

Economic decision analysis: Concepts and applications

Jeffrey M. Keisler

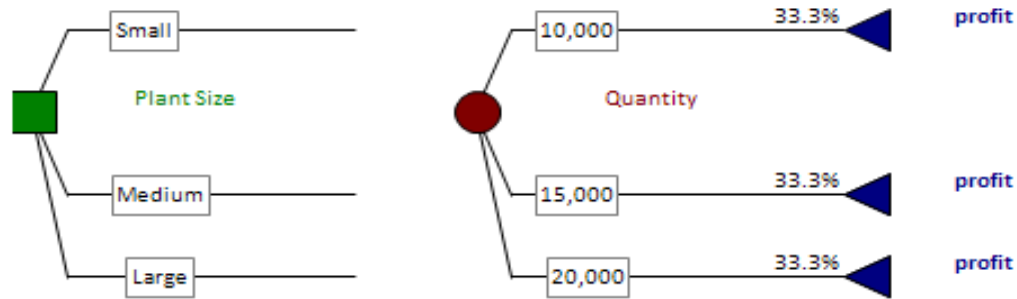
Stockholm, 23 May 2016



My background and this work

- Education in DA and Economics
- Government and industry consulting
- Portfolio DA
- The nature of modeling

Textbook DA problem: plant size vs. and sales volume

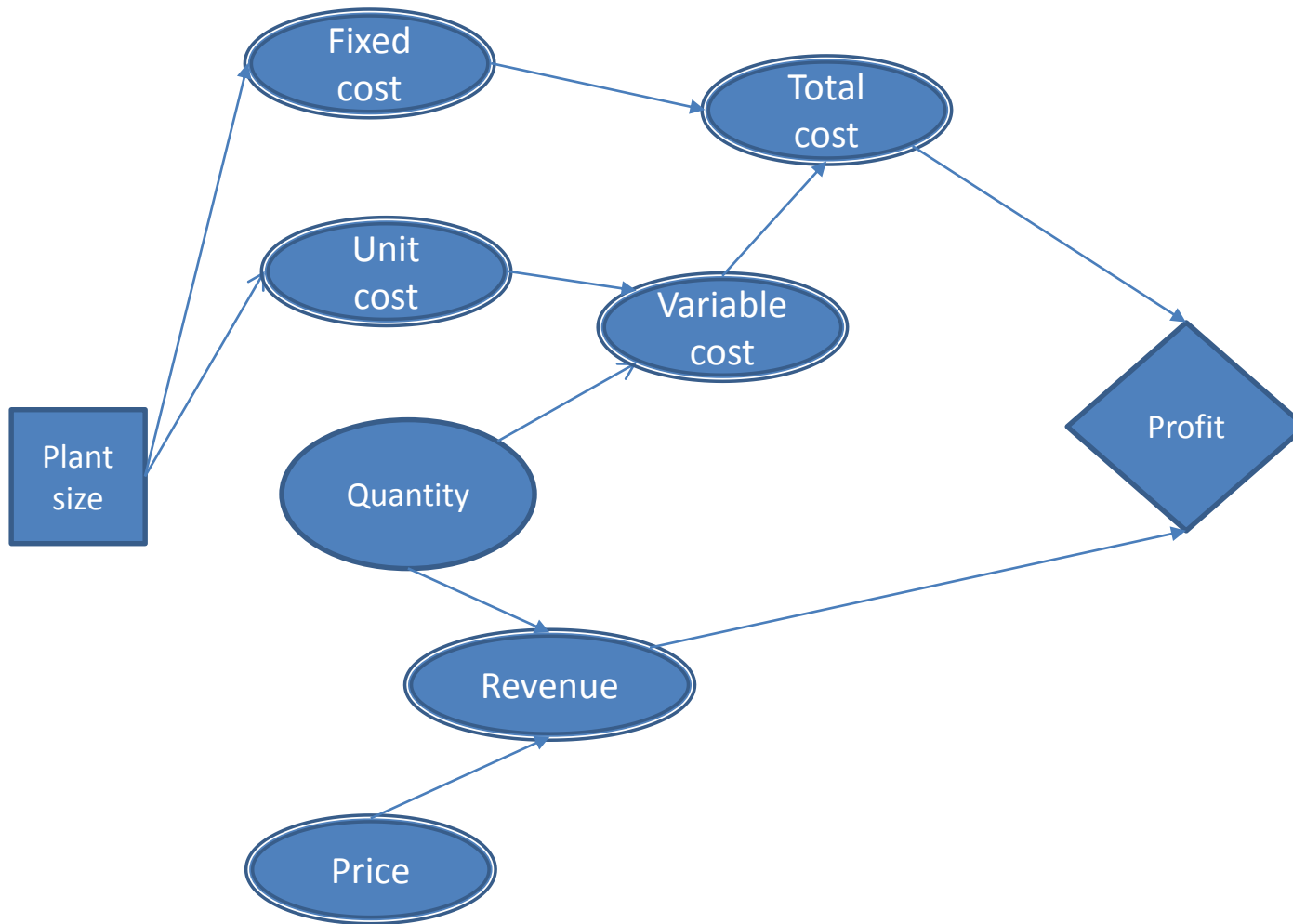


Profit =
quantity * (100 - variable cost) - fixed cost

Plant size	Fixed cost	Variable cost per unit
Small	\$75,000	\$70
Medium	\$325,000	\$50
Large	\$650,000	\$30

		TOTAL PROFIT		
Quantity	Probability	Small plant	Medium plant	Large plant
10,000	33%	\$225,000	\$175,000	\$50,000
15,000	33%	\$375,000	\$425,000	\$400,000
20,000	33%	\$525,000	\$675,000	\$750,000
Expected value		\$375,000	\$425,000	\$400,000

The influence diagram serves as a map for constructing the model

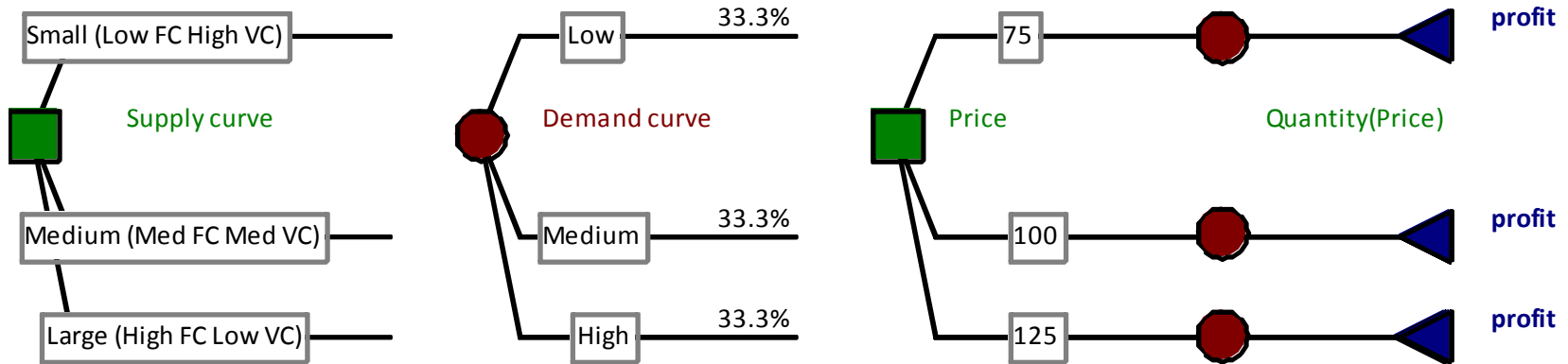


Is price a decision \square or uncertainty \circ ? What about quantity?

