

AAE 706

Optimization and Equilibrium

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Suppose that irregardless of relative prices, Suzy always has one soda before and one soda after eating an ice cream. What utility function is consistent with these choices? Write down demand functions which could extrapolate her optimal choices to any expenditure (m) and prices (p_1 and p_2).

Perfect complement preferences have the form:

$$U(x_1, x_2) = \min\left(\frac{x_1}{a_1}, \frac{x_2}{a_2}\right)$$

in which the ratio $\frac{a_1}{a_2}$ determines the ratio in which goods 1 and 2 are consumed. In the present example, we have:

$$U(x_1, x_2) = \min\left(x_1, \frac{x_2}{2}\right)$$

and demand functions given by:

$$x_1 = \frac{Y}{p_1 + 2p_2}$$

and

$$x_2 = 2 \frac{Y}{p_1 + 2p_2}$$



When Joe gets to the bar, he always asks about the price of peanuts and the price of beer. If two beers cost less than one bag of peanuts, he spends all of his money on beer. Otherwise he buys peanuts. What utility function is consistent with these choices? Write down demand functions which could extrapolate her optimal choices to any expenditure (m) and prices (p_1 and p_2).

General *perfect substitutes preferences* have the form:

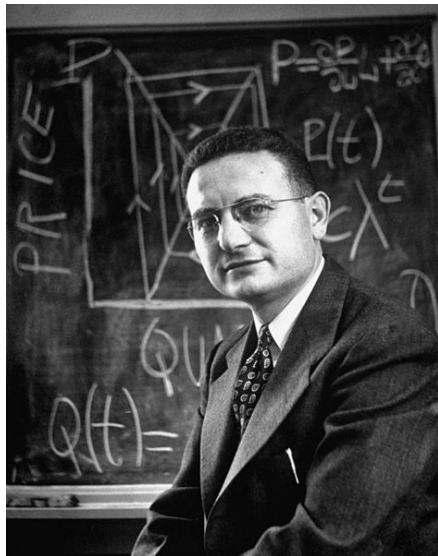
$$U(x_1, x_2) = a_1 x_1 + a_2 x_2$$

in which the ratio $\frac{a_1}{a_2}$ represents the marginal rate of substitution of good 1 for good 2. The demand functions for these preferences are given by:

$$x_1 = \begin{cases} 0 & \text{when } \frac{p_1}{p_2} > \frac{a_1}{a_2} \\ \frac{M}{p_1} & \text{otherwise} \end{cases}$$

$$x_2 = \begin{cases} 0 & \text{when } \frac{p_1}{p_2} < \frac{a_1}{a_2} \\ \frac{M}{p_2} & \text{otherwise} \end{cases}$$

Paul Samuelson in 1950



- University of Chicago (B.A.) – Harvard University (Ph.D.),
- Enrolled college at age 16
- Full professor at age 32
- First American to win the Nobel Memorial Prize in Economic Sciences: “[Samuelson] has done more than any other contemporary economist to raise the level of scientific analysis in economic theory.”
- Recruited numerous Nobel laureates at MIT: Robert M. Solow, Paul Krugman, Franco Modigliani, Robert C. Merton and Joseph E. Stiglitz.

A Theory which is both True and Nontrivial



Stanislaw Ulam once challenged Samuelson to name one theory in all of the social sciences which is both true and nontrivial.

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Several years later, Samuelson responded with David Ricardo's theory of comparative advantage:

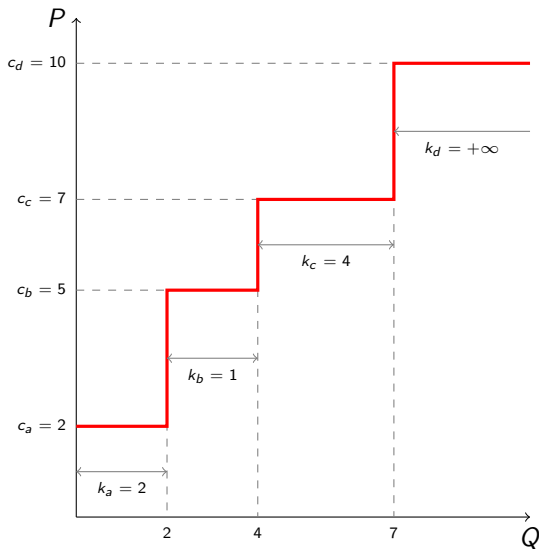
That it is logically true need not be argued before a mathematician; that is not trivial is attested by the thousands of important and intelligent men who have never been able to grasp the doctrine for themselves or to believe it after it was explained to them.

