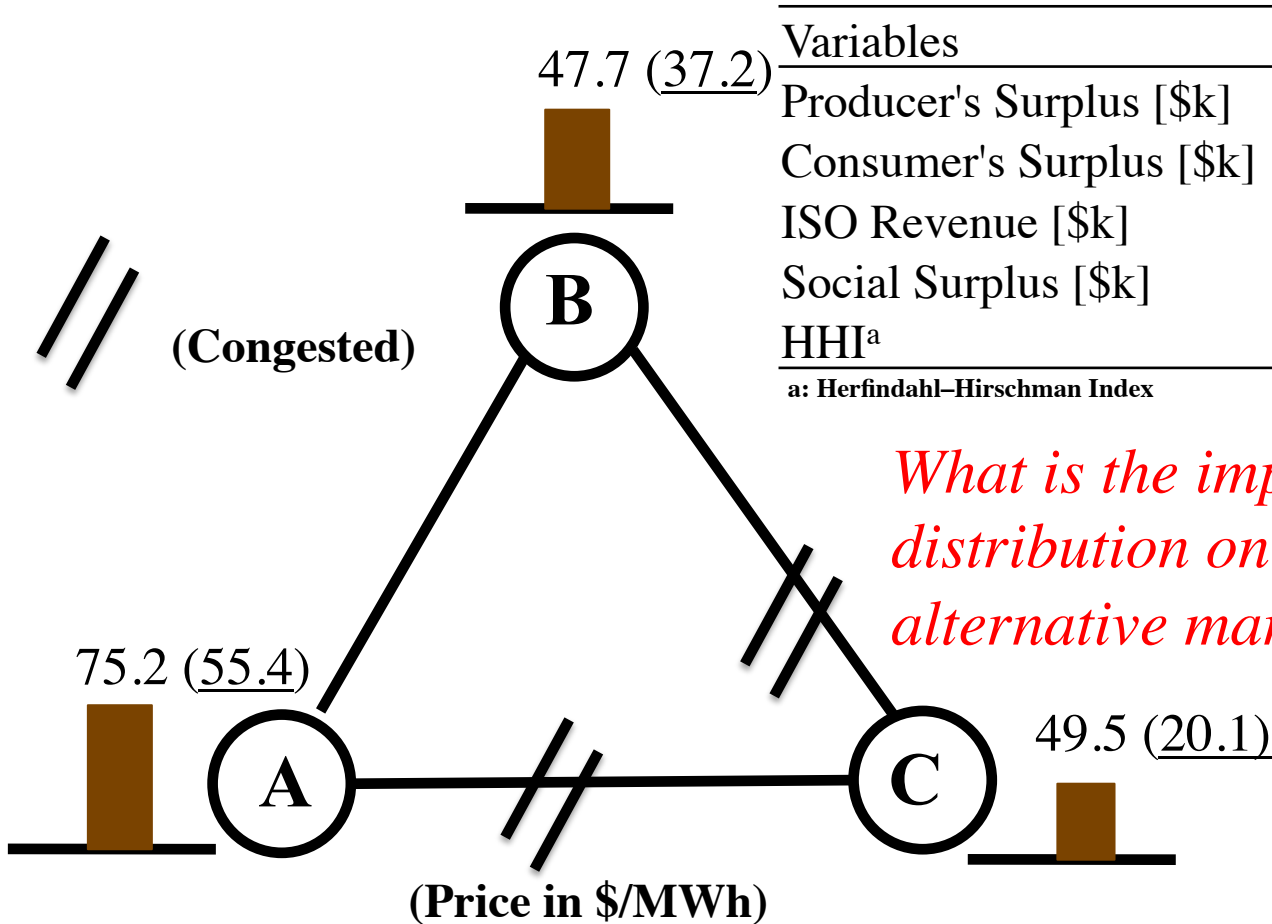
Variables	Firm 1	Firm 2	Firm 3
Producer's Surplus [\$k]	5.9	3.8	1.1
Consumer's Surplus [\$k]			131.8
ISO Revenue [\$k]			1.6
Social Surplus [\$k]			153.7
HHI ^a			3500