- “We will have the highest profit margin and the highest volume.” $\square P \square Q$
- “What we make is what we will sell.” $\circ P \square Q$
- “If our customer’s price drops, we’ll have to suffer along with them.” $\circ P \circ Q$
- “We will reduce risk *and* cost by pushing all risk to our suppliers.” $\square P \square Q$

*All real examples

Using an uncertain demand function improves the decision by revealing a hidden option

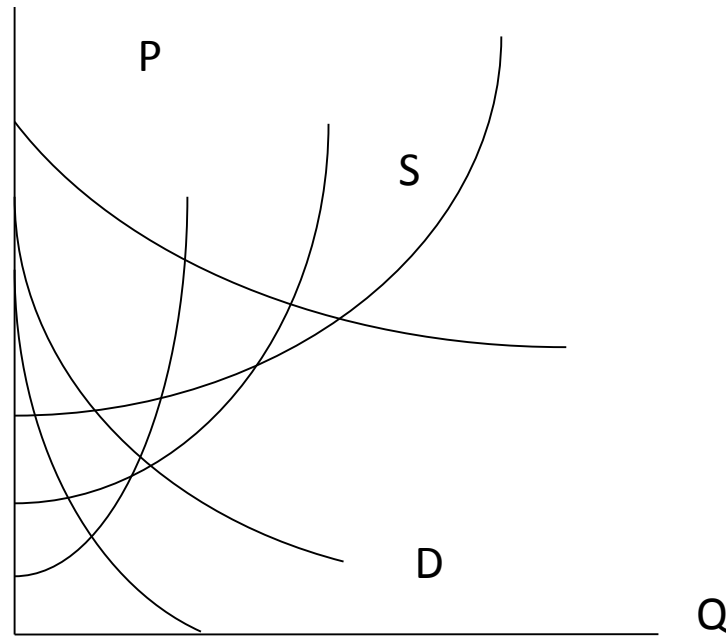
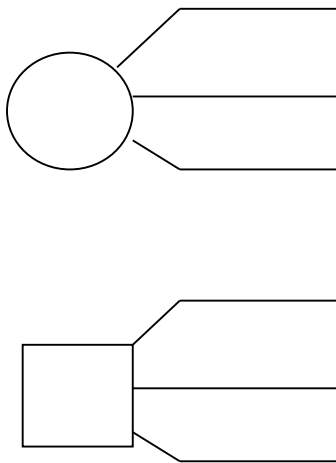


Profit =
 $\text{quantity}(\text{price}) * (\text{price} - \text{variable cost}) - \text{fixed cost}$

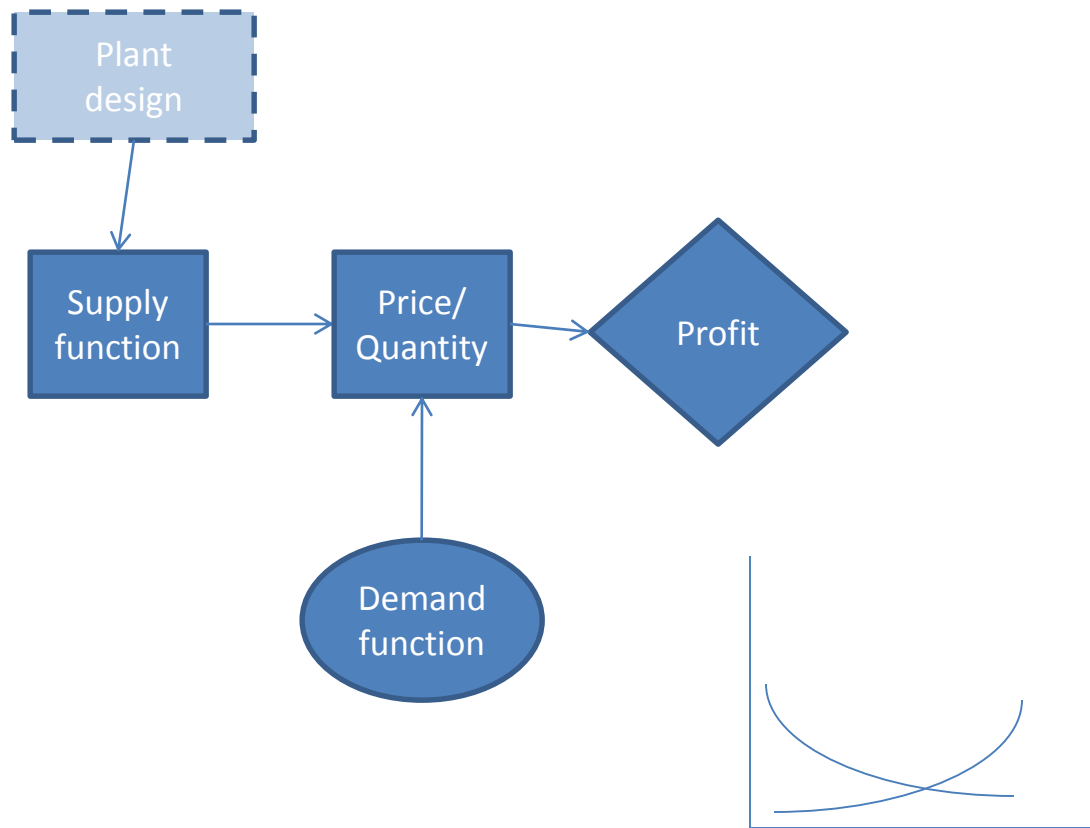
Demand function: $\text{quantity} = 200 - k_2 * \text{price}$,
 $k_2 = 150$ (high demand), 175 or 200

Profit (optimal price)	Small plant	Medium plant	Large plant
Low demand	\$225,000 (100)	\$175,000 (100)	\$50,000 (100)
Medium demand	\$475,000 (125)	\$425,000 (100 or 125)	\$400,000 (100)
High demand	\$750,000 (125)	\$800,000 (125)	\$775,000 (125)
EV	\$483,333	\$466,667	\$408,333

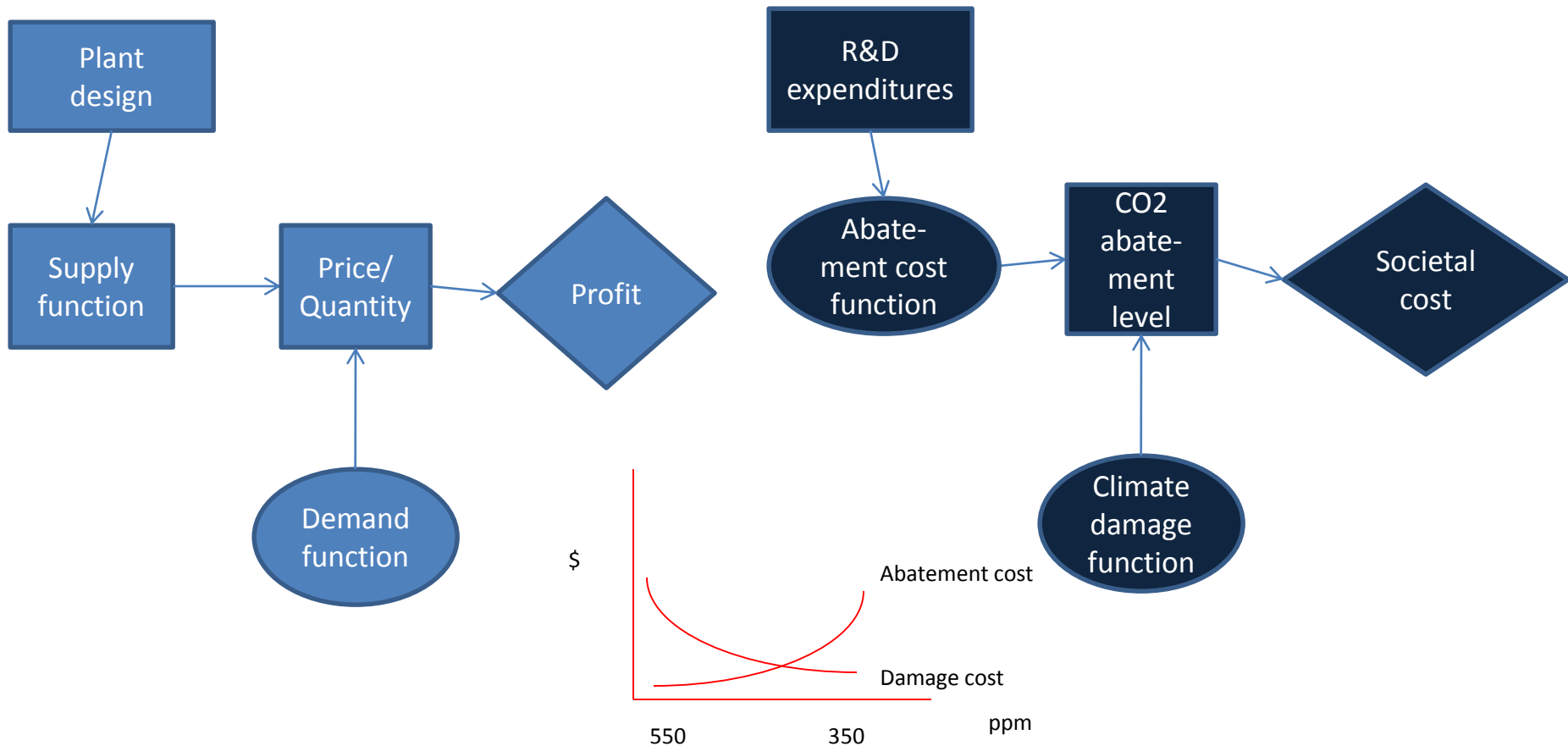
Economic derivation of price, quantity and resulting surplus for all scenarios