When there are a discrete set of production technologies, each characterized by a marginal cost and a capacity, the supply curve becomes a step function corresponding to the sorted sequence of plant capacities.

Consider a market in which the commodity is supply by the following four technologies:

	c_j	k_j
a	2	2
b	5	2
c	7	4
d	10	∞

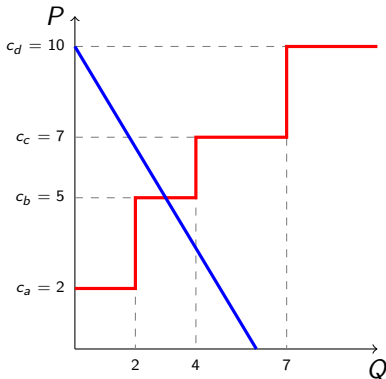
Activity Analysis Supply Curve



Market Equilibrium with Activity Analysis



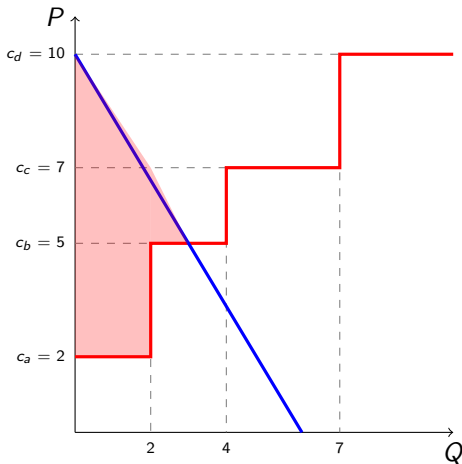
Consider a market equilibrium when there are multiple discrete supply technologies. As in the conventional continuous Marshallian model, the equilibrium price and quantity is defined by the intersection of the supply and demand schedules:



Market Equilibrium and Social Surplus



A convenient property of the competitive market allocation is that it *maximizes* social surplus, as illustrated in this figure:



Constrained Optimization Approach



Let $Q_t \geq 0$ denote output from technology t , P denote the equilibrium price, PS and CS denote producer and consumer surplus. The market equilibrium then solves:

$$\max PS + CS$$

subject to:

- Market supply equals technology output:

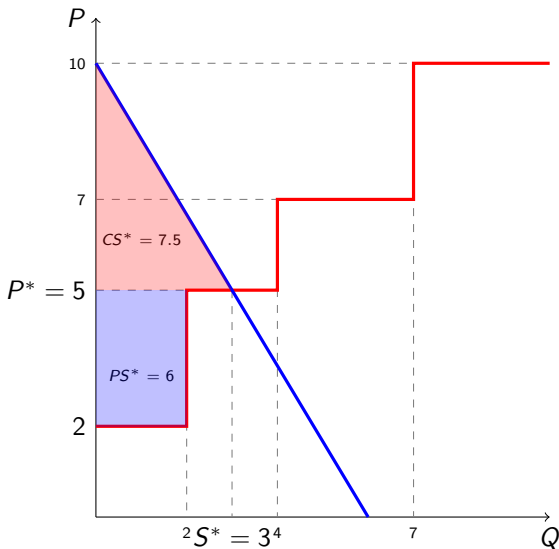
$$S = \sum_t Q_t$$

- Market equilibrium price is on the demand curve:

$$P = 10 - \frac{5}{6}S$$

- Producer surplus is the area below the market price and above the cost of production: $PS = \sum_t (P - c_t) Q_t$
- Consumer surplus is the area under the demand curve: $CS = \frac{(10-P)}{2} S$

Geometric Interpretation of the Equilibrium



```
$title    surplus maximization and market equilibrium

set       t /a,b,c,d/;

table     tech      Technology
          cost      cap
a         2         2
b         5         2
c         7         4
d         10        inf;

parameter      c(t)      Cost by technology;
c(t) = tech(t,"cost");
```



```
nonnegative variables    P,PS,CS,s,Q(t);
free variable            obj;
equations                price, supply, psurplus, csurplus, objective;

price..                  P =e= 10 - S*10/6;

supply..                 S =e= sum(t, Q(t));

psurplus..               PS =e= sum(t, (P-c(t))* Q(t));

csurplus..               CS =e= (10 - P)*S/2;

objective..              OBJ =e= CS + PS;

Q.UP(t) = tech(t,"cap");

model equil /all/;
solve equil using nlp maximizing OBJ;
```

	LOWER	LEVEL	UPPER	MARGINAL
---- VAR P	.	5.0000	+INF	.
---- VAR PS	.	6.0000	+INF	.
---- VAR CS	.	7.5000	+INF	.
---- VAR Q				
a	.	2.0000	2.0000	3.0000
b	.	1.0000	2.0000	EPS
c	.	.	4.0000	-2.0000
d	.	.	+INF	-5.0000
---- VAR obj	-INF	13.5000	+INF	.