a: Herfindahl-Hirschman Index

$$HHI = \sum_i (100S_i)^2$$



Results: Oligopoly Cournot Markets

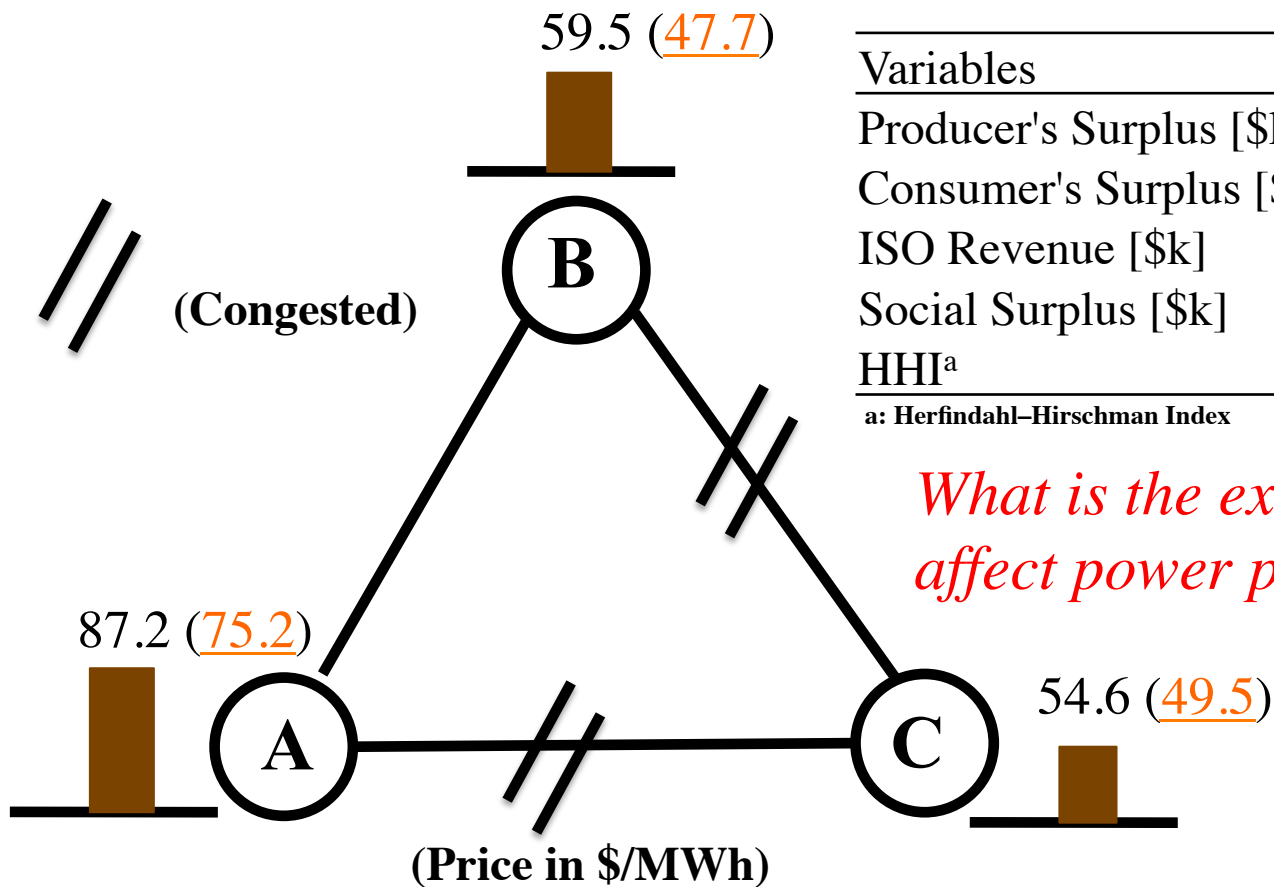


Variables	Firm 1	Firm 2	Firm 3
Producer's Surplus [\$k]	19.0 (5.9)	13.5(3.8)	20.7(1.1)
Consumer's Surplus [\$k]			92.1 (131.8)
ISO Revenue [\$k]			5.1 (1.6)
Social Surplus [\$k]			150.5 (153.7)
HHI ^a			3350 (3500)

a: Herfindahl-Hirschman Index

What is the impact on power prices & distribution on surplus under an alternative market structure?

Results: Duopoly Markets (Merger Analysis)

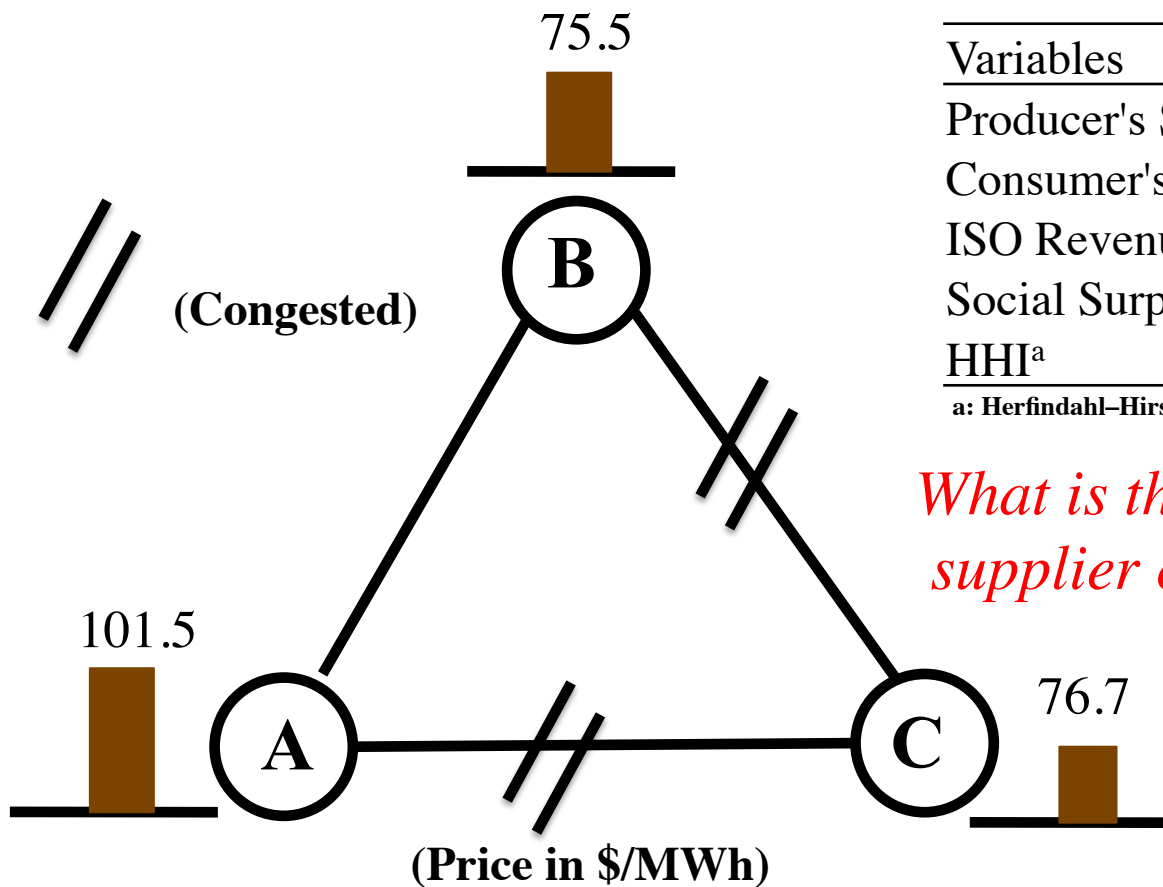


Variables	Firm 1	Firm 2
Producer's Surplus [\$k]	27.5	37.2
Consumer's Surplus [\$k]		76.3 (92.1)
ISO Revenue [\$k]		5.1 (5.1)
Social Surplus [\$k]	145.6	(150.5)
HHI ^a	5030	(3350)

a: Herfindahl-Hirschman Index

What is the extent that a merge might affect power prices?

