



Mercados Eléctricos

Despacho

Subastas

Contratos

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Ciclo de 1 semana – 6 sesiones
Sesión 4- 4 de diciembre de 2025

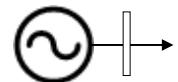


COMPETENCIA PERFECTA

EJEMPLO - 1 GENERADOR

Perfect Competition

Competencia Perfecta: Un Generador - Demanda Elástica



$$P^{\max} = 100 \frac{\$}{\text{MWh}}, m_D = 20 \frac{\$}{\text{MW}^2\text{h}} \quad P^{\min} = 5 \frac{\$}{\text{MWh}}, m_O = 15 \frac{\$}{\text{MW}^2\text{h}}$$

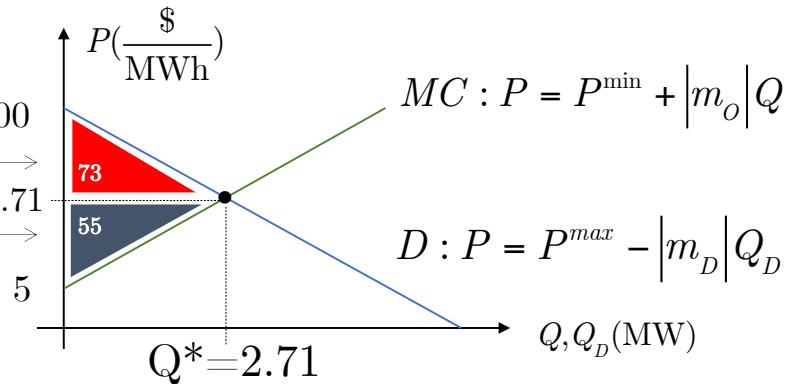
$$\begin{bmatrix} Q^* \\ Q_D^* \\ \lambda^* \end{bmatrix} = \begin{bmatrix} 15 & 0 & -1 \\ 0 & -20 & -1 \\ -1 & 1 & 0 \end{bmatrix}^{-1} \begin{bmatrix} -5 \\ -100 \\ 0 \end{bmatrix} = \begin{bmatrix} 2.7143 \\ 2.7143 \\ 45.7143 \end{bmatrix}$$

$$E_p(Q = 2.71) = 1 - \frac{100}{20 \cdot 2.7} = -0.85$$

$$CSP^* = P^{\max}Q^* - \frac{1}{2}|m_D|Q^{*2} - P^*Q^* \rightarrow P^* = 45.71$$

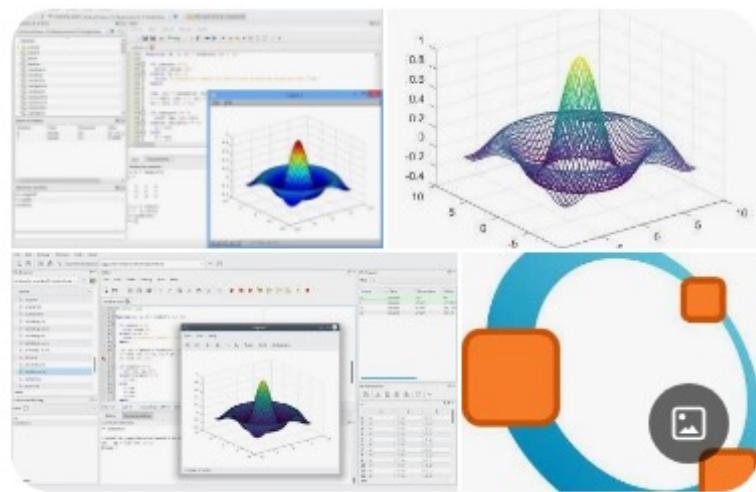
$$PSP^* = P^*Q^* - P^{\min}Q^* - \frac{1}{2}|m_O|Q^{*2} \rightarrow P^* = 45.71$$

SW* = 128 \\$/h



GNU Octave

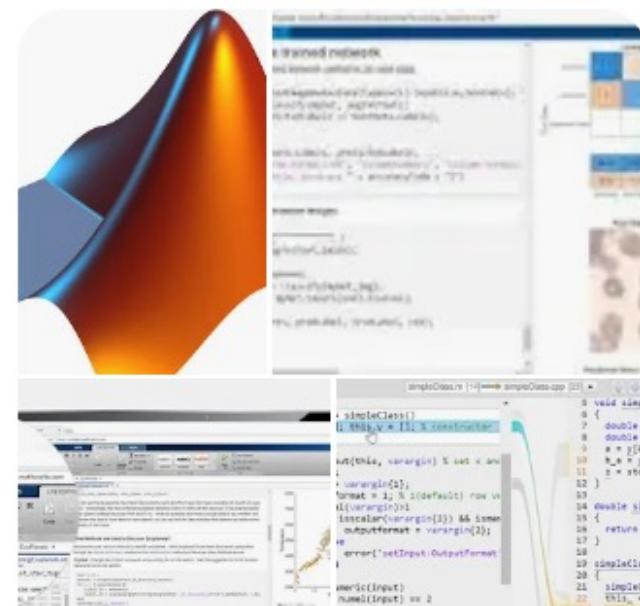
Programming language ::



<https://octave.org/>

MATLAB

Programming language ::



```
%Clearing the market (1 producer - 1 demand)
% cp: perfect competition
m01=15;pmax=100;mD=20;p1min=5
B=[-p1min; -pmax; 0];
Acp=[m01 0 -1;
      0 -mD -1;
      -1 1 0];
Solcp=inv(Acp)*B
Past=Solcp(3);
Qast=Solcp(1);
CSP = 0.5*Qast*(pmax-Past) % Área bajo la curva superior
CSPx= pmax*Qast-0.5*mD*Qast*Qast-Past*Qast
PSPx= Past*Qast-p1min*Qast-0.5*m01*Qast*Qast
PSP = 0.5*Qast*(Past-p1min) % Área bajo la curva inferior
SW=PSP+CSP
```