- Sets
 - s Load segments
 - j Generating units, e.g. existing capacity, new investment options
 - i Demand categories, e.g. residential, commercial, industrial
 - f Fuel types, e.g. hard coal, soft coal, natural gas, uranium

h_s Segment durations, hours

$\bar{p}_s, \bar{D}_{is}, \epsilon_{is}$ Demand characteristics as might be represented by representative price-quantity pairs and elasticities of demand (price expressed in € per KW, demand in KW and elasticity is dimensionless)

- ϕ_{fj} Heat rates describing input fuel requirements per unit generation (PJ per KWH)
- \bar{K}_j Capacities of existing generating units, TW
- c_f Fuel costs (€ per PJ)
- α_{js} Average availability factor for generating units, reflecting need for repair and intermittency of renewable energy sources (dimensionless)
- r_j^K Rental price of *new* generating capacity, (€ per KW per year), typically computed on the basis of capital cost, depreciation rate, capital cost and fixed maintenance and operating costs:

$$r_j^K = \begin{cases} p_j^K(r + \delta) + c_j^M & \text{New plants} \\ c_j^M & \text{Extant plants} \end{cases}$$

- p_j^M Variable maintenance and operating costs, (€ per KWH)

- Primal Variables : quantities
 - X_{js} Generation and dispatch
 - K_j Generating utilization (extant and new vintage)
- Dual Variables : prices
 - p_s Wholesale prices by load segment
 - π_{js} Profit margins
 - μ_j Shadow price on installed (extant) capacity

- Aggregate demand:

$$D_s = \sum_i \bar{D}_{is} (1 - |\epsilon_{is}|(p_s/\bar{p}_s - 1)) \quad \perp \quad D_s$$

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- Market clearance:

$$D_s = \sum_j X_{js} \quad \perp \quad p_s$$

- Aggregate demand:

$$D_s = \sum_i \bar{D}_{is} (1 - |\epsilon_{is}|(p_s/\bar{p}_s - 1)) \quad \perp \quad D_s$$

- Market clearance:

$$D_s = \sum_j X_{js} \quad \perp \quad p_s$$

- Feasibility of generation:

$$\alpha_{js} K_j \geq X_{js} \geq 0 \quad \perp \quad \pi_{js} \geq 0$$

- Aggregate demand:

$$D_s = \sum_i \bar{D}_{is} (1 - |\epsilon_{is}|(p_s/\bar{p}_s - 1)) \quad \perp \quad D_s$$

- Market clearance:

$$D_s = \sum_j X_{js} \quad \perp \quad p_s$$

- Feasibility of generation:

$$\alpha_{js} K_j \geq X_{js} \geq 0 \quad \perp \quad \pi_{js} \geq 0$$

- Capacity:

$$\bar{K}_j \geq K_j \quad \perp \quad \mu_j \geq 0$$

- Profitability – arbitrage in dispatch:

$$\pi_{js} + p_j^M + \sum_f c_f \phi_{fj} \geq p_s \quad \perp \quad X_{js} \geq 0$$

- Profitability – arbitrage in dispatch:

$$\pi_{js} + p_j^M + \sum_f c_f \phi_{fj} \geq p_s \quad \perp \quad X_{js} \geq 0$$

- Profitability – arbitrage in investment:

$$r_j^K + \mu_j \geq \sum_s h_s \alpha_{js} \pi_{js} \quad \perp \quad K_j \geq 0$$

Integrability: Equilibrium Allocation = Optimal Allocation

$$\begin{aligned} \max \quad & \sum_{i,s} \bar{p}_s D_{is} \left(1 + 1/|\epsilon_{is}| \left(1 - \frac{D_{is}}{2D_{is}} \right) \right) \\ & - \sum_{sj} X_{js} h_s \left(\sum_f c_f \phi_{jf} + p_j^M \right) - \sum_j K_j r_j^K \end{aligned}$$

subject to:

$$\sum_{is} D_{is} = \sum_j X_{js}$$

$$\alpha_{js} K_j \geq X_{js} \geq 0$$

$$\bar{K}_j \geq K_j$$

$$K_j \geq 0$$

```
variables      TOTCOST Objective function (dispatch cost);

nonnegative
variables      Y(j,s) Dispatch;

equations      costdef, demand;

costdef..      TOTCOST =e= sum((j,s), mc(j)*Y(j,s));

demand(s)..    sum(j, Y(j,s)) =e= qref(s);

model mincost /costdef, demand/;

Y.UP(j,s) = supply(j,"cap");

solve mincost minimizing TOTCOST using LP;
```

```
parameter          pref(s)          Reference price,
                   dref(i,s)        Reference demand;

pref(s) = demand.m(s);
dref(i,s) = qref(s) * load(s,i);

parameter          epsilon(i)        Elasticity of demand /rsd 0.1,  com 0.2,  ind 0.5/;

nonnegative
variables          D(i,s)  Aggregate demand,
                   PI(j,s) Shadow price on capacity,
                   P(s)    Market price;

equations aggdemand, supplydemand, profit, capacity;

aggdemand(i,s)..   D(i,s) =e= dref(i,s) * (1 - epsilon(i)*(P(s)/pref(s)-1));

supplydemand(s)..  sum(j,Y(j,s)) =e= sum(i,D(i,s));

profit(j,s)..      mc(j) + PI(j,s) =g= P(s);

capacity(j,s)..    cap(j) =g= Y(j,s);

model equil /aggdemand.D, supplydemand.P, profit.Y, capacity.PI/;

Y.UP(j,s) = inf;  P.L(s) = pref(s);  D.L(i,s) = dref(i,s);  PI.L(j,s) = -Y.M(j,s);

equil.iterlim = 0;
SOLVE equil USING mcp;
```

Electricity Dispatch GAMS Code: Equivalent QP Model

```
variable      SURPLUS      Sum of consumer and producer surplus
              K(j)         Capacity of technology j;

equation      surplusdef    Defines the surplus;

surplusdef..  SURPLUS =e= sum((i,s), pref(s)*D(i,s) *
                          (1 + 1/epsilon(i) * (1 - D(i,s)/(2*dref(i,s)))))
                          - TOTCOST;

MODEL SAMUELSON /surplusdef, supplydemand, costdef, capacity/;

K.FX(j) = supply(j,"cap");
Y.UP(j,s) = +inf;

SOLVE samuelson USING nlp MAXIMIZING surplus;
```


The *isoelastic* demand function is an alternative functional form:

$$d(p) = ap^b$$

- 1 Derive values of a and b which produce a demand function which is locally consistent with the following *linear* demand curve at $p = \bar{p}$:

$$d(p) = \bar{d} \left(1 - |\epsilon| \left(\frac{p}{\bar{p}} - 1 \right) \right)$$

- 2 Formulate a representation of the isoelastic demand based on \bar{q} , \bar{p} and ϵ rather than a and b .
- 3 Produce MCP and NLP models with iso-elastic demand, and demonstrate that these are calibrated.
- 4 Impose a supply shock (a phase out of coal generation) and compare results from the linear and isoelastic models.

- i Supply nodes
- j Demand nodes
- c_{ij} Unit shipment costs
- μ_i Unit (marginal) production cost
- \bar{S}_i Supply limit (upper bound)
- \bar{D}_j Demand quantity

$$\min \sum_i \mu_i S_i + \sum_{i,j} c_{ij} X_{ij}$$

subject to:

$$\begin{aligned} S_i &\geq \sum_j X_{ij} \\ \sum_i X_{ij} &\geq D_j \end{aligned}$$

$$D_j = \bar{D}_j, \quad S_i \leq \bar{S}_i$$

```
$title  A Calibrated Spatial Price Equilibrium Model
```

```
$ontext
```

We first formulate a linear programming model which minimizes the cost of production and distribution on a transportation network with supply nodes and demand nodes. Using the primal and dual values from the LP model we calibrate an economic equilibrium model with price elastic demand and supply for which the reference equilibrium corresponds precisely to the LP optimum.

```
$offtext
```

```
*      Generate a random instance of the problem:
```

```
set      i      Supply nodes /1*5/  
        j      Demand nodes /1*5/;
```

```
parameter      d0(j)  Demands  
               s0(i)  Supply  
               mu(i)  Marginal cost of production,  
               c(i,j) Transport cost;
```

```
c(i,j) = uniform(0,1);  
d0(j) = round(uniform(1,100));  
s0(i) = round(uniform(1,200));  
mu(i) = uniform(0.5,1.5);
```

```
*      Here I illustrate the lazy way to declare variables.  When
*      a variable is declared with no arguments, the dimensionality
*      is inferred at the first use and the domains are assumed
*      to be the universe, e.g. X(*,*).
```

```
*      The disadvantage of this approach is that domain errors
*      may be undetected and difficult to trace. It is a good idea
*      to use explicit domain wherever possible:
```

```
nonnegative variables    X,S,D;
```

```
free variable    TOTCOST          Objective function;
```

```
equations        objdef, supply, demand;
```

```
objdef..         TOTCOST =e= sum((i,j), c(i,j) * X(i,j)) + sum(i, mu(i)*S(i));
```

```
*      Orient both equations as >= so that the Lagrange multipliers
*      are non-negative:
```

```
supply(i)..      S(i) =g= sum(j, X(i,j));
```

```
demand(j)..      sum(i, X(i,j)) =g= D(j);
```

```
model transport /all/;
```

```
*      Fix demand and place an upper bound on supply in order
*      that the marginal cost of supply is included in the
*      shadow prices at the equilibrium point:
```

```
S.UP(i) = s0(i);  D.FX(j) = d0(j);
```

```
solve transport using LP MINIMIZING TOTCOST;
```