Results: Monopoly Markets

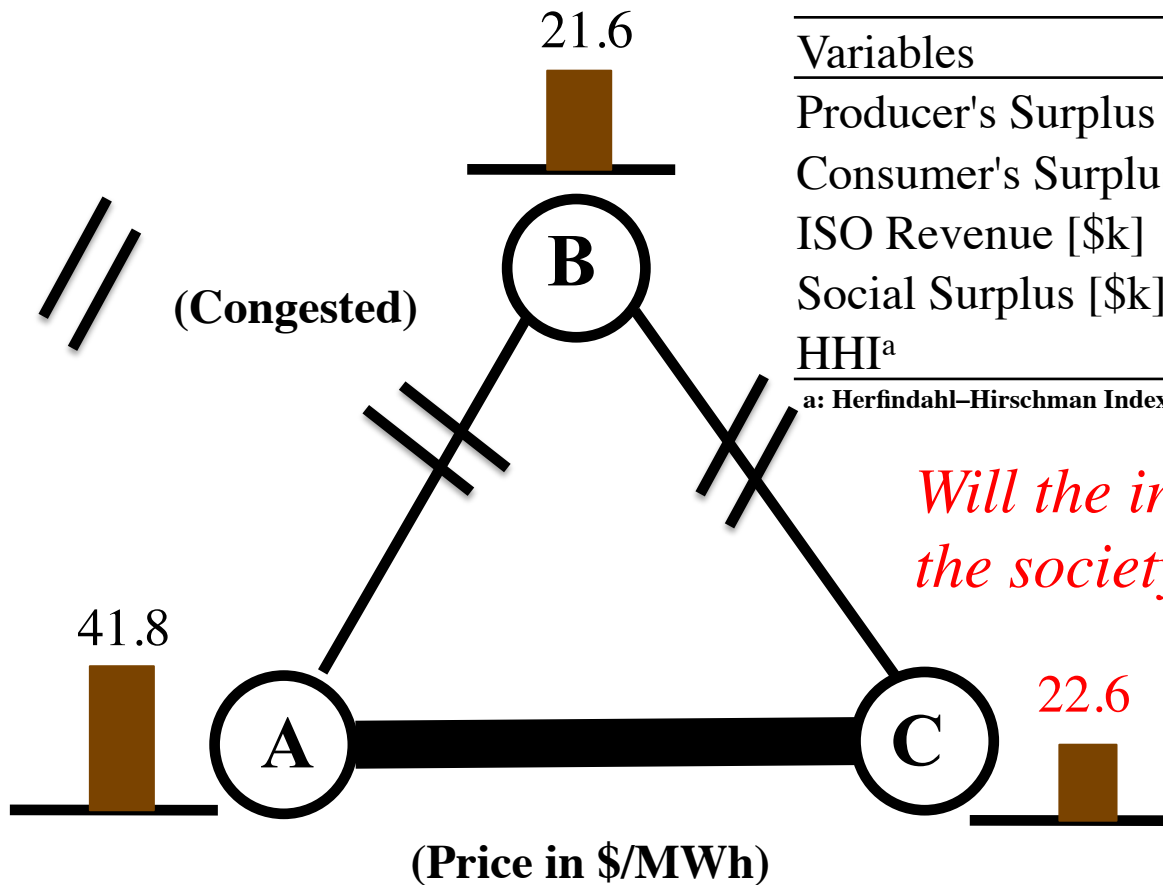


Variables	Firm 1
Producer's Surplus [\$k]	70.5
Consumer's Surplus [\$k]	54.6
ISO Revenue [\$k]	4.8
Social Surplus [\$k]	129.9
HHI ^a	10000

a: Herfindahl-Hirschman Index

What is the extent that a single supplier can exercise market power?