ED_cp1.m

Octave Online

Variables	
[3x3] Acp	octave:>1> %Clearing the market (1 producer - 1 demand)
[3x1] B	% cp: perfect competition
# CSP	m01=15;pmax=100;mD=20;p1min=5
# CSPx	B=[-p1min; -pmax; 0];
# PSP	Acp=[m01 0 -1; 0 -mD -1; -1 1 0];
# PSPx	Solcp=inv(Acp)*B
# Past	Past=Solcp(3);
# Qast	Qast=Solcp(1);
# SW	CSP = 0.5*Qast*(pmax-Past) % Area bajo la curva superior
[3x1] Solcp	CSPx= pmax*Qast-0.5*mD*Qast*Qast-Past*Qast
# ans	PSPx= Past*Qast-p1min*Qast-0.5*m01*Qast*Qast
# mD	PSP = 0.5*Qast*(Past-p1min) % Area bajo la curva inferior
# m01	SW=PSP+CSP
# p1min	p1min = 5
# pmax	Solcp = 2.7143 2.7143 45.7143
	 CSP = 73.673 CSPx = 73.673 PSPx = 55.255 PSP = 55.255 SW = 128.93

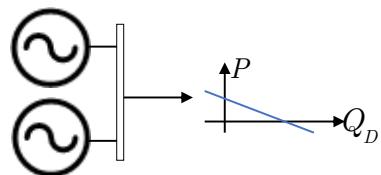
COMPETENCIA PERFECTA

EJEMPLO - 2 GENERADORES

Perfect Competition



Competencia Perfecta: Dos Generadores - Demanda Elástica



$$P_{\max} = 100 \frac{\$}{\text{MWh}}, m_D = 20 \frac{\$}{\text{MW}^2\text{h}} \quad P_1^{\min} = 5 \frac{\$}{\text{MWh}}, m_{o1} = 15 \frac{\$}{\text{MW}^2\text{h}}$$

$$P_2^{\min} = 10 \frac{\$}{\text{MWh}}, m_{o2} = 10 \frac{\$}{\text{MW}^2\text{h}}$$

$$\begin{bmatrix} Q_1^* \\ Q_n^* \\ Q_D^* \\ \lambda \end{bmatrix} = \begin{bmatrix} |m_{o1}| & 0 & 0 & -1 \\ 0 & |m_{o_n}| & 0 & -1 \\ 0 & 0 & -|m_D| & -1 \\ -1 & -1 & 1 & 0 \end{bmatrix}^{-1} \begin{bmatrix} -P_1^{\min} \\ -P_n^{\min} \\ -P^{\max} \\ 0 \end{bmatrix} = \begin{bmatrix} 1.61 \\ 1.92 \\ 3.53 \\ 29.23 \end{bmatrix}$$

$$PSP_1^* = P^* \cdot Q_1^* - \int_0^{Q_1^*} P_1^{\min} + |m_{o1}| Q_1 dQ_1 = 19.57$$

$$PSP_2^* = P^* \cdot Q_2^* - \int_0^{Q_2^*} P_2^{\min} + |m_{o2}| Q_2 dQ_2 = 18.49$$

$$PSP^* = PSP_1^* + PSP_2^* = 38.06$$

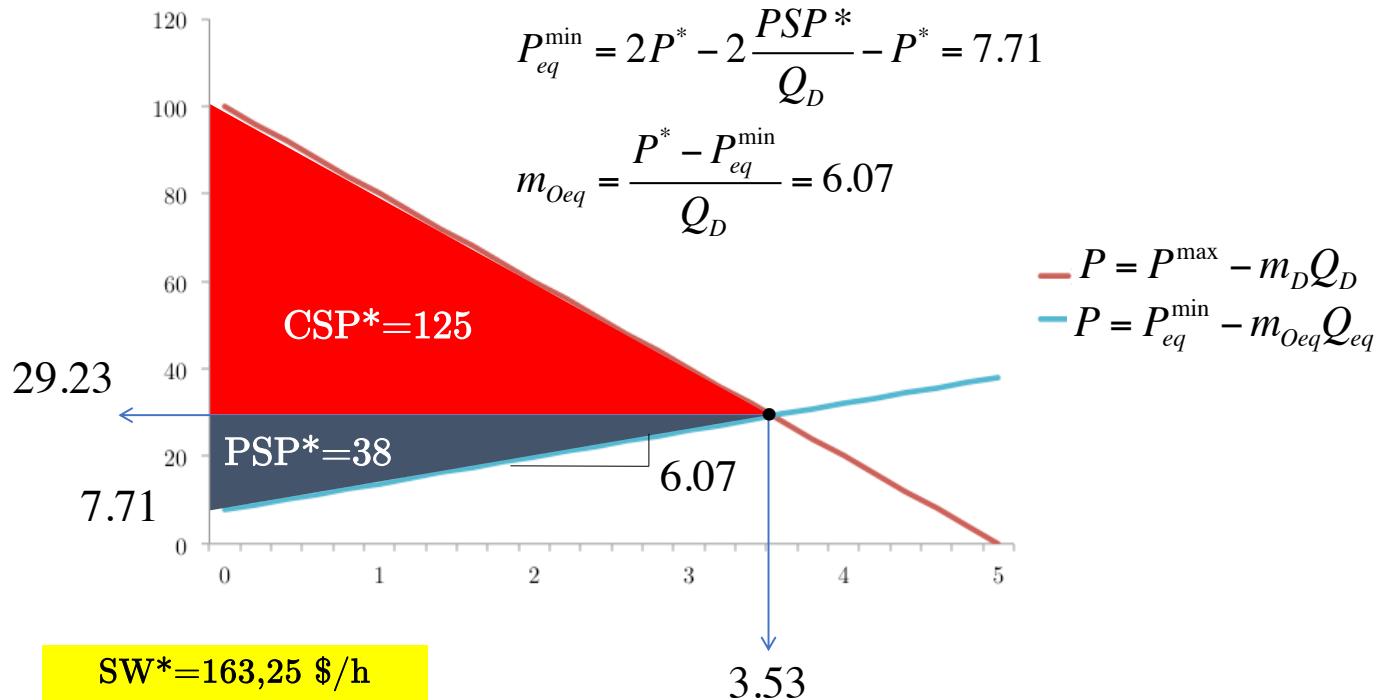
$$CSP^* = \int_0^{Q_D^*} P^{\max} - |m_D| Q_D dQ_D - P^* \cdot Q_D^* = 125.2$$