An influence diagram represents this problem with nodes for supply & demand functions

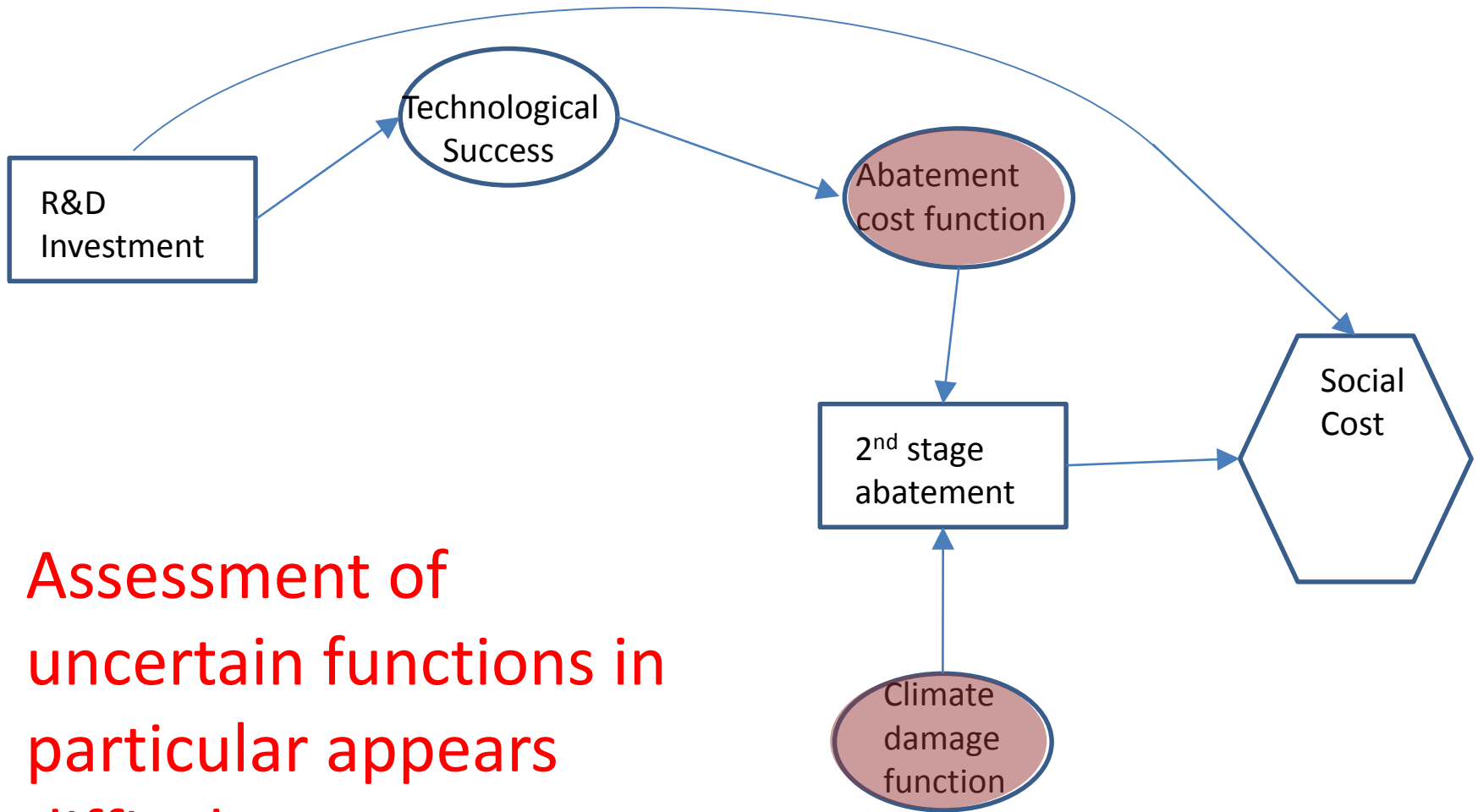


This is similar to a problem in the economics of climate change



*work with Erin Baker

Application to climate change problem

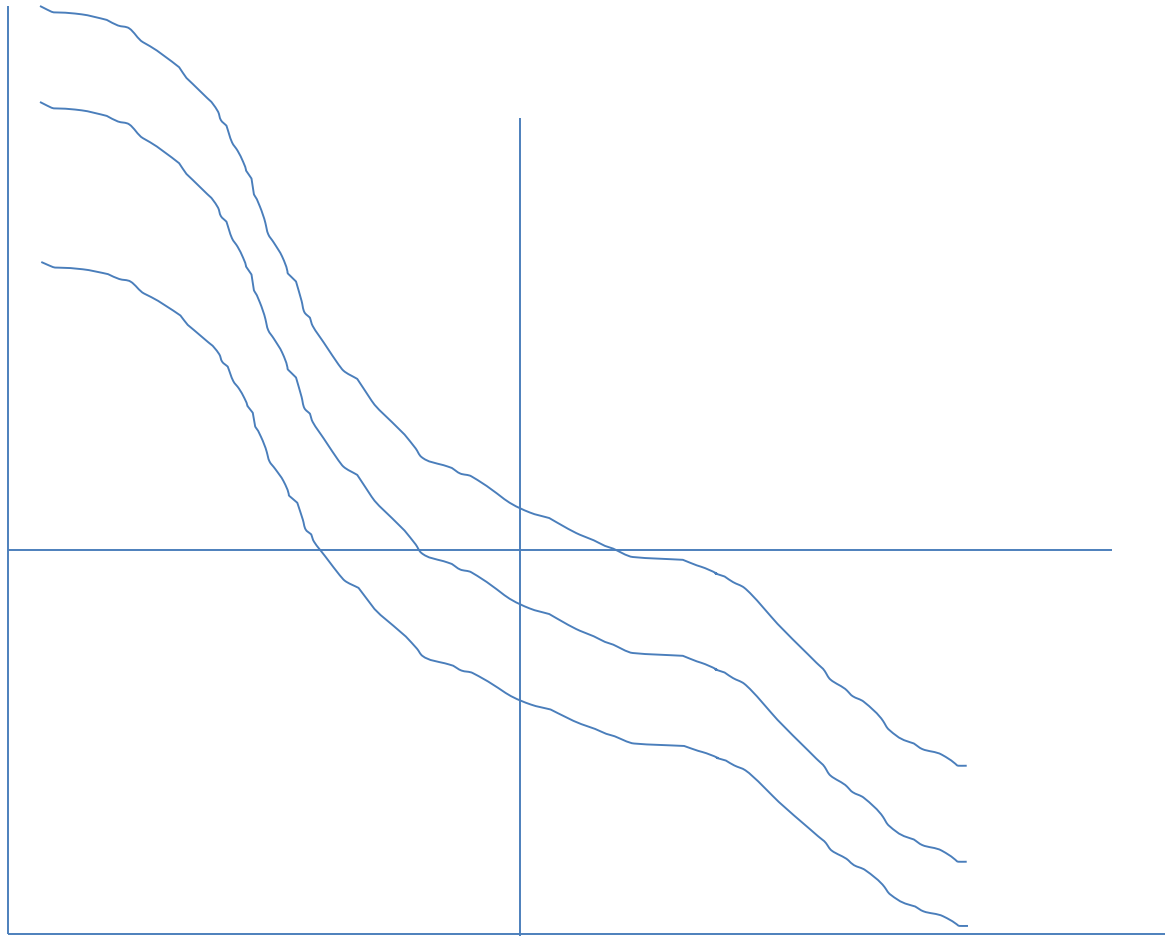


Assessment of uncertain functions in particular appears difficult

Variables that are elements of function spaces naturally extend standard DA

- **C: the space continuous functions from $\mathbb{R} \rightarrow \mathbb{R}$**
 - Often bounded, e.g., $C(0,1): \mathbb{R}(0,1) \rightarrow \mathbb{R}(0,1)$
- **Precedents**
 - Random utility functions in BDT choice models
 - Econometrics approaches involving uncertain functions
- **DA approach**
 - Mathematically consistent with axioms of DA and probability theory
- **Need to develop practical methods**