Results: Transmission Investment

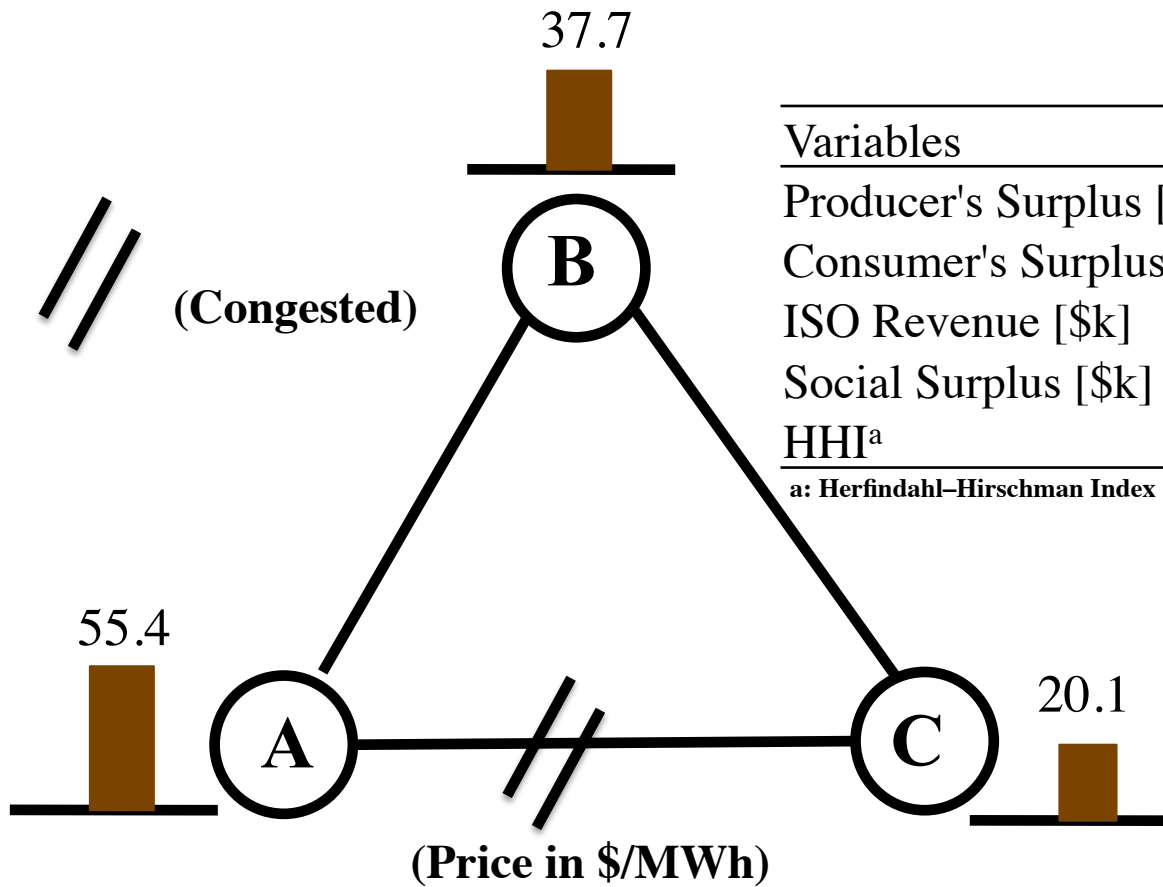


Variables	Firm 1	Firm 2	Firm 3
Producer's Surplus [\$k]	1.5	0.3	4.3
Consumer's Surplus [\$k]			151.1
ISO Revenue [\$k]			12.2
Social Surplus [\$k]			153.7
HHI ^a			43

a: Herfindahl-Hirschman Index

Will the investment be justified from the society's perspective?

Results: Competitive Market



Variables	Firm 1	Firm 2	Firm 3
Producer's Surplus [\$k]	5.9	3.8	10.6
Consumer's Surplus [\$k]			131.8
ISO Revenue [\$k]			1.6
Social Surplus [\$k]			153.7
HHI ^a			3500

a: Herfindahl-Hirschman Index

$$HHI = \sum_i (100S_i)^2$$

Advantages of Process Models for Policy & Markets Analysis

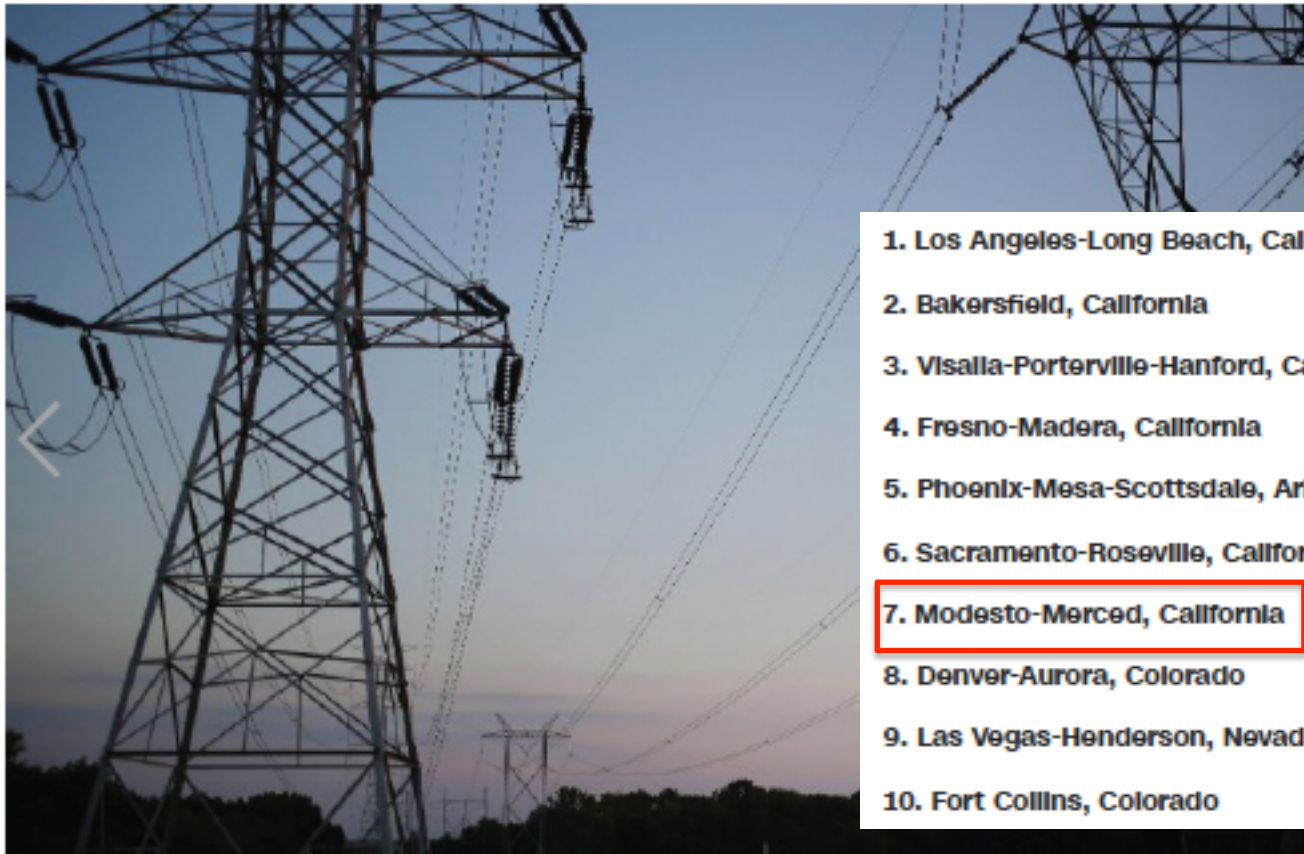
- **Explicitness:**
 - assumptions can be laid bare
 - changes in technology, policies, prices, objectives can be modeled by altering:
 - decision variables
 - objective function coefficients
 - constraints
- **Descriptive uses:**
 - show detailed cost, emission, technology choice impacts of policy changes
 - show changes in market prices, consumer welfare
- **Normative:**
 - identify better solutions through use of economic models
 - show tradeoffs among policy objectives