Formulated as a capacity-constrained supply with constant marginal cost, the shadow prices at supply and demand nodes reflect both the production and transportation costs:

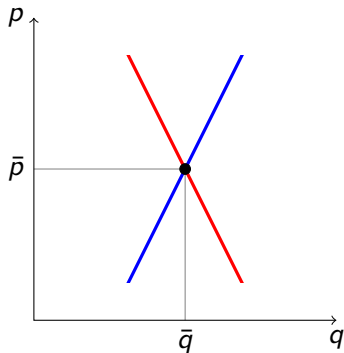
---- EQU supply

	LOWER	LEVEL	UPPER	MARGINAL
1	.	.	+INF	1.3821
2	.	.	+INF	1.3298
3	.	.	+INF	1.2227
4	.	.	+INF	1.1282
5	.	.	+INF	1.2468

---- EQU demand

	LOWER	LEVEL	UPPER	MARGINAL
1	.	.	+INF	1.5539
2	.	.	+INF	1.2878
3	.	.	+INF	1.3783
4	.	.	+INF	1.3969
5	.	.	+INF	1.3534

Calibrated Supply and Demand Functions



Given the following *data*:

\bar{q} Reference quantity supplied (and demanded)

\bar{p} Reference demand price

$\bar{\mu}$ Reference supply price

ϵ Magnitude of the price elasticity of demand

η Magnitude of the price elasticity of supply

We can write the demand and supply functions as:

$$d(p) = \bar{d} \left(1 - \epsilon \left(\frac{p}{\bar{p}} - 1 \right) \right)$$

and

$$s(\mu) = \bar{s} \left(1 + \eta \left(\frac{\mu}{\bar{\mu}} - 1 \right) \right)$$

①

$$\frac{d}{dQ} \int_{q=0}^Q p(q) dq = p(Q)$$

② The first order conditions for

$$\max \sum_i f_i(S_i) + \sum_j g_j(D_j)$$

s.t.

$$\begin{array}{rcl} S_i & \geq & \sum_j X_{ij} \quad \perp \mu_i \\ \sum_i X_{ij} & \geq & D_j \quad \perp p_j \end{array}$$

are

$$\frac{df_i(S_i)}{dS_i} = -\mu_i$$

and

$$\frac{dg_j(D_j)}{dD_j} = p_j$$

The calibrated inverse demand function corresponding to $D_j(p_j)$ is

$$p_j(q) = \bar{p}_j \left(1 + (1 - q/\bar{D}_j) / \epsilon_j \right)$$

and the calibrated inverse supply function corresponding to $S_i(\mu_i)$ is

$$\mu_i(q) = \bar{\mu}_i \left(1 + (q/\bar{S}_i - 1) / \eta_i \right)$$

Integrating, we have consumer surplus

$$CS_j(D_j) = \int^{D_j} p_j(q) dq = \bar{p}_j D_j \left(1 + \left(1 - \frac{D_j}{2\bar{D}_j} \right) / \epsilon_j \right)$$

and total cost

$$TC_i(S_i) = \int^{S_i} \mu_i(q) dq = \bar{\mu}_i S_i \left(1 + \left(\frac{S_i}{2\bar{S}_i} - 1 \right) / \eta_i \right)$$

$$\max \sum_j \underbrace{\int_0^{D_j} p_j(q) dq}_{CS_j(D_j)} - \sum_i \underbrace{\int_0^{S_i} \mu_i(q) dq}_{TC_i(S_i)} - \sum_{ij} c_{ij} X_{ij}$$

s.t.