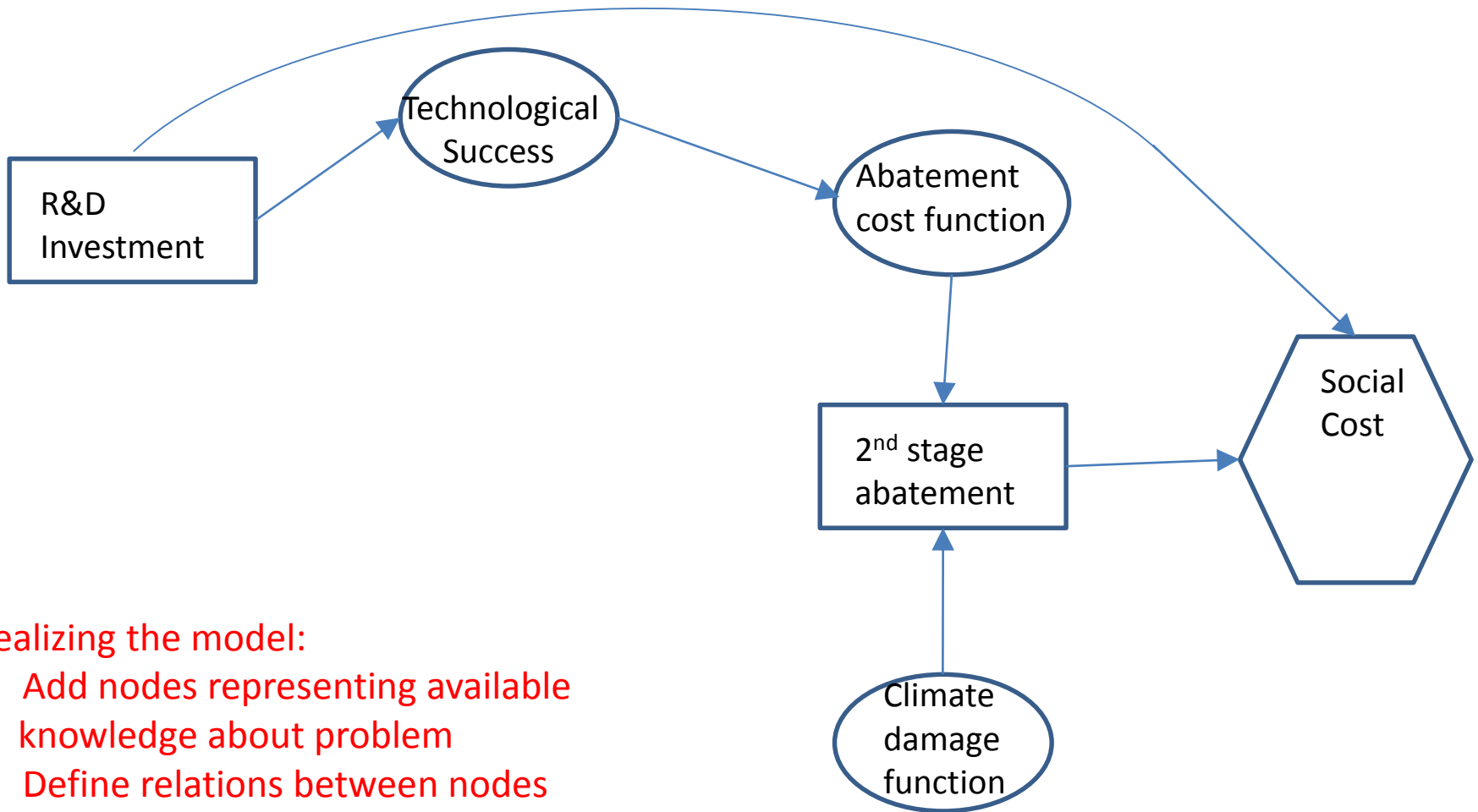
Challenge: Assessing probability measures on space of functions



Assessment methods analogous to those for real-valued variables

- With real variables
 - estimate probabilities of discrete outcomes
 - assumptions about the shape of distribution
- With functions
 - characterize in terms of real parameters
 - assumptions about shape of function
- Choose structure that avoids most difficult elicitations

Application to climate change problem



Realizing the model:

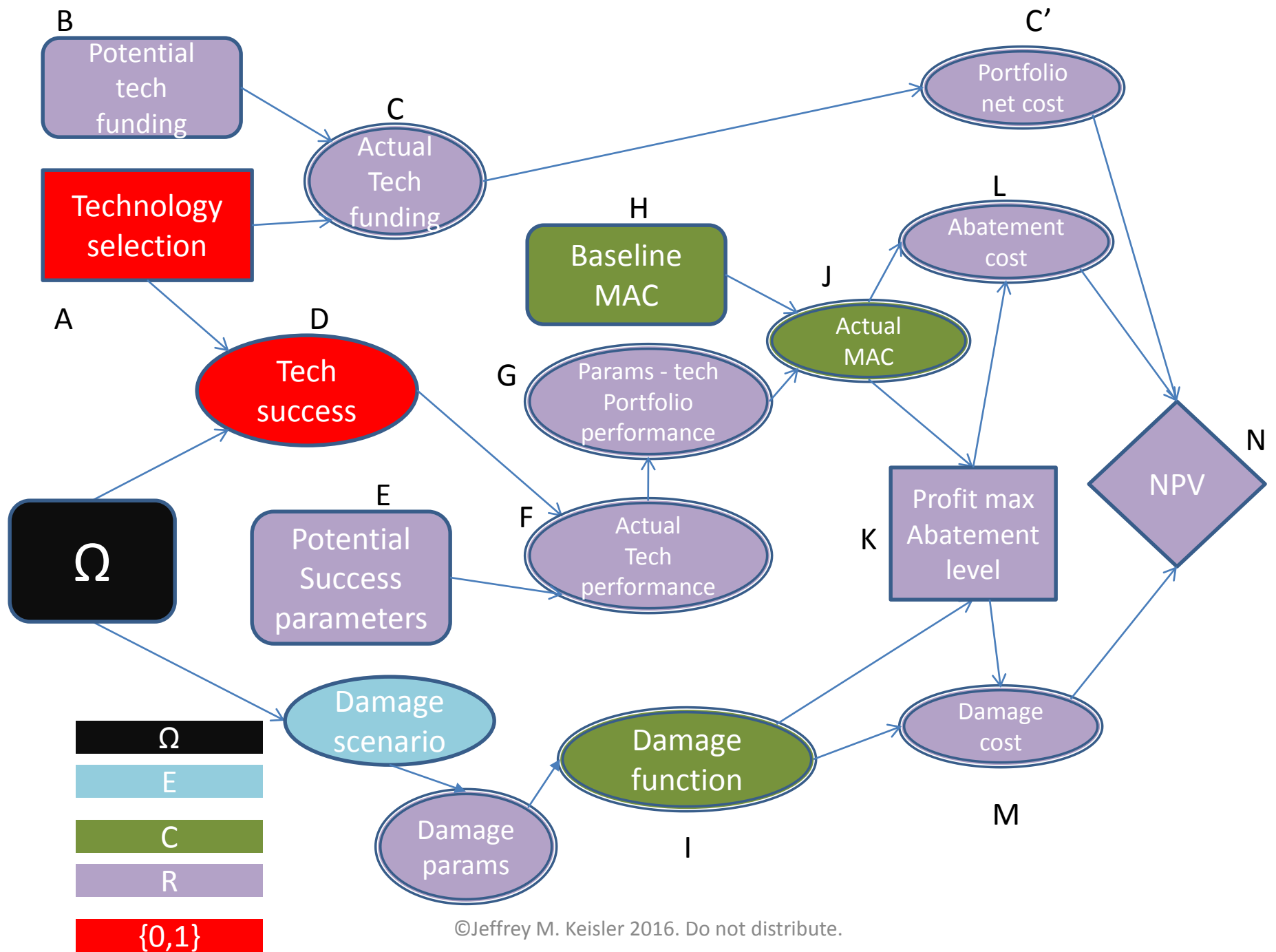
- Add nodes representing available knowledge about problem
- Define relations between nodes

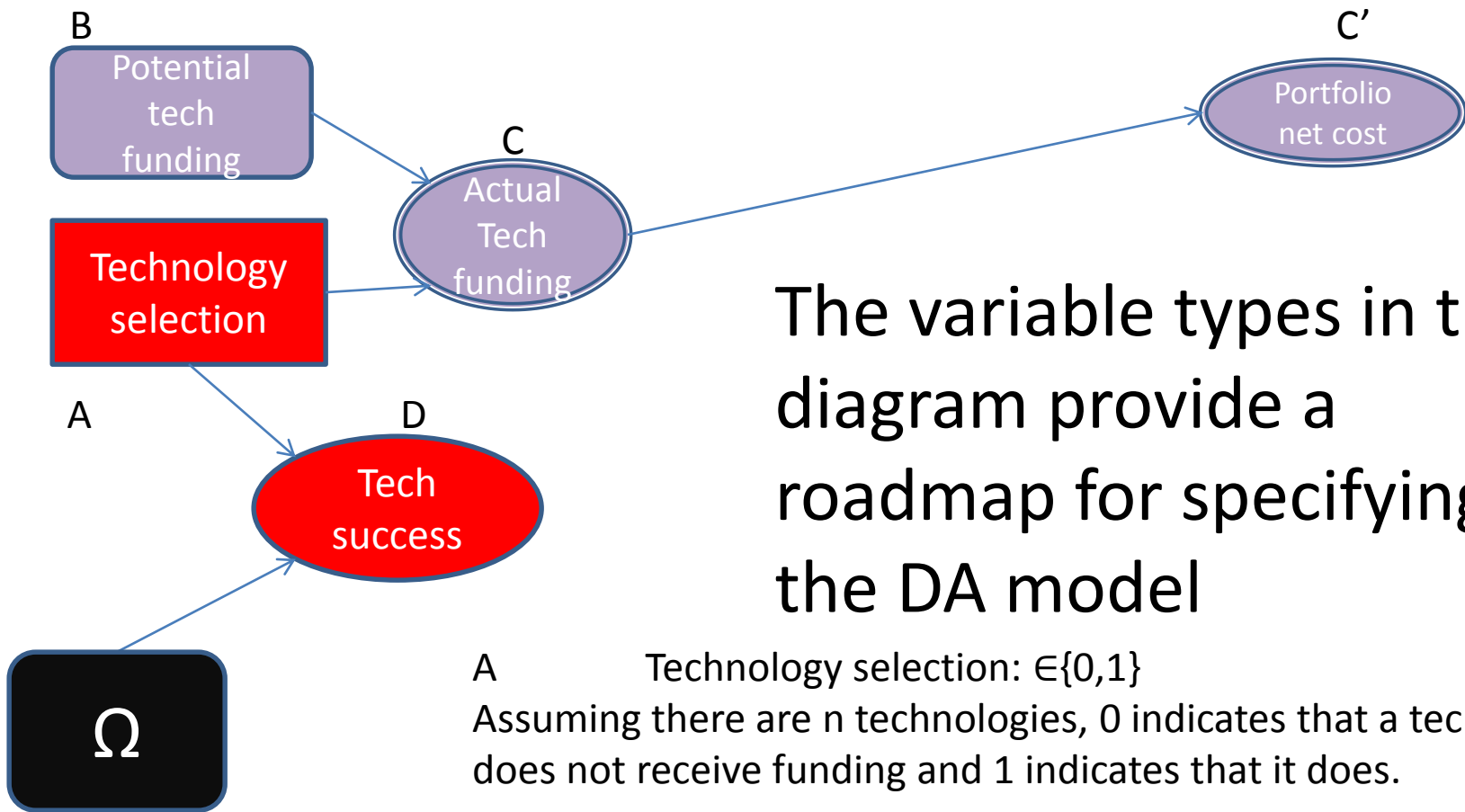
The art of modeling

- **Structure so as to model what is hard to assess:**
 - Uncertain demand and supply functions
 - Uncertain variables conditional on supply and demand functions
 - Transformations of uncertain supply and demand functions
 - Impacts on supply and demand functions
- **Structure so as to assess what is hard to model**
 - Likelihood of success
 - Future states

Modeling with malice aforethought

- **Composing functions** – simplifying by directly modeling or assessing a relationship in a single step
- **Decomposing functions** – simplify by breaking complicated variables into parts where it is clearer how to assess or construct connections
- **Ordering nodes** – Can rearrange
 - Bayes' rule holds for function-valued variables
- **Leads to a workable influence diagram, e.g., as follows**





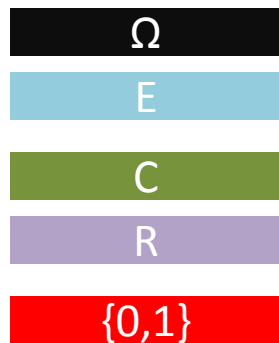
The variable types in the diagram provide a roadmap for specifying the DA model

A Technology selection: $\in \{0,1\}$
 Assuming there are n technologies, 0 indicates that a technology does not receive funding and 1 indicates that it does.

B Potential Funding for technologies: $\in R^n$.
 For each technology, we defined a funding trajectory to be assumed for later judgments; the NPV of a funded project is a social cost.

C Actual funding portfolio $\{0,1\}^n \times R^n \rightarrow R^n$
 Simply multiplies A and B.

C' Total NPV of funding for the portfolio (simply sums values from C)



Chance nodes represent mappings; elicit probability functions

- $\Omega \rightarrow \{0,1\}, \Omega \rightarrow E, \Omega \rightarrow R$: standard DA assessments
- $\Omega \times E \rightarrow E, \Omega \times R \rightarrow E$ etc.: Standard conditional assessments
- $\Omega \rightarrow C$: Exotic assessment methods
- $\Omega \times E \rightarrow C$: Exotic conditional assessments (difficult)

Deterministic nodes and relationships are modeled with standard math

- $E \rightarrow E$ or $E \rightarrow R$ or $R \rightarrow E$, $R \rightarrow R$

– Simple spreadsheet functions, operations, formulas

COMMON IN ECONOMIC ANALYSIS

- $C \rightarrow R$

– Functionals, e.g., Short programs, such as integration

- $R \rightarrow C$

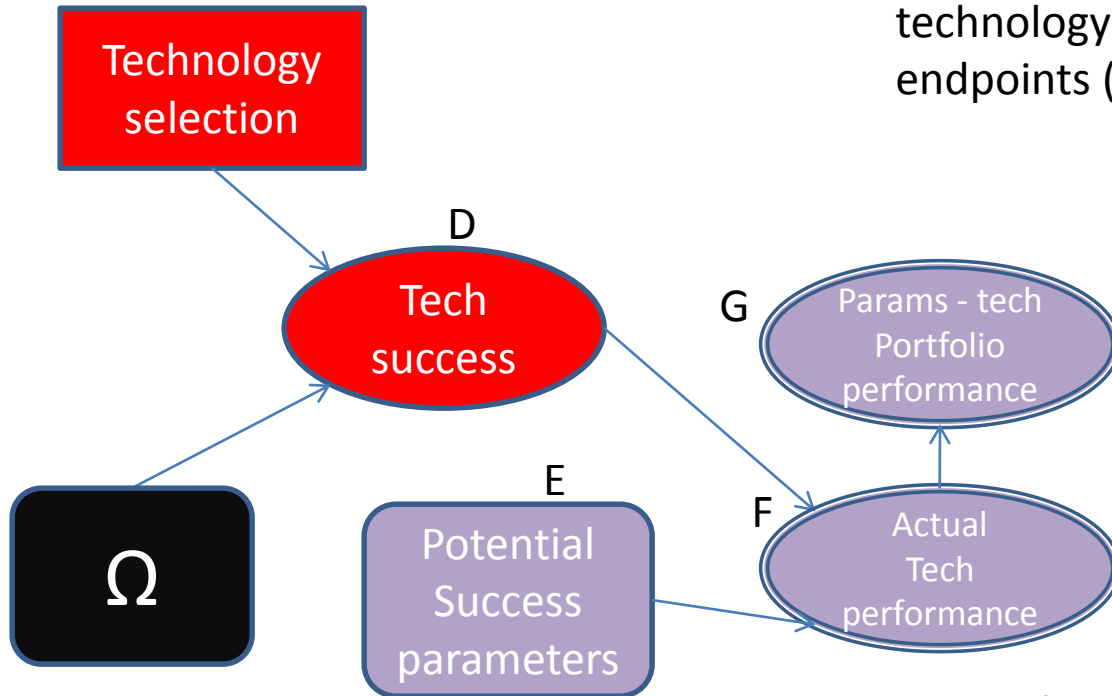
– Creating parametric functions, Spreadsheet formulas

- $C \rightarrow C$

– Operators, e.g., specialized programs

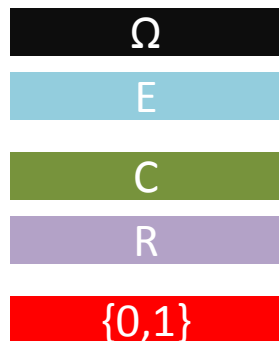
D Technology success: $\{0,1\}^n \times \Omega \rightarrow \{0,1\}^n$
 Standard R&D portfolio probability assessments

E Potential Success parameters for a technology: $R^{m \times n}$ using carefully defined endpoints (looking ahead)



F Actual successful technology performance:
 $\{0,1\}^n \times R^{m \times n} \rightarrow R^{m \times n}$ Simply multiplies D and E

G Technology portfolio performance: $R^{m \times n} \rightarrow R^m$
 Combines impact of all successful projects (F), as additive parameters to be used to calculate vertical shift, horizontal shift, etc. of the MAC

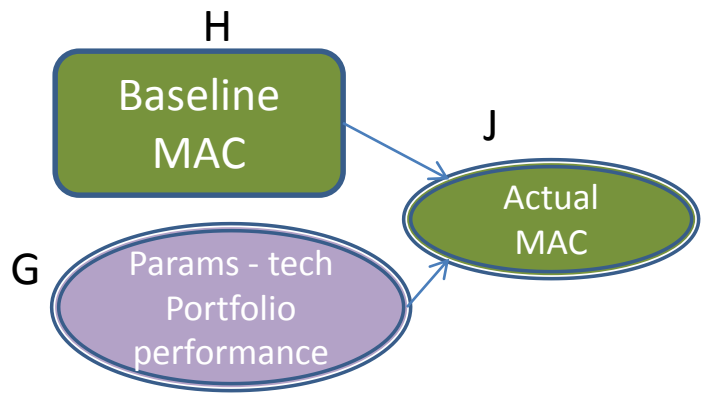


H Baseline abatement curve: $\in C$

We used the curve for the standard scenario already developed for Minicam.

J Actual abatement cost function: $C \times R^n \rightarrow C$

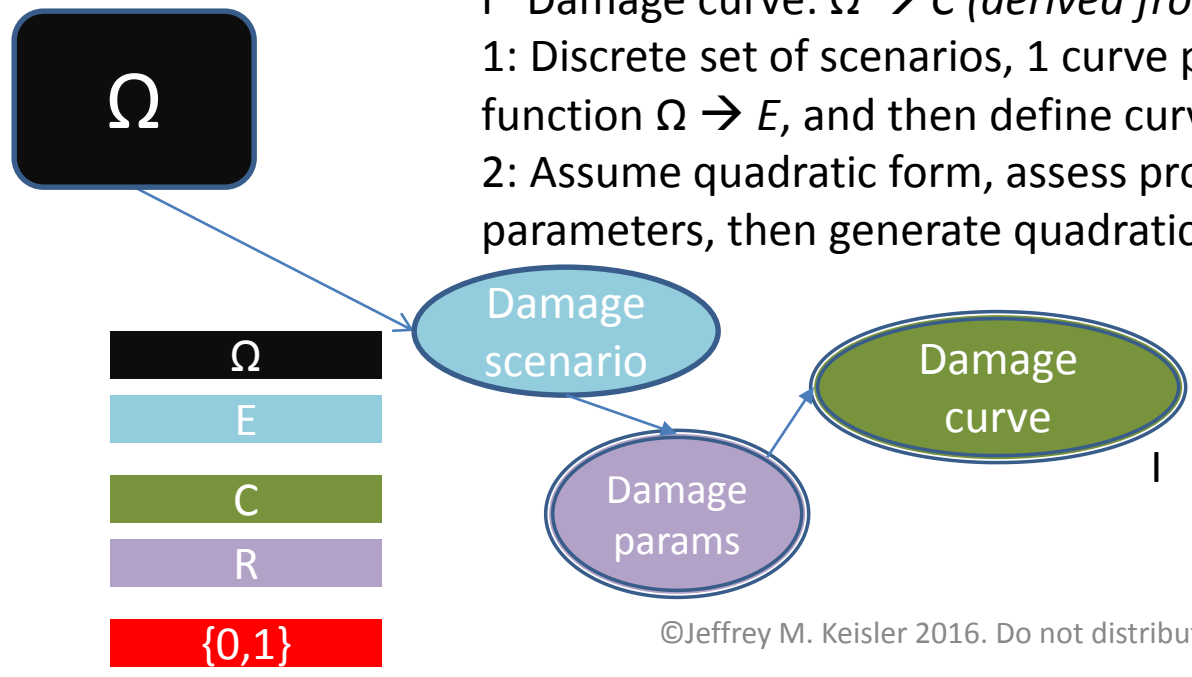
Uses various **linear operators** applied to the function in H and the parameter values from G.



I Damage curve: $\Omega \rightarrow C$ (derived from literature)

1: Discrete set of scenarios, 1 curve per scenario, assess probability function $\Omega \rightarrow E$, and then define curve for each event $E \rightarrow C$.

2: Assume quadratic form, assess probability function $\Omega \rightarrow R^3$ on parameters, then generate quadratic function $R^3 \rightarrow C$.

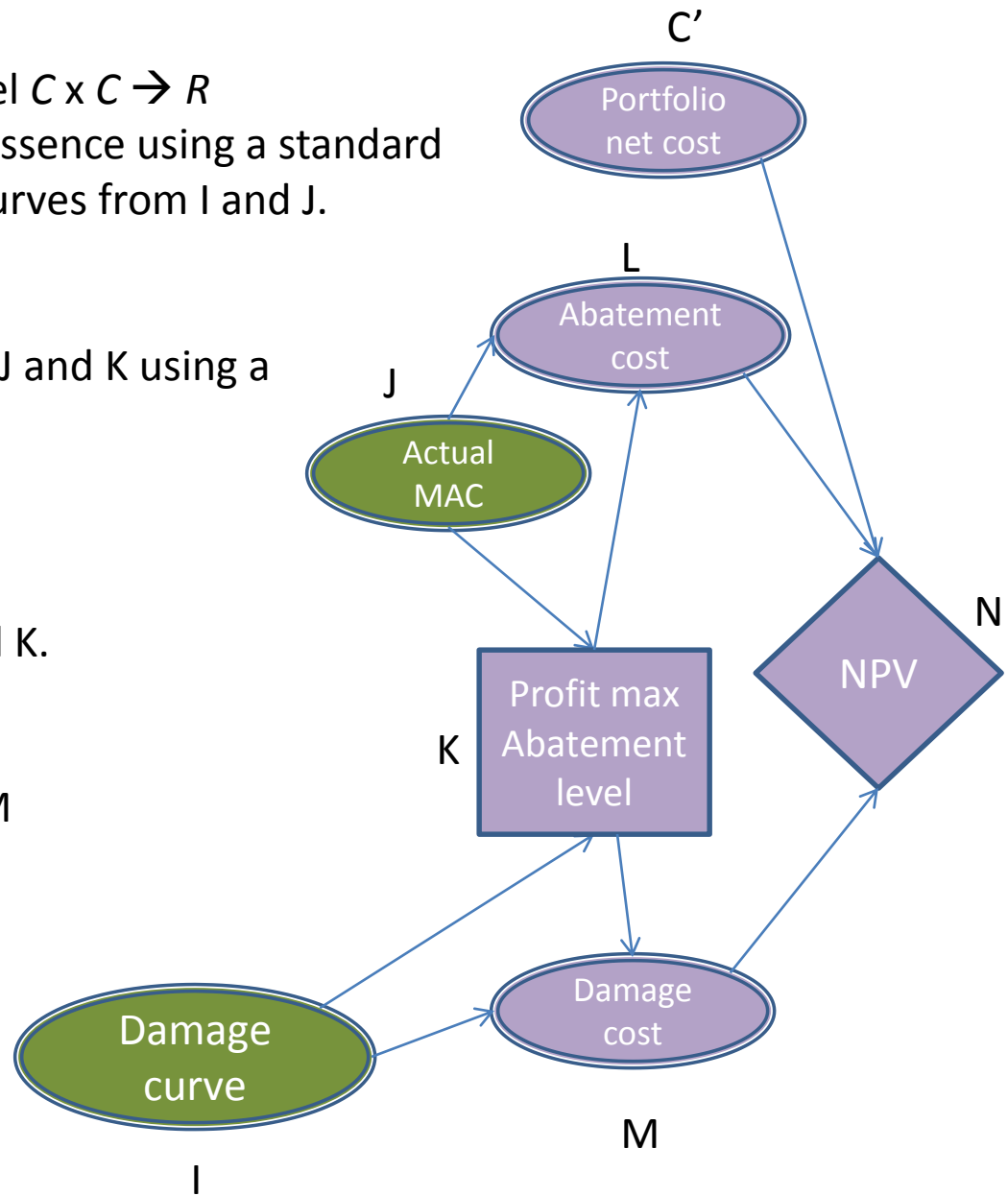


K Profit maximizing abatement level $C \times C \rightarrow R$
 This is implemented in Minicam, in essence using a standard economic **functional** based on the curves from I and J.

L Abatement cost $R \times C \rightarrow R$
 This is calculated from the results of J and K using a simple economic **functional** – reading a value off the curve.

M Damage cost $R \times C \rightarrow R$
 Similar to L, using the results of I and K.

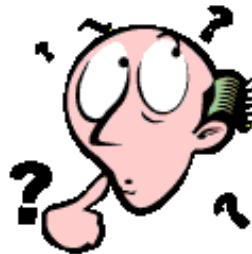
N Societal cost: $R^3 \rightarrow R$
 Simply adds the results of C', L and M (with appropriate discounting)



- Ω
- E
- C
- R
- {0,1}

The composition of these functions is used to calculate expected societal cost for any given R&D portfolio

- $N(C'(A,B),L(J(H,G(F(D(A,\Omega),E))),K(I(\Omega),J(H,G(F(D(A,\Omega),E))))),M(I(\Omega),K(I(\Omega),J(H,G(F))))))$



- We'll let the computer handle that one!
- Simpler to compute but impossible to assess would be $E[N(A, \Omega)]$ for each alternative

Implementation

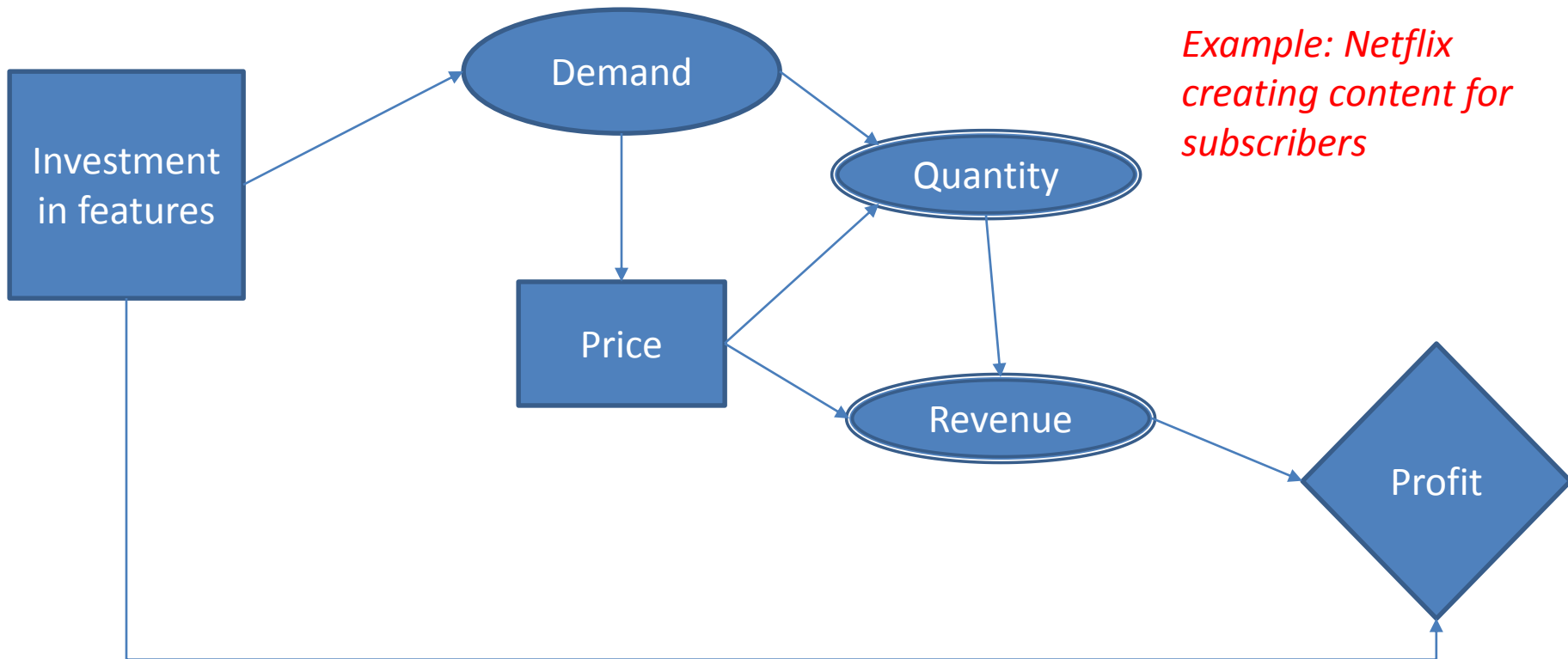
- **Structured assessments according to the plan to anticipate connection to economic analysis models**
 - Identified technical hurdles
 - Assessed probability of success as function of funding
 - Endpoints of R&D success were individual technology parameters (e.g., cost/Kg) that could be combined into economy-wide parameters used to derive economy wide abatement cost curve, or allow direct calculation of amount of “shift”, “pivot” of functions, etc.
 - Defined and estimated functional relationships
- **Range of possible damage curves from published literature**
 - Based on scientific climate models and economic models
- **Modeling in Minicam/DICE (Baker & Solak) produced suggestive results**

Platform ecosystems

(if we have more time)

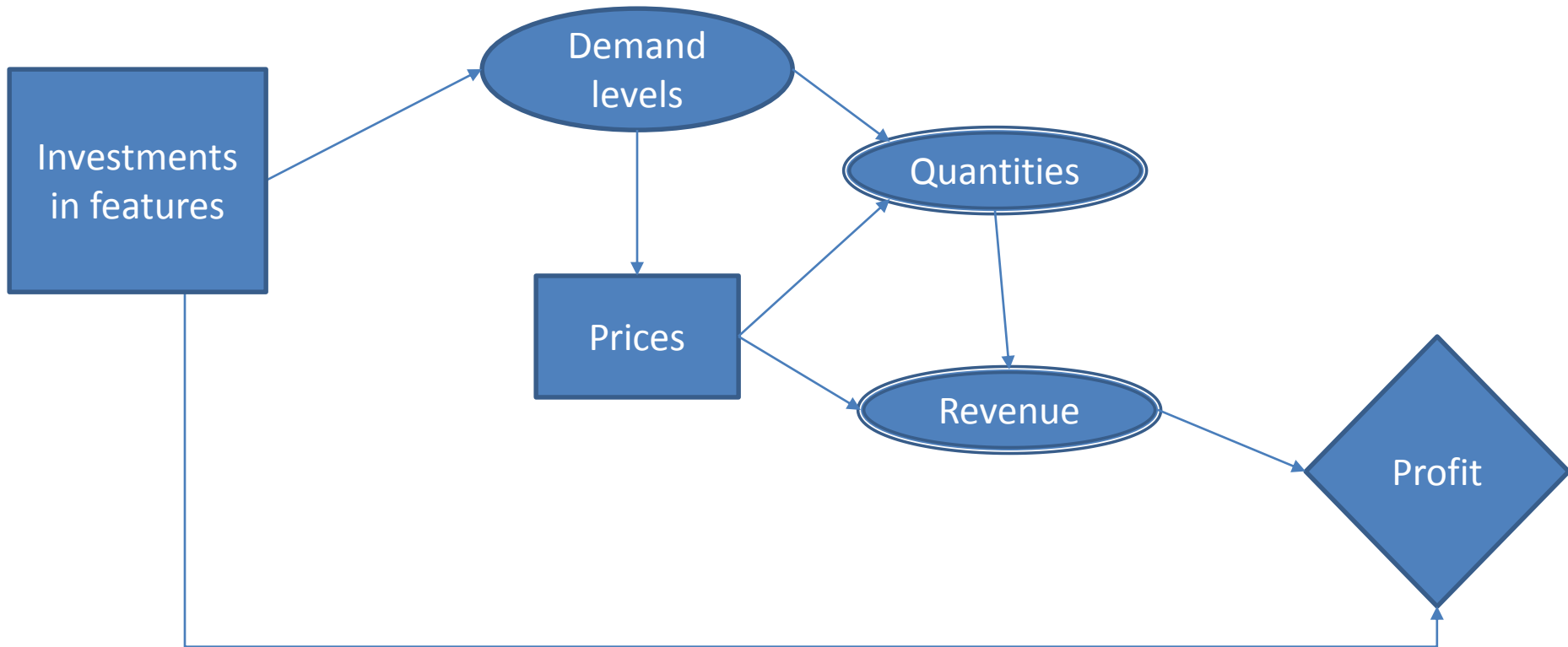
- Two sided markets
 - Value to buyers depends on number of sellers
 - Value to sellers depends on number of buyers
 - Extends to multi-sided markets
- Economic / strategy theory since ~2000
- Current efforts
 - specifying decision analytic approach
 - starting simple

One-sided market platform model is variation on earlier examples

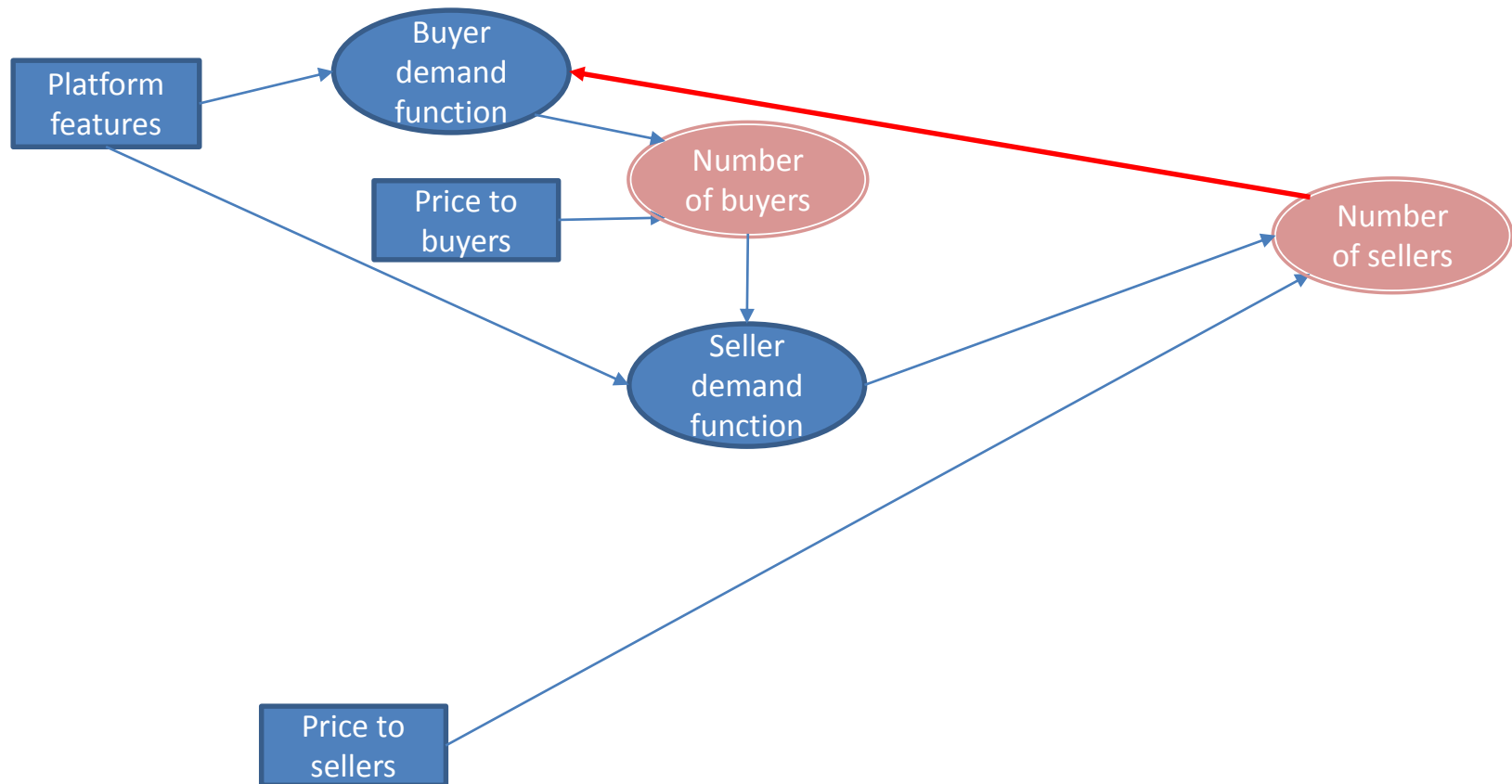


Assume quantity represents number of users, price is fee per user, with no additional modeling of individual transactions

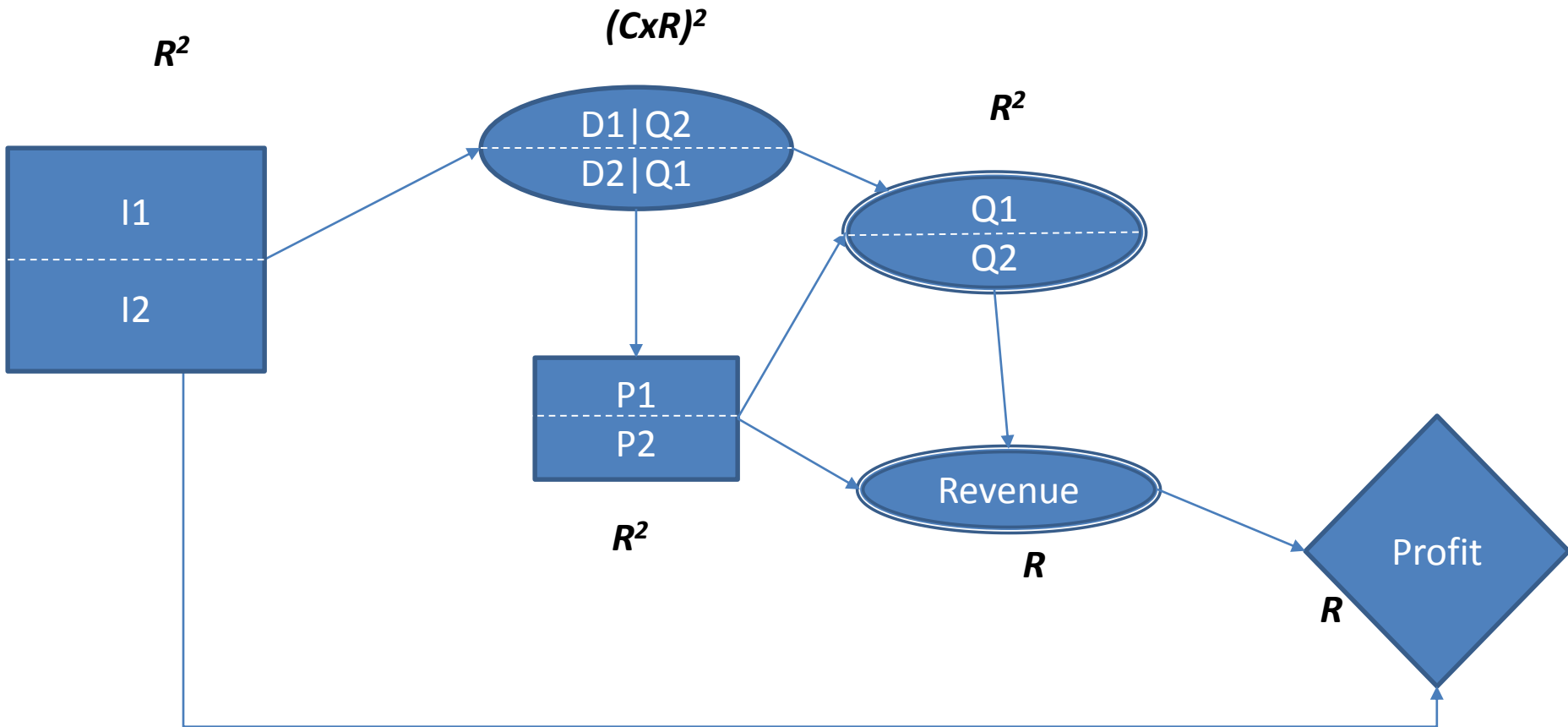
Two sided market – Same diagram but more complicated implementation