Other Research: Air Pollution & Public Transportation

Cities with most air pollution revealed

By **Jareen Imam**, CNN

🕒 Updated 4:59 PM ET, Wed April 20, 2016



10 photos: What's causing climate change? Meet the top 10 villains

<http://www.cnn.com/2016/04/20/health/air-pollution-report-irpt/index.html>

Other Research: Air Pollution & Public Transportation

Cities with most air pollution revealed

By **Jareen Imam**, CNN

🕒 Updated 4:59 PM ET, Wed April 20, 2016

George Thurston, professor of Environmental Medicine at New York University stated “ ... One of the cities with the best air quality is Salinas, California... ”



1. Los Angeles-Long Beach, California
2. Bakersfield, California
3. Visalia-Porterville-Hanford, California
4. Fresno-Madera, California
5. Phoenix-Mesa-Scottsdale, Arizona
6. Sacramento-Roseville, California
7. Modesto-Merced, California
8. Denver-Aurora, Colorado
9. Las Vegas-Henderson, Nevada
10. Fort Collins, Colorado

10 photos: What's causing climate change? Meet the top 10 villains

<http://www.cnn.com/2016/04/20/health/air-pollution-report-irpt/index.html>

Other Research: Air Pollution & Public Transportation

American Economic Journal: Economic Policy 2012, 4(1): 58–97
<http://dx.doi.org/10.1257/pol.4.1.58>

Green Infrastructure: The Effects of Urban Rail Transit on Air Quality[†]

By YIHSU CHEN AND ALEXANDER WHALLEY*

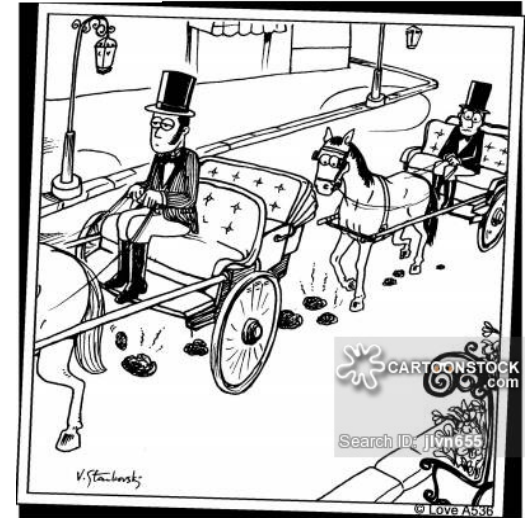
The transportation sector is a major source of air pollution worldwide, yet little is known about the effects of transportation infrastructure on air quality. This paper quantifies the effects of one major type of transportation infrastructure—urban rail transit—on air quality using the sharp discontinuity in ridership on opening day of a new rail transit system in Taipei. We find that the opening of the Metro reduced air pollution from one key tailpipe pollutant, carbon monoxide, by 5 to 15 percent. Little evidence that the opening of the Metro affected ground level ozone pollution is found however. (JEL L92, Q53, R41, R53)





Motivation

- Mass Transit sector is very large and growing:
 - 155 million people ride mass transit everyday in 110 cities.
 - 25 mass transit systems are currently under expansion or construction worldwide
 - Since 2000 – 37 new systems have opened.
 - Seen as potential ‘Green’ Policy: Beijing to become ‘Public transit city’
- Travel by mass transit produces significantly less pollution than private vehicle travel per mile.
- We know little about the impact of of urban mass transit on air pollution, as the size of two key elasticities is not clear:
 - Substitution towards low-emissions travel
 - Substitution towards total travel



And you thought exhaust problems were bad today.