$$S_i \geq \sum_j X_{ij} \quad \perp \mu_i$$

$$\sum_i X_{ij} \geq D_j \quad \perp p_j$$

$$X_{ij} \geq 0$$

Price-Responsive Demand (QCP Formulation)



```
*      Extract the solution with fixed demand as a reference equilibrium:

parameter          muref(i)          Reference marginal cost
                   pref(j)          Reference demand price
                   sref(i)          Reference supply
                   dref(j)          Reference demand
                   epsilon(j)       Demand elasticity at the reference point;

muref(i) = supply.m(i); pref(j) = demand.m(j); sref(i) = S.L(i); dref(j) = D.L(j);

epsilon(j) = uniform(0.5, 2);

free variable      SURPLUS          Social surplus;

equation           csurplus          Social surplus with horizontal supply curves (Cs);

csurplus..         SURPLUS =e= -sum((i,j), c(i,j) * X(i,j)) - sum(i, mu(i)*S(i))
                   + sum(j, pref(j) * D(j) * (1 + (1-0.5*D(j)/dref(j)) / epsilon(j)));

model elasticdemand /supply, demand, csurplus/;

*      Remove upper and lower bounds on demand:

D.LO(j) = 0; D.UP(j) = +inf;

solve elasticdemand using QCP maximizing SURPLUS;
```

Formulated as a maximization problem, Lagrange multipliers on the supply and demand markets change sign, but they have identical magnitude as compared with the LP solution. This implies that we have “replicated the benchmark equilibrium”, having removed upper and lower bounds on demand but introduced the consumer surplus measure which results in no change in prices or quantities.

---- EQU supply

	LOWER	LEVEL	UPPER	MARGINAL
1	.	.	+INF	-1.3821
2	.	.	+INF	-1.3298
3	.	.	+INF	-1.2227
4	.	.	+INF	-1.1282
5	.	.	+INF	-1.2468

---- EQU demand

	LOWER	LEVEL	UPPER	MARGINAL
1	.	.	+INF	-1.5539
2	.	.	+INF	-1.2878
3	.	.	+INF	-1.3783
4	.	.	+INF	-1.3969
5	.	.	+INF	-1.3534

Price-Responsive Supply and Demand (QCP Formulation)

```
parameter          eta(i)  Price elasticity of supply from node i;

eta(i) = uniform(0.5, 2);

equation           ssurplus      Social surplus with price elastic supply;

ssurplus..         SURPLUS =e= -sum((i,j), c(i,j) * X(i,j))
                    + sum(j,  pref(j) * D(j) * (1 + (1-0.5*D(j)/dref(j)) / epsilon(j)))
                    - sum(i,  muref(i) * S(i) * (1 + (0.5*S(i)/sref(i)-1)/eta(i)));

model equilibrium /supply, demand, ssurplus/;

*      Remove the upper bound so as to accommodate price-elasticity:

S.UP(i) = +inf;

solve equilibrium using QCP maximizing SURPLUS;
```

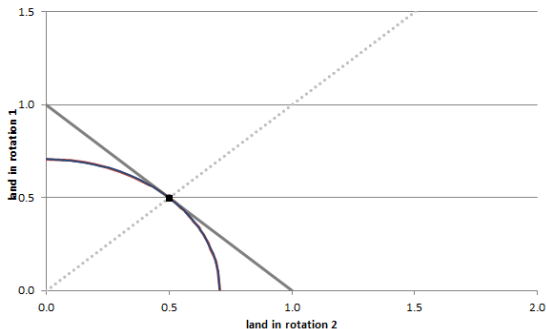
- The Marshallian model fails to account for the *interconnectedness* of supply for goods produced jointly.
- Consider the supply of agricultural land to produce crops. The supply depends on the choice of *crop rotation*
- An increase in supply of one crop (i) may imply increases or decreases in the supply of another crop (j), depending on whether i and j are complements or substitutes in production.
- Aggregate land area varies in the CET model. Extensions of the model are required to incorporate a fixed land supply (some other cost determines decisions over rotations).

- In the American midwest soybeans are alternated with corn because soybeans have nodules on their roots that host bacteria which fix atmospheric nitrogen. Thus, soybeans require less nitrogen to be applied to the field. As a result, an increase in the supply of corn might then increase the supply of soybeans. In short, these are *complementary joint products*.
- In REAP production response to changes in crop prices involves land allocation across alternative *rotations*. The supply of crops is then implicitly determined by the allocation of land to rotations and acreage and crop yields associated with those rotations.
- Consider the allocation of aggregate land (\mathcal{L}) to land farmed with one of two rotations r_1 and r_2 . For example, technique r_1 may represent a crop rotation of corn and soybeans on alternate years. r_2 might represent a rotation in which corn is planted for two years and soybeans for one.