$$SW = CSP^* + PSP^* = 163.25$$

Perfect Competition



Competencia Perfecta: Dos Generadores - Demanda Elástica



```

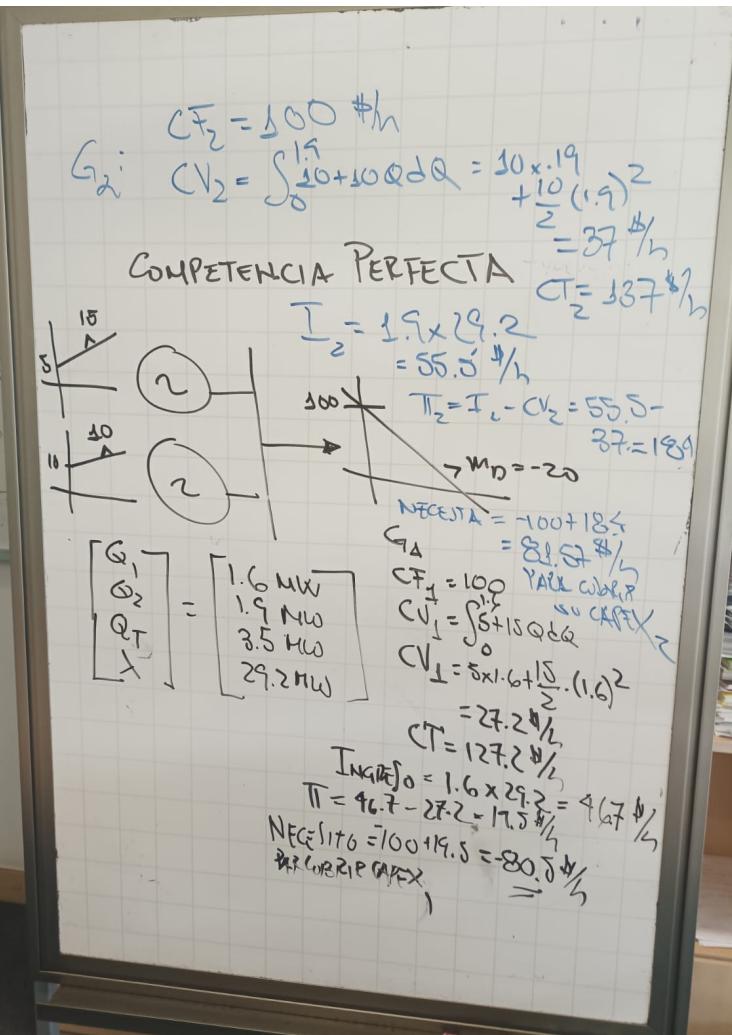
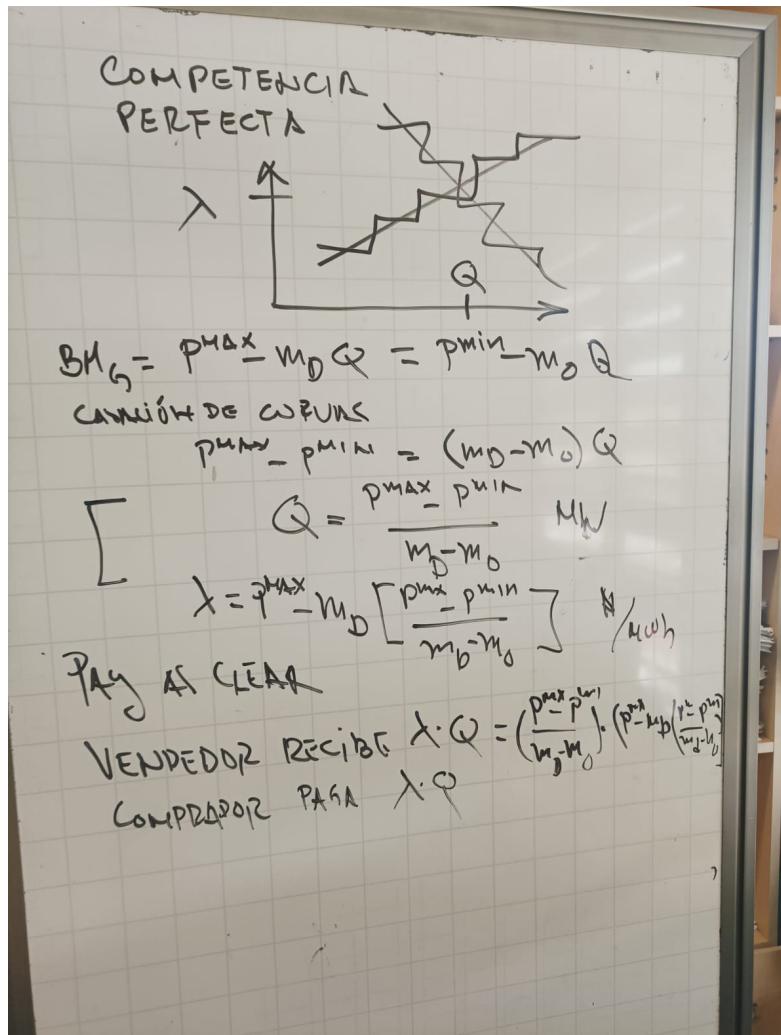
%Clearing the market (2 producers - 1 demand)
% cp: perfect competition
mO1=15;mO2=10;pmax=100;mD=20;p1min=5;p2min=10;
B=[-p1min; -p2min; -pmax; 0];
Acp=[mO1 0 0 -1;
      0 mO2 0 -1;
      0 0 -mD -1;
      -1 -1 1 0];
Solcp=inv(Acp)*B
Past=Solcp(4);
Qast1=Solcp(1);
Qast2=Solcp(2);
Qast=Solcp(3);
PSP1=Past*Qast1-p1min*Qast1-0.5*mO1*Qast1*Qast1
PSP2=Past*Qast2-p2min*Qast2-0.5*mO2*Qast2*Qast2
PSP=PSP1+PSP2
CSP = pmax*Qast-0.5*mD*Qast*Qast-Past*Qast
SW=PSP+CSP
%Elasticity at equilibrium point
DP=(pmax-Past);
DQ=0-Qast;
e=(DQ/Qast)/(DP/Past)

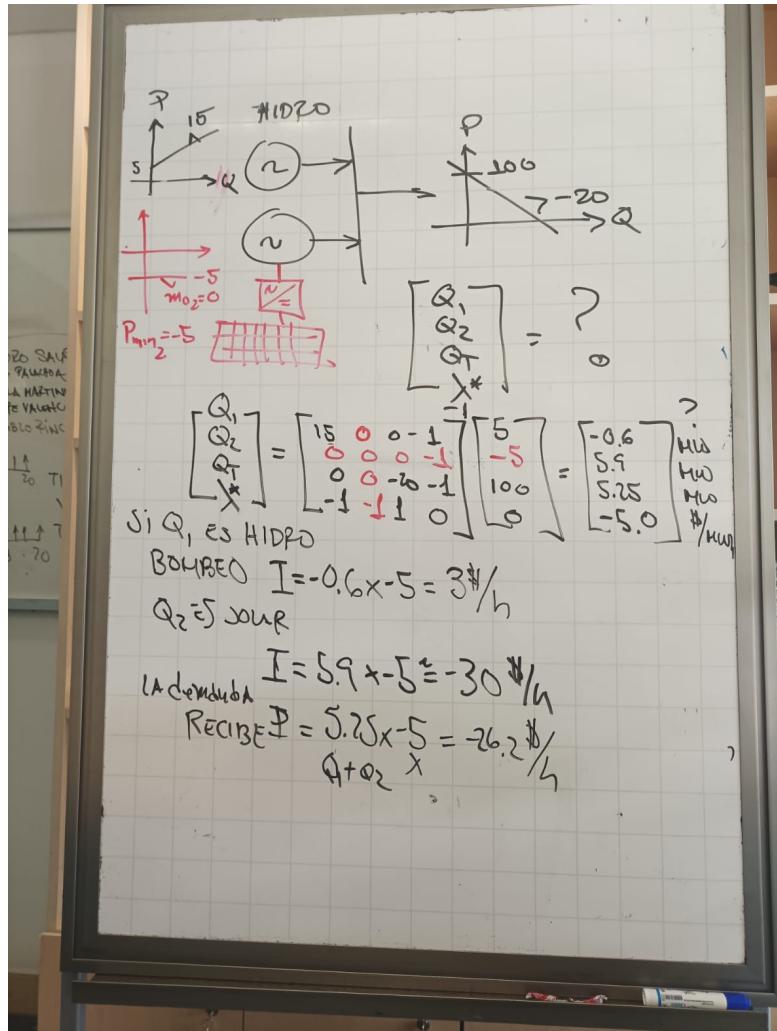
```