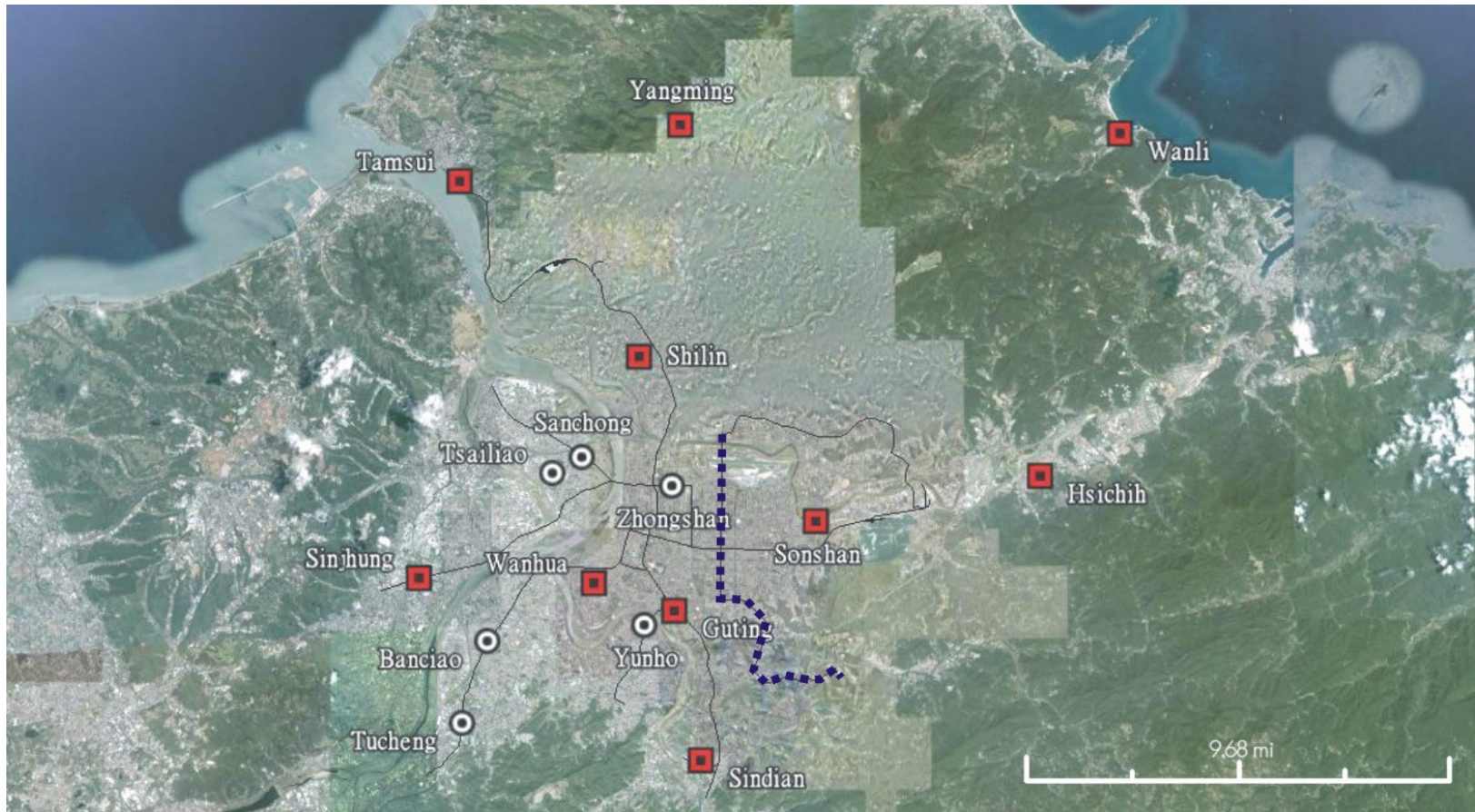
Research Questions

- Does urban mass transit infrastructure effect air quality?
- How do tailpipe and indirect pollutants respond?
- How big are the effects and their economic benefits?

What We Do

- Study the effects of the opening of the Metropolitan Rapid Transit (MRT) system in 1996 in Taipei
- Quantify the effects of the opening of the MRT on air quality

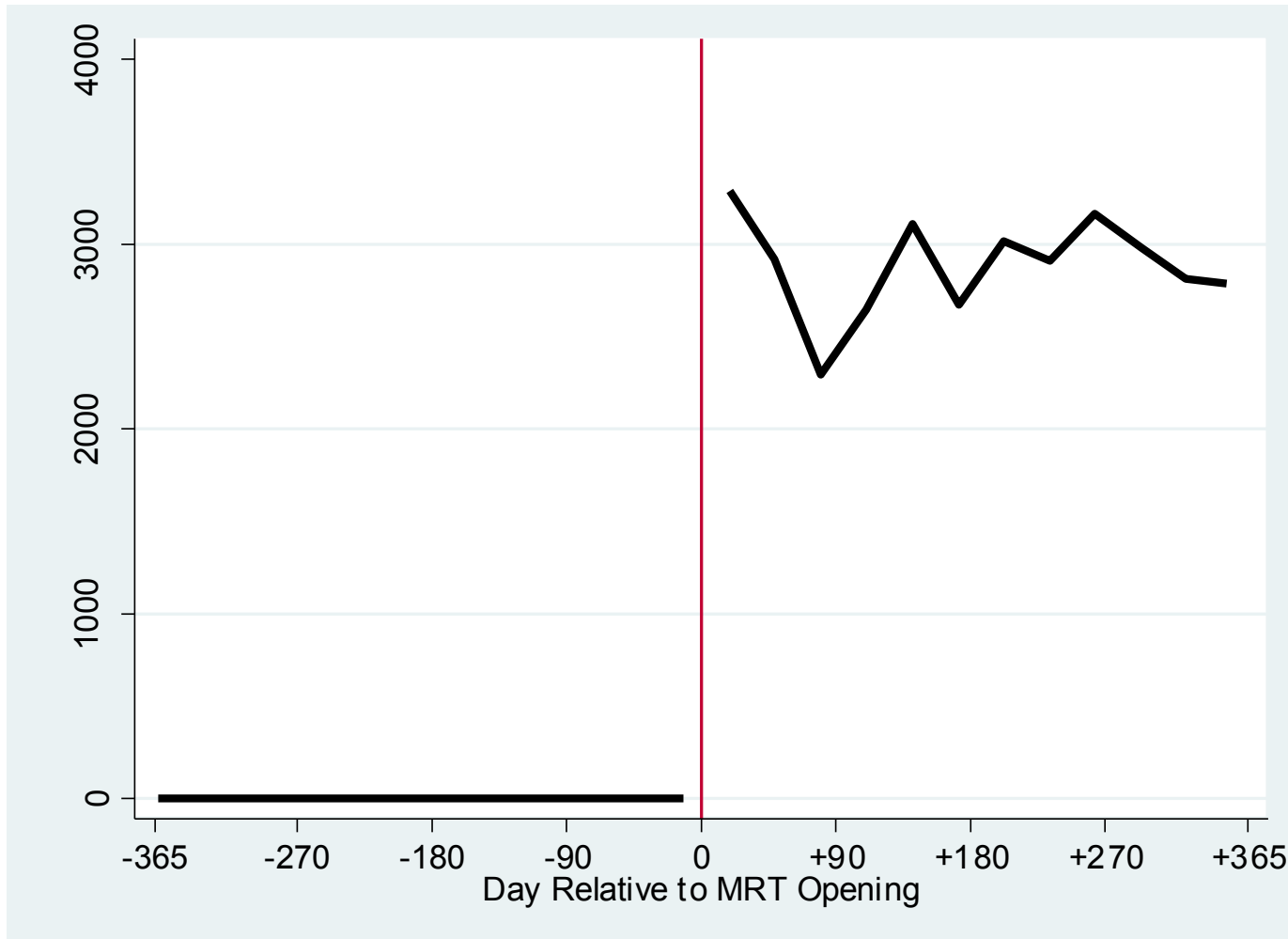
Map of Monitoring Stations and Metropolitan Rapid Transit System in Taipei



Map of Metropolitan Rapid Transit System in Taipei



Other Research: Air Pollution & Public Transportation



Empirical Approach

Regression Discontinuity Based Specification:

$$y_t = \delta_0 + \delta_1 \text{MetroOpen}_t + \delta_2 \mathbf{x}_t + \delta_3 \mathbf{P}(t) + \delta_4 \mathbf{P}(t) \times \text{MetroOpen}_t + e_t$$

y_t = pollution outcome in log scale

MetroOpen_t = MRT open

x_t = regulation, humidity, wind and temperature controls

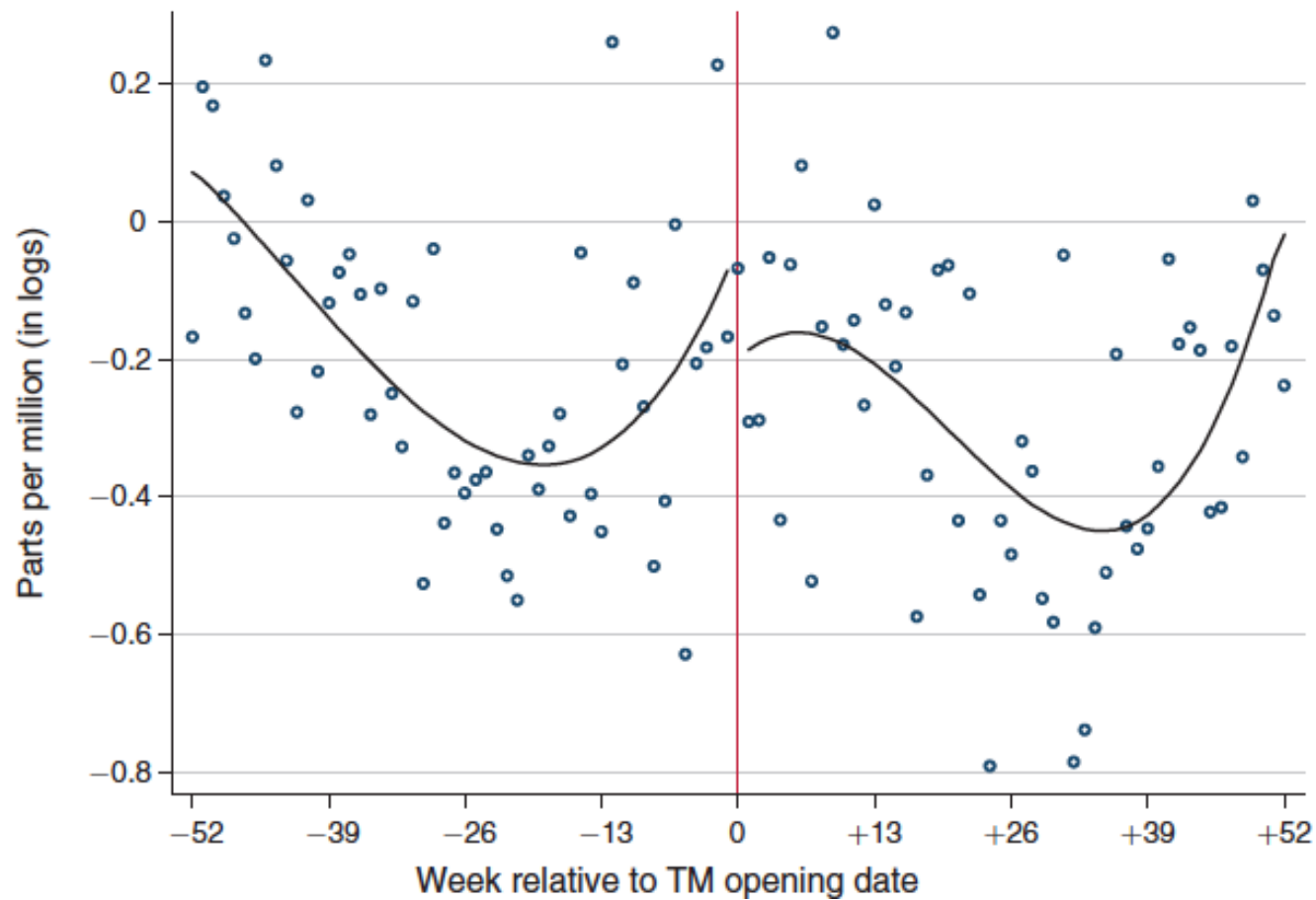
$P(t)$ = 3rd Order Polynomial time trend

- cluster standard errors at 5 week level

Results: Main Regressions

Dependent variable:	Log(CO)	Log(NO _x)	Log(O ₃)
Model:	DB-OLS (1)	DB-OLS 2)	DB-OLS (3)
Taipei Metro Open	-0.156** (0.059)	-0.083 (0.052)	-0.037 (0.063)
Observations	17,076	16,466	17,070

CO Mean Daily Pollution Level in Taipei, Polynomial Time Trend



Results: Falsification Tests

City: Model:	Taipei DB-OLS (1)	East coast DB-OLS (2)	Kaohsiung DB-OLS (3)
<i>Dependent variable =</i> Log(PM10)	0.041 (0.141)		
Log(SO ₂)	-0.249 (0.199)		
Log(CO)		0.088 (0.096)	-0.037 (0.052)
Log(NO _x)		0.062 (0.058)	-0.085 (0.057)
Log(O ₃)		0.165 (0.225)	0.056 (0.221)

MRT Health Benefits

MRT leads health effects valued at \$85 million USD the first year

- 1) CO \downarrow 1ppm, infant mortality \downarrow 2.5% (Currie, et *al.*, 2008)
- 2) Value per life: 1.2 million USD (Liu and Hammit 1999)
 \rightarrow 1.7 infant lives saved, \$8.6 million USD
(=0.834ppm \times 0.156 \times 0.025 \times 0.00666 \times 77029)
 \rightarrow (speculative) 58 elderly lives saved, \$76.4 million USD
- 3) Total benefits = \$85 million USD
- 4) Per passenger-mile benefit (\$) $\frac{8.6M}{40,000 \times 365 \times 9.3} = 0.063$.

Other Research: Biofuel -- Environment & Economics

Increased estimates of air-pollution emission from Brazilian sugar-cane ethanol

C-C. Tsao¹, J. E. Campbell^{1*}, M. Mena-Carrasco², S. N. Spak³, G. R. Carmichael³ and Y. Cher

Accelerating biofuel production has been promoted as an opportunity to enhance energy security, offset greenhouse-gas emissions and support rural economies. However, large uncertainties remain in the impacts of biofuels on air quality and climate^{1,2}. Sugar-cane ethanol is one of the most widely used biofuels, and Brazil is its largest producer³. Here we use a life-cycle approach to produce spatially and temporally explicit estimates of air-pollutant emissions over the whole life cycle of sugar-cane ethanol in Brazil. We show that even in regions where pre-harvest field burning has been eliminated on half the croplands, regional emissions of air pollutants continue to increase owing to the expansion of sugar-cane growing areas, and burning continues to be the dominant life-cycle stage for emissions. Comparison of our estimates of burning-phase emissions with satellite estimates of burning in São Paulo state suggests that sugar-cane field burning is not fully accounted for in satellite-based inventories, owing to the small spatial scale of individual fires. Accounting for this effect leads to revised regional estimates of burned area that are four times greater than some previous estimates. Our revised emissions maps thus suggest that biofuels may have larger impacts on regional climate forcing and human health than previously thought.

Air-pollutant emissions from biofuel production and combustion may have significant impacts on climate and air quality. The change in vehicle emissions that would result from a large-scale conversion from gasoline to E85 (a blend of up to 85% ethanol with gasoline or another hydrocarbon) in the United States could have significant health consequences, by increasing tropospheric ozone concentrations⁴, for example. Monetizing the health and climate impacts of US ethanol emissions (fuel production and

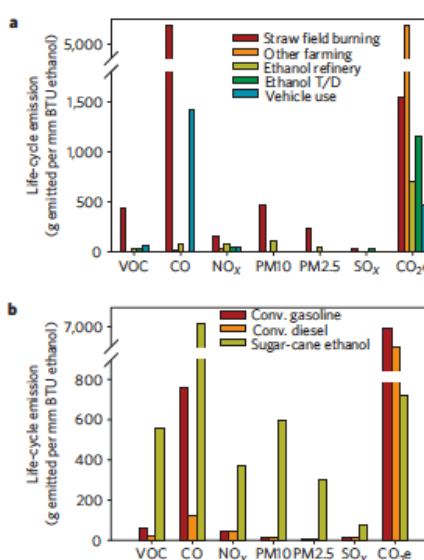
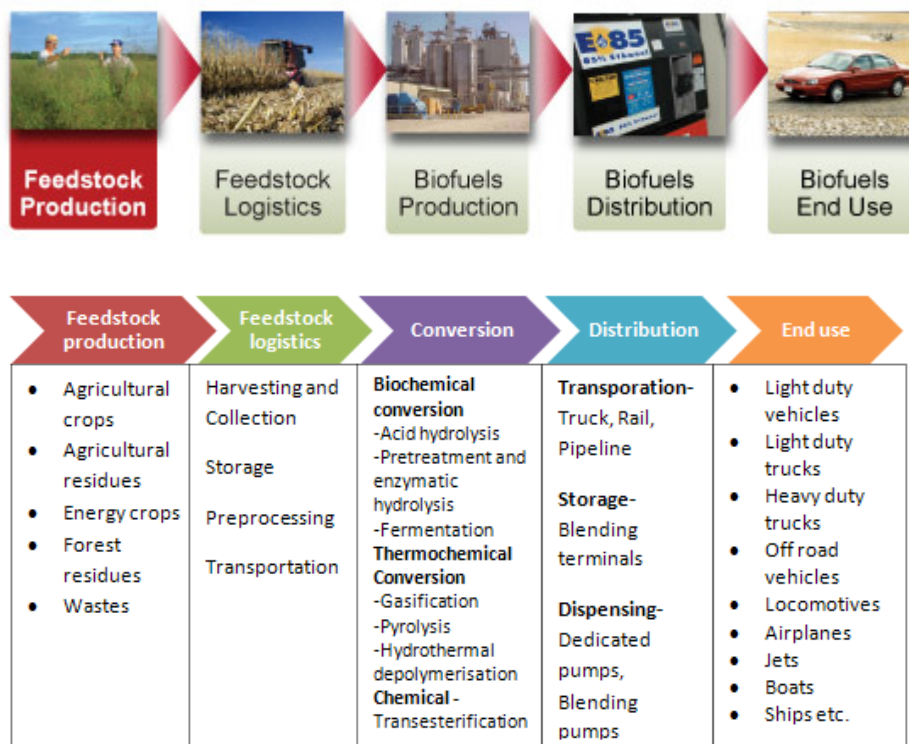


Figure 1 | Comparisons of life-cycle emissions for sugar-cane ethanol and conventional liquid fuels. a, Life-cycle emissions per unit energy of sugar-cane ethanol produced within five life-cycle phases. Although our life-cycle emissions account for a mix of sugar-cane fields where the burning practice is used and not used, the burning-phase emissions shown

Biofuels Supply Chain



Robert T. Clemen
Terence Reilly

MAKING HARD DECISIONS

with DecisionTools®



TIM 165

Decision
Analysis

Coming soon in
UCSC
Fall, 2016



Thank you!
Yihsu Chen, Ph.D.

Email: yihsuchen@ucsc.edu