Joint Products: Application to Land



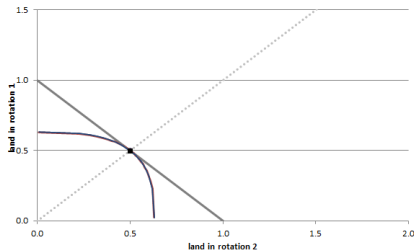
Land is heterogeneous. Allocation of aggregate land to land under cultivation implies that as land is moved among rotations, the aggregate efficiency changes. Specialization in a more profitable rotation imposes a cost denominated in the ratio of the cultivated land to the aggregate supply.



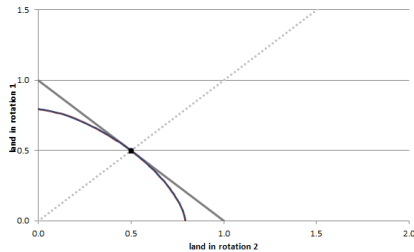
CET Model: Alternative Values of η



- Low elasticity: $\eta = 0.5$



- High elasticity: $\eta = 2$



Project Independence was an initiative announced by U.S. President Richard Nixon on November 7, 1973, in reaction to the OPEC oil embargo and the resulting 1973 oil crisis. Recalling the Manhattan Project, the stated goal of Project Independence was to achieve energy self-sufficiency for the United States by 1980 through a national commitment to energy conservation and development of alternative sources of energy. Nixon declared that American science, technology and industry could free America from dependence on imported oil (energy independence).



Some of the important initiatives to emerge from Project Independence included lowering highway speeds to 55 mph (89 km/h), converting oil power plants to coal, completion of the Trans-Alaska Pipeline System and diverting federal funds from highway construction to mass transit.



- The PIES initiative was notable for its ambition, the United States has had limited success with energy independence. In 1976 the General Accounting Office reviewed the complex PIES forecasting and analysis system which had been developed by the Federal Energy Administration (FEA) used in preparing the November 1974 Project Independence report.
- The key recommendations of the GAO review were that FEA should add to its plan (i) an analysis of problems resulting from the static nature of the system and the procedures which can be used to alleviate them; (ii) an analysis of the limitations in the environmental impact analysis and the procedures which can be applied to correct them; and (iii) a comprehensive, well-documented verification, validation, and sensitivity testing effort. In spite of these limitations, the analysis represented by the PIES model was well ahead of its time.

The Energy System

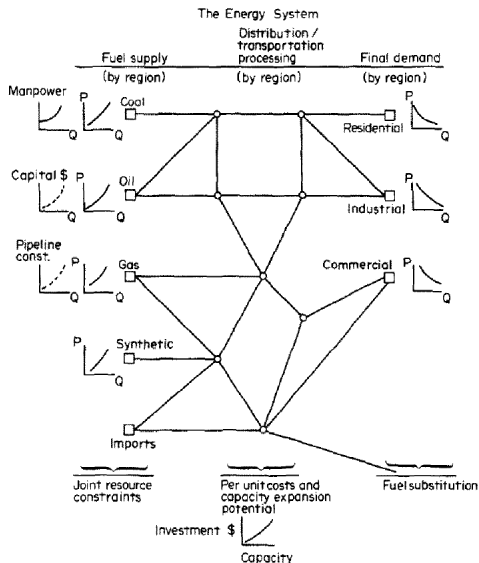


Fig. 1. The energy system.

A Framework for Analysis

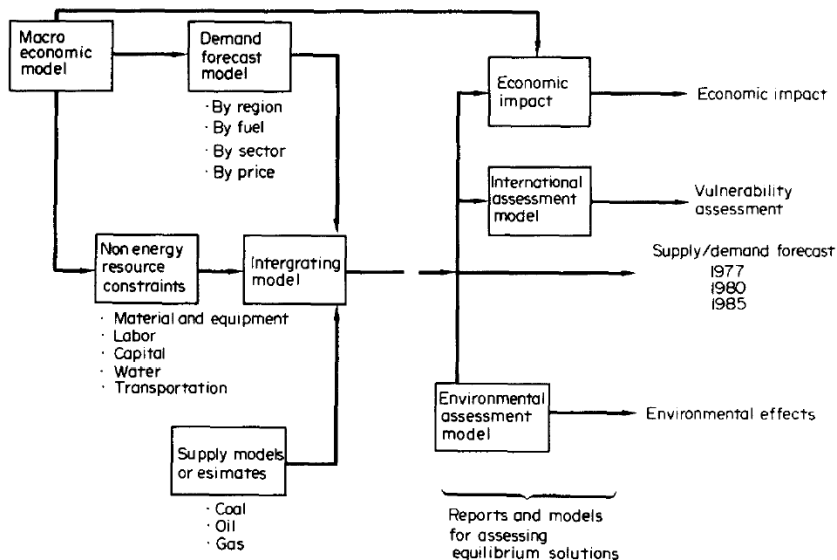


Fig. 2. The FEA project independence evaluation system.