Octave Online

Variables	octave:12> %Clearing the market (2 producers - 1 demand)
[4x4] Acp	% cp: perfect competition
[4x1] B	m01=15;m02=10;pmax=100;mD=20;p1min=5;p2min=10;
# CSP	B=[-p1min; -p2min; -pmax; 0];
# CSPx	Acp=[m01 0 0 -1;
# DP	0 m02 0 -1;
# DQ	0 0 -mD -1;
# PSP	-1 -1 1 0];
# PSP1	Solcp=inv(Acp)*B
# PSP2	Past=Solcp(4);
# PSPx	Qast1=Solcp(1);
# Past	Qast2=Solcp(2);
# Qast	Qast=Solcp(3);
# Qast1	PSP1=Past*Qast1-p1min*Qast1-0.5*m01*Qast1*Qast1
# Qast2	PSP2=Past*Qast2-p2min*Qast2-0.5*m02*Qast2*Qast2
# SW	PSP=PSP1+PSP2
[4x1] Solcp	CSP = pmax*Qast-0.5*mD*Qast*Qast-Past*Qast
# ans	SW=PSP+CSP
# e	%Elasticity at equilibrium point
# mD	DP=(pmax-Past);
# m01	DQ=0-Qast;
# m02	e=(DQ/Qast)/(DP/Past)
# p1min	Solcp =
# p2min	1.6154
# pmax	1.9231
	3.5385
	29.2308
	PSP1 = 19.571
	PSP2 = 18.491
	PSP = 38.062
	CSP = 125.21
	SW = 163.27
	e = -0.4130







G2 es una planta PV
 con $m_{02} =$
 y $p_{min2} = -5 \$/\text{MWh}$

Perfect Competition

MAX SOCIAL WELFARE: CSP+PSP WITH PRODUCTION CONSTRAINTS (MAX-MIN)

Lagrangiano:

$$L = \sum_{i=1}^n \left[FC + \rho_i^{\min} P_{Gi} + \frac{1}{2} |m_{O_i}| (P_{Gi})^2 \right] - \left[\rho^{\max} P_D - \frac{1}{2} |m_D| (P_D)^2 \right] + \lambda \left[P_D - \sum_{i=1}^n P_{Gi} \right]$$

$$+ \sum_{i=1}^n \mu_i [P_{Gi} - P_{Gi}^{\max}] + \sum_{i=1}^n \nu_i [P_{Gi}^{\min} - P_{Gi}]$$

$$\frac{\partial L}{\partial P_{Gi}} = \rho_i^{\min} + |m_{O_i}| P_{Gi} - \lambda + \mu_i - \nu_i = 0 \quad i = 1, \dots, n$$

$$\frac{\partial L}{\partial P_D} = -\rho^{\max} + |m_D| P_D + \lambda = 0$$

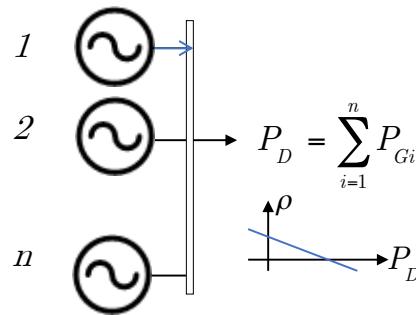
See Matlab program: [EDLimits.m](#)

$$\frac{\partial L}{\partial \lambda} = P_D - \sum_{i=1}^n P_{Gi} = 0,$$

$$\frac{\partial L}{\partial \mu_i} = P_{Gi} - P_{Gi}^{\max} + \beta_i = 0, \quad \beta_i \geq 0, \quad i = 1, \dots, n \quad \frac{\partial L}{\partial \nu_i} = P_{Gi}^{\min} - P_{Gi} + \vartheta_i = 0, \quad \vartheta_i \geq 0, \quad i = 1, \dots, n$$

$$\mu_i (P_{Gi} - P_{Gi}^{\max}) = 0, \quad i = 1, \dots, n \quad \nu_i (P_{Gi}^{\min} - P_{Gi}) = 0, \quad i = 1, \dots, n$$

Perfect Competition

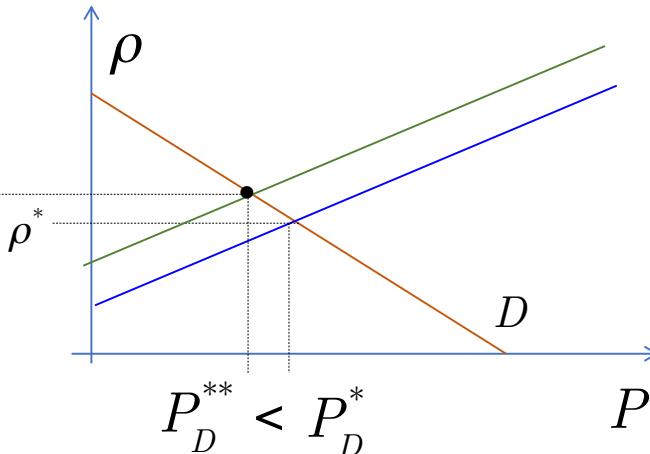


$$P_{Gi} < P_{Gi}^{\max} \Rightarrow \mu_i = 0 \quad i = 1, \dots, n$$

$$\rho^{**} = \lambda = \rho_1^{\min} + |m_{o1}| P_{G1}^{**} + \mu_1 = \frac{\partial SC}{\partial P_{G1}} > \rho^*$$

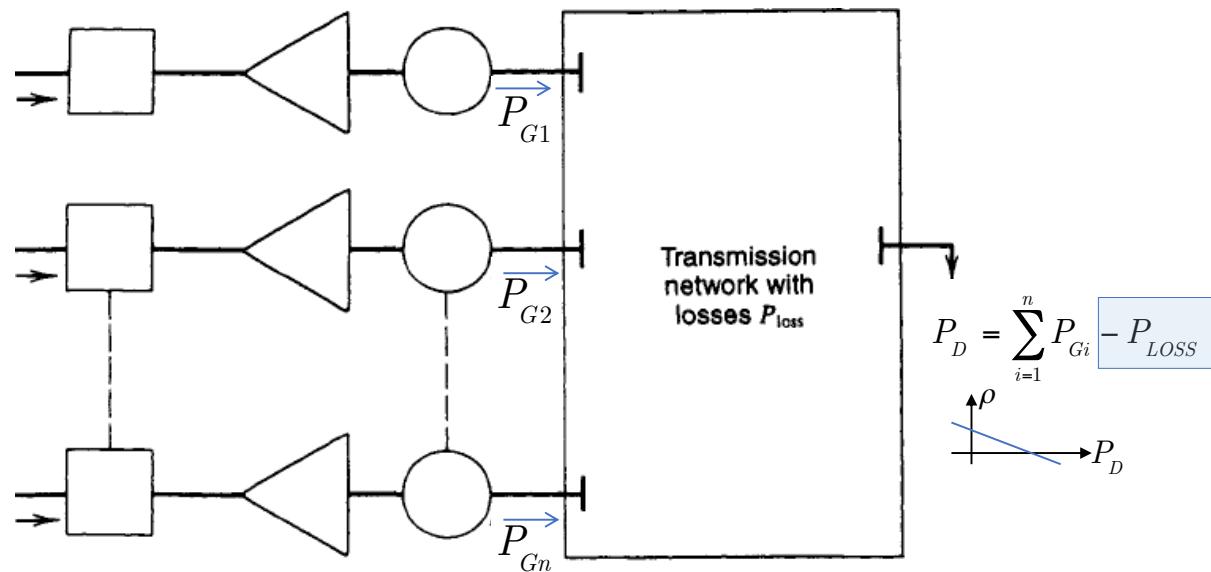
$$\rho^{**} = \lambda = \rho_2^{\min} + |m_{o2}| P_{G2}^{**} + \mu_2 = \frac{\partial SC}{\partial P_{G2}} > \rho^*$$

	Benefit	Cost
$G1$	$P_{G1}^{**}(\lambda - \mu_1)$	$\rho_1^{\min} P_{G1}^{**} + \frac{1}{2} m_{o1} (P_{G1}^{**})^2$
$G2$	$P_{G2}^{**}(\lambda - \mu_1)$	$\rho_2^{\min} P_{G2}^{**} + \frac{1}{2} m_{o2} (P_{G2}^{**})^2$
D	$\rho^{\max} P_D^{**} - \frac{1}{2} m_D (P_D^{**})^2$	$P_D^{**}(\lambda - \mu_1)$



Perfect Competition

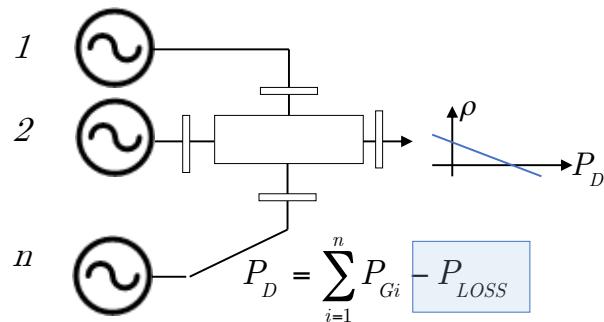
MAX SOCIAL WELFARE DISPATCH
 WITH PRODUCTION CONSTRAINTS (MAX-MIN)
 AND NETWORK LOSSES



Perfect Competition



MAX SOCIAL WELFARE DISPATCH WITH PRODUCTION CONSTRAINTS (MAX-MIN) AND NETWORK LOSSES



$$\sum_{i=1}^n P_{Gi} \cdot \lambda_{Gi} \neq P_D \lambda_D$$

Network Revenue:

$$NR = P_D \lambda_D - \sum_{i=1}^n P_{Gi} \cdot \lambda_{Gi}$$

$$\min SC = -SW = CSP + PSP - NR$$

$$\min TC - U = \sum_{i=1}^n \left[\int_0^{P_{Gi}} MC(P_{Gi}) dP_{Gi} \right] - \int_0^{P_D} D(P_D) dP_D$$

$$\text{s.t. } P_D = \sum_{i=1}^n P_{Gi} - P_{LOSS}$$

Perfect Competition



MAX SOCIAL WELFARE: CSP+PSP WITH PRODUCTION CONSTRAINTS (MAX-MIN) AND NETWORK LOSSES

Lagrangiano:

$$L = \sum_{i=1}^n \left[FC + \rho_i^{\min} P_{Gi} + \frac{1}{2} |m_{O_i}| (P_{Gi})^2 \right] - \left[\rho^{\max} P_D - \frac{1}{2} |m_D| (P_D)^2 \right] + \lambda \left[P_D - \sum_{i=1}^n P_{Gi} - P_{LOSS} \right]$$

$$+ \sum_{i=1}^n \mu_i [P_{Gi} - P_{Gi}^{\max}] + \sum_{i=1}^n \nu_i [P_{Gi}^{\min} - P_{Gi}]$$

$$\frac{\partial L}{\partial P_{Gi}} = \rho_i^{\min} + |m_{O_i}| P_{Gi} - \lambda \left(1 - \frac{dP_{LOSS}}{dP_{Gi}} \right) + \mu_i - \nu_i = 0 \quad i = 1, \dots, n$$

$$\frac{\partial L}{\partial P_D} = -\rho^{\max} + |m_D| P_D + \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = P_D - \sum_{i=1}^n P_{Gi} - P_{LOSS} = 0,$$

$$\frac{\partial L}{\partial \mu_i} = P_{Gi} - P_{Gi}^{\max} + \beta_i = 0, \quad \beta_i \geq 0, \quad i = 1, \dots, n \quad \frac{\partial L}{\partial \nu_i} = P_{Gi}^{\min} - P_{Gi} + \vartheta_i = 0, \quad \vartheta_i \geq 0, \quad i = 1, \dots, n$$

$$\mu_i (P_{Gi} - P_{Gi}^{\max}) = 0, \quad i = 1, \dots, n \quad \nu_i (P_{Gi}^{\min} - P_{Gi}) = 0, \quad i = 1, \dots, n$$

Coeficiente
Incremental de Pérdidas del nodo de generación i

Perfect Competition

MAX SOCIAL WELFARE: CSP+PSP
 WITH PRODUCTION CONSTRAINTS (MAX-MIN)
 AND NETWORK LOSSES

$$P_{LOSS} = \mathbf{P}_G \mathbf{B}_{LOSS} \mathbf{P}_G + \mathbf{B}_0 \mathbf{P}_G + B_{00}$$

$$P_{LOSS} \approx \sum_{i=1}^n B_{ii} P_{Gi}^2 = \begin{bmatrix} P_{G1} & \dots & P_{Gn} \end{bmatrix} \begin{bmatrix} B_{11} & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & B_{nn} \end{bmatrix} \begin{bmatrix} P_{G1} \\ \vdots \\ P_{Gn} \end{bmatrix}$$

See Bloss Formula: See Kirchmayer 1958,
 Wood & Wollemborg 1984

Perfect Competition



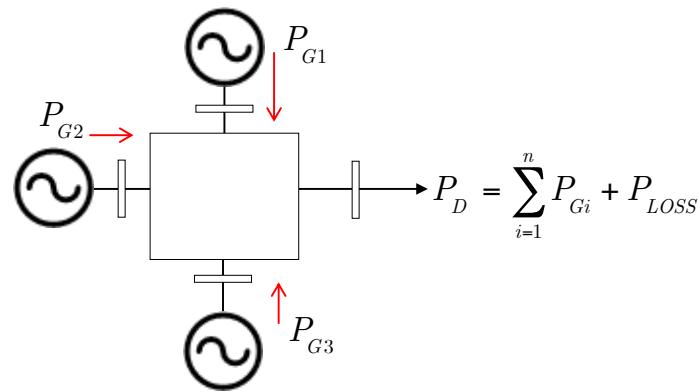
Wood & Wollemborg Example 3C [BUT WITH ELASTIC DEMAND]

$$F_1(P_1) = H_1(P_1) \times 1.1 = 561 + 7.92P_1 + 0.001562P_1^2 \text{ R/h}$$

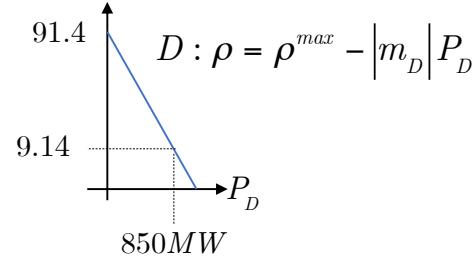
$$F_2(P_2) = H_2(P_2) \times 1.0 = 310 + 7.85P_2 + 0.00194P_2^2 \text{ R/h}$$

$$F_3(P_3) = H_3(P_3) \times 1.0 = 78 + 7.97P_3 + 0.00482P_3^2 \text{ R/h}$$

$$P_{\text{loss}} = 0.00003P_1^2 + 0.00009P_2^2 + 0.00012P_3^2$$



$$P_D = \sum_{i=1}^n P_{Gi} + P_{LOSS}$$



Perfect Competition

**MAX SOCIAL WELFARE: CSP+PSP
WITH PRODUCTION CONSTRAINTS (MAX-MIN)
AND NETWORK LOSSES**

Lagrangiano:

$$L = \sum_{i=1}^3 \left[FC + \rho_i^{\min} P_{Gi} + \frac{1}{2} |m_{oi}| (P_{Gi})^2 \right] - \left[\rho_D^{\max} P_D - \frac{1}{2} |m_D| (P_D)^2 \right] + \lambda \left[P_D - \sum_{i=1}^3 P_{Gi} + P_{LOSS} \right]$$

$$\frac{\partial L}{\partial P_{Gi}} = \rho_i^{\min} + |m_{oi}| P_{Gi} - \lambda(1 - 2B_{ii} P_{Gi}) = 0, \quad i = 1, \dots, 3$$

$$\frac{\partial L}{\partial P_D} = -\rho_D^{\max} + |m_D| P_D + \lambda = 0$$

$$\frac{\partial L}{\partial \lambda} = P_D - \sum_{i=1}^3 P_{Gi} + \sum_{i=1}^3 B_{ii} P_{Gi}^2 = 0,$$

5 ecuaciones 5 incógnitas: $P_{G1}, P_{G2}, P_{G3}, P_D, \lambda$

$$\lambda_{Gi} = \lambda(1 - \frac{dP_{LOSS}}{dP_{Gi}}) = \lambda(1 - 2B_{ii} P_{Gi}), \quad i = 1, \dots, 3$$

Perfect Competition



Network Revenue is not null

$$NR = P_D \lambda - \sum_{i=1}^n P_{Gi} \cdot \lambda_{Gi}$$

In general, taking into account generation and transmission limits:

$$\lambda_{Gi} = \lambda \left(1 - \frac{dP_{LOSS}}{dP_{Gi}}\right) - \mu_i^G + \nu_i^G - \mu_i^T$$

$$pf_{Gi} = \frac{1}{\left(1 - \frac{dP_{LOSS}}{dP_{Gi}}\right)}$$

$$pf_{Gi} > 0 \Rightarrow \frac{dP_{LOSS}}{dP_{Gi}} > 0 \Rightarrow \Delta P_{Gi} \propto \Delta P_{LOSS}$$

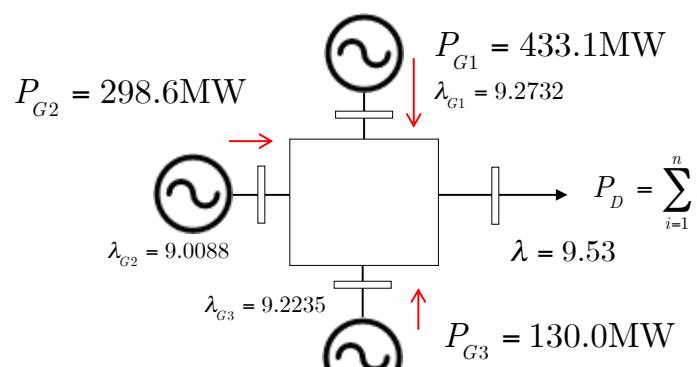
Perfect Competition



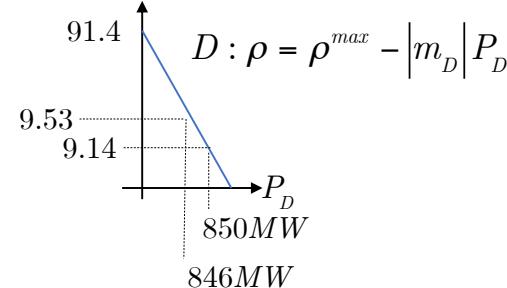
Wood & Wollemberg Example 3C [BUT WITH ELASTIC DEMAND]

$$\lambda_{G1} = \lambda - \lambda \frac{dLoss}{dP_{G1}} = \lambda(1 + 2B_{11}P_{G1}) = 9.53(1 - 0.0264) = 9.2732$$

Penalty
Factor
 $PF_1 = 1/0.97 > 1$



See Matlab program: [EDLosses.m](#)



Bulk Generation

$$P_D = \sum_{i=1}^n P_{Gi} + P_{LOSS} = 861.84 + 15.68 = 846.1 \text{ MW}$$

Losses Demand

Social Cost = -3.5373e+04

Perfect Competition



Methods Dispatch Solution

When no Kuhn-Tucker coefficients are active, the ED problem is Determined. Minimum is obtained from n non-linear equations With n unknowns.

Linear programming (e.g. Simplex) for linearized cost and loss functions

In general, NonLinear Optimization Solvers should be applied.

Gradient

Interior Point

Newton

Tools: Minos/Gams, Frontline Solver
Matlab, SciPy, etc.

Other: Genetic Algorithm, Neural Networks

(See Zhu, 4.8, 4.9)



COMPETENCIA PERFECTA

SUBASTA
PAY AS CLEAR

In a Pay-as-Clear market, **all** winning power plants receive the same price per megawatt-hour (MWh), regardless of what they originally bid. That price is determined by the **most expensive unit** needed to meet the demand (the "Marginal Unit"). [🔗](#)

The Demand (The Buyers)

- **Load 1:** 50 MW
- **Load 2:** 50 MW
- **Total Market Demand:**

$$50 + 50 = 100 \text{ MW}$$

The Supply (The Sellers) We have two power plants with different technologies and costs. To illustrate how the auction works effectively, we will assume **Plant A** (perhaps Solar or Wind) has a limitation (availability) or a strategy that restricts its bid volume, forcing the market to use the more expensive plant.

- **Plant A (Renewable/Cheap):**
 - **Capacity:** 100 MW
 - **Bid Offer:** 50 MW at \$20/MWh
 - *(Note: Even though it has 100 MW capacity, let's assume availability is low or it only offers 50 MW into the auction).*
- **Plant B (Gas/Expensive):**
 - **Capacity:** 100 MW
 - **Bid Offer:** 100 MW at \$80/MWh

2. The Auction (The Merit Order)

The System Operator (SO) ranks the bids from cheapest to most expensive to create the "Merit Order." [🔗](#)

Rank	Generator	Quantity Offered	Bid Price	Cumulative Supply
1	Plant A	50 MW	\$20	50 MW
2	Plant B	100 MW	\$80	150 MW

3. Clearing the Market

The System Operator fills the 100 MW demand starting with the cheapest power.

1. **First 50 MW:** The operator takes all 50 MW from **Plant A** because it is the cheapest (\$20).
 - *Demand met: 50 MW.*
 - *Demand remaining: 50 MW.*
2. **Next 50 MW:** The operator must now turn to **Plant B**. They take 50 MW from Plant B to finish filling the demand.
 - *Demand met: 100 MW.*
 - *Demand remaining: 0 MW.*

The Market Clearing Price: The auction stops at **Plant B**. Because Plant B is the last unit needed to satisfy the demand (the Marginal Unit), its bid sets the price for everyone.

$$\text{Clearing Price} = \$80/\text{MWh}$$

4. The Result: Pay-as-Clear

This is the crucial part. Even though Plant A offered to sell its power for \$20, it gets paid the Clearing Price of \$80.

Why is this good for Plant A?

Plant A receives a **surplus** (profit). This creates an incentive to invest in cheap, efficient technologies (like wind, solar, or hydro).

- **Plant A Revenue:** $50 \text{ MW} \times \$80 = \$4,000$
- *If it were "Pay-as-Bid", they would have only made $50 \times \$20 = \$1,000$.*

The Final Payouts

Generator	Power Generated	Bid Price	Paid Price	Total Revenue
Plant A	50 MW	\$20	\$80	\$4,000
Plant B	50 MW	\$80	\$80	\$4,000
Plant B (Remaining)	50 MW	\$80	N/A	\$0 (Not dispatched)

1. Mathematical Formulation

Sets & Indices:

- $g \in \{1, 2\}$: The set of generators (Plant A, Plant B).
- $L \in \{1, 2\}$: The set of loads.

Parameters:

- C_g : Bid price of generator g (\$/MWh).
- P_g^{max} : Maximum capacity of generator g (MW).
- D_l : Demand of load l (MW). Total Demand $D_{total} = \sum D_l$.

Variables:

- $x_g \geq 0$: Power generated by plant g (MW).

Objective Function: Minimize Total Cost:

$$\min Z = \sum_g C_g \cdot x_g$$

Constraints:

- Power Balance (Demand Constraint):** The total generation must equal the total load. The **dual variable** (shadow price) of this constraint is the **Clearing Price**.

$$\sum_g x_g = \sum_l D_l \quad [\lambda]$$

- Capacity Constraints:**

$$x_g \leq P_g^{max} \quad \forall g$$

```

Gams
$ scenarios / Scen1, Scen2 /;

Parameters
  Bid(g)      Bid Price ($ per MWh)
  Cap(g)      Capacity (MW)
  Demand      Total Demand (MW) / 100 /;

* Define Prices
Bid('PlantA') = 20;
Bid('PlantB') = 80;

* Define Variables
Positive Variable x(g);
Variable z;

* Define Equations
CostObj   Objective Function
BalEq     Supply Demand Balance
CapLim(g) Capacity Limit;

CostObj..  z == sum(g, Bid(g) * x(g));
BalEq..    sum(g, x(g)) == Demand;
CapLim(g).. x(g) =l= Cap(g);

Model Auction /all/;

* Reporting Parameter
Parameter Report(s, *);

* --- SCENARIO 1: Plant A restricted to 50MW ---
Cap('PlantA') = 50;
Cap('PlantB') = 100;

Solve Auction using lp minimizing z;

Report('Scen1', 'Gen_A') = x.l('PlantA');
Report('Scen1', 'Gen_B') = x.l('PlantB');
Report('Scen1', 'ClearingPrice') = BalEq.m;

* --- SCENARIO 2: Plant A full 100MW ---
* Note: To strictly ensure the price drops to 20 in LP,
* we assume A has margin or is the marginal unit.
Cap('PlantA') = 100;
Cap('PlantB') = 100;

Solve Auction using lp minimizing z;

Report('Scen2', 'Gen_A') = x.l('PlantA');
Report('Scen2', 'Gen_B') = x.l('PlantB');
Report('Scen2', 'ClearingPrice') = BalEq.m;

Display Report;

```



1. Mathematical Formulation

Sets & Indices:

- $g \in \{1, 2\}$: The set of generators (Plant A, Plant B).
- $L \in \{1, 2\}$: The set of loads.

Parameters:

- C_g : Bid price of generator g (\$/MWh).
- P_g^{max} : Maximum capacity of generator g (MW).
- D_l : Demand of load l (MW). Total Demand $D_{total} = \sum D_l$.

Variables:

- $x_g \geq 0$: Power generated by plant g (MW).

Objective Function: Minimize Total Cost:

$$\min Z = \sum_g C_g \cdot x_g$$

Constraints:

1. **Power Balance (Demand Constraint):** The total generation must equal the total load. The **dual variable** (shadow price) of this constraint is the **Clearing Price**.

$$\sum_g x_g = \sum_l D_l \quad [\lambda]$$

2. **Capacity Constraints:**

$$x_g \leq P_g^{max} \quad \forall g$$

```

import numpy as np
from scipy.optimize import linprog

def solve_auction(scenario_name, cap_A, cap_B):
    # 1. Setup Data
    # Cost vector c = [Price_A, Price_B]
    c = [20, 80]

    # Equality Constraint: 1*A + 1*B = 100 (Demand)
    A_eq = [[1, 1]]
    b_eq = [100]

    # Bounds: 0 <= A <= cap_A, 0 <= B <= cap_B
    bounds = [(0, cap_A), (0, cap_B)]

    # 2. Solve LP
    # method='highs' is recommended for modern scipy versions
    res = linprog(c, A_eq=A_eq, b_eq=b_eq, bounds=bounds, method='highs')

    # 3. Extract Results
    gen_A = res.x[0]
    gen_B = res.x[1]

    # The Clearing Price is the dual value (shadow price) of the equality constraint
    # Note: Scipy's 'highs' method stores equality duals in res.dual
    # but strictly it's often easier to inspect the logic or use t
    if hasattr(res, 'eqlin'):
        # Negative sign convention sometimes used in optimization
        clearing_price = abs(res.eqlin.marginals[0])
    else:
        # Fallback for older scipy versions
        clearing_price = abs(res.con[0]) if res.con else 0

    # Output
    print(f"--- {scenario_name} ---")
    print(f"Capacity A: {cap_A} MW | Capacity B: {cap_B} MW")
    print(f"Dispatch: Plant A = {gen_A:.1f} MW, Plant B = {gen_B:.1f} MW")
    print(f"Total Cost: ${res.fun:.2f}")
    print(f"Clearing Price (Shadow Price): ${clearing_price:.2f}/MWh")
    print("-" * 30)

    # Run Scenario 1: Plant A limited to 50 MW
    solve_auction("Scenario 1 (Scarcity)", 50, 100)

    # Run Scenario 2: Plant A has full 100 MW (plus epsilon to resolve ties)
    solve_auction("Scenario 2 (Abundance)", 100.001, 100)

```



Would you like to see how this scenario changes if Plant A bids its full 100 MW capacity?

1. The New Setup

- **Total Market Demand:** 100 MW (Load 1 + Load 2)
- **Plant A (Cheap):** Offers 100 MW at \$20/MWh
- **Plant B (Expensive):** Offers 100 MW at \$80/MWh

2. The New Merit Order

The System Operator ranks the bids again. This time, the "cheap" stack is much larger.

Rank	Generator	Quantity Offered	Bid Price	Cumulative Supply
1	Plant A	100 MW	\$20	100 MW
2	Plant B	100 MW	\$80	200 MW

3. Clearing the Market

The operator fills the 100 MW demand starting with the cheapest power.

1. **First 100 MW:** The operator takes the full 100 MW from **Plant A** because it is the cheapest (\$20).

- *Demand met: 100 MW.*
- *Demand remaining: 0 MW.*

The Market Clearing Price: The auction stops at **Plant A**. Since Plant A is the last unit needed to satisfy the demand (the Marginal Unit), and no expensive power from Plant B was needed, the price is set by Plant A's bid.

$$\text{New Clearing Price} = \$20/\text{MWh}$$

(Note: *The price collapsed from \$80 down to \$20 simply because the expensive plant was not needed.*)

4. The Result: A Win for Consumers, A Loss for Plant A

This creates a fascinating paradox often seen in energy markets. By supplying more power, Plant A actually made less money than in the first example.

The Financial Comparison

Metric	Scenario 1 (Plant A restricted)	Scenario 2 (Plant A full capacity)
Clearing Price	\$80 / MWh	\$20 / MWh
Plant A Generation	50 MW	100 MW
Plant A Revenue	\$4,000	\$2,000
Plant B Revenue	\$4,000	\$0
Total Cost to Consumers	\$8,000	\$2,000