It is well-known that modelers and policy analysts gain access to policymaking arenas based on what they know. Therefore, critics of models are quick to employ various types of technical standards when evaluating policy models in order to assess validity and reliability of claims to knowledge. This article argues that, in the effort to make models better, overreliance on technical standards misses the important political and policy reasons to model: models call attention to the modelers and to their advice about important policy problems of the day. In this sense, models are used as symbols, as claims to authority, whether or not the underlying knowledge is technically up to snuff. Drawing on the experience of energy policy models, this article explores the problem of models as knowledge versus models as symbols and it examines the muddle that conflicts between them produce. (PB Hammond, *Policy Sciences* February 1984, Volume 16, Issue 3, pp 227-243)

William Hogan, “Energy Policy Models for Project Independence”,
Computers and Operations Research Vol 2, pp 251–271, 1975.

```
set      j      Consumption regions /j1, j2/

        i      Coal supply regions /i1, i2/

        k      Oil supply regions  /k1, k2/

        r      Refineries          /r1, r2/

        c      Increments for coal /L, M, H/,

        o      Increments for oil  /L, H/,

        p      Energy products      /Light, Heavy, Coal/,

g(p)     Grades of refined oil / Light, Heavy/;
```

Table 1. Resource requirements for production levels

	Production (ton/D)	Minimum price/ton(\$)	New capital/ton	Steel/ton
Coal region 1	0-300	5	1	1
	300-600	6	5	2
	600-1000	8	10	3
Coal region 2	0-200	4	1	1
	200-500	5	5	4
	500-1000	7	6	5

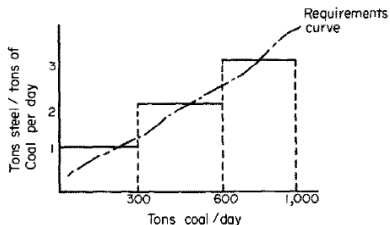


Fig. 3. Steel requirement coal region 1.

NONNEGATIVE VARIABLES

COAL(i,c)	Coal in region i at increment c,
OIL(k,o)	Oil in region k at increment o,Cent
XC(i,j)	Coal transport from i to j,
XO(k,r)	Oil transport from k to refinery r,
Z(r)	Level of operation in refinery r,
Y(p,r,j)	Oil products from r to j,

$P_C(i)$	Supply price of coal,
$P_O(k)$	Supply price of oil,
$P_{XO}(r)$	Delivered price of oil,
$P_R(g,r)$	Price of refinery outputs,
$P_D(p,j)$	Demand price of products,
P_S	Shadow price of the steel,
P_K	Shadow price of the capital;

```
oilresource(k)..      sum(o, Q0(k,o)) =g= sum(r, X0(k,r));  
crudeoil(r)..        sum(k, X0(k,r)) =g= QR(r);  
refinedoil(r,g)..    QR(r)*table5(g,r) =g= sum(j, XR(r,g,j));  
oildemand(j,g)..     sum(r, XR(r,g,j)) =g= D(g,j);  
coalsupply(i)..      sum(c, QC(i,c)) =g= sum(j, XC(i,j));  
coaldemand(j)..      sum(i, XC(i,j)) =g= D("coal",j);
```

* Transportation costs:

```
transportcost..       TCOST =e= sum((r,g,j), XR(r,g,j)*table6(r,j)) +  
                                  sum((k,r),    X0(k,r)*table4(k,r)) +  
                                  sum((i,j),    XC(i,j)*table2(i,j));
```

* Other costs:

```
othercost..           OCOST =e= sum((k,o), Q0(k,o)*table3(k,o,"c0")) +  
                                  sum((i,c), QC(i,c)*table1(i,c,"c0")) +  
                                  sum(r,        QR(r)*table5("cost",r));
```

* First use a simple integrable demand function to define consumer surplus

```
surplus..           CS =e= sum((p,j), D(p,j)*pref(p) * (1 + (1-D(p,j))/(2*qref(p))/ep
```

```
objdef..           OBJ =e= CS - TCOST - OCOST;
```



```
QC.UP(i,c) = table1(i,c,"cap");  
QO.UP(k,o) = table3(k,o,"cap");  
  
model pies /all/;  
solve pies using QCP maximizing obj;
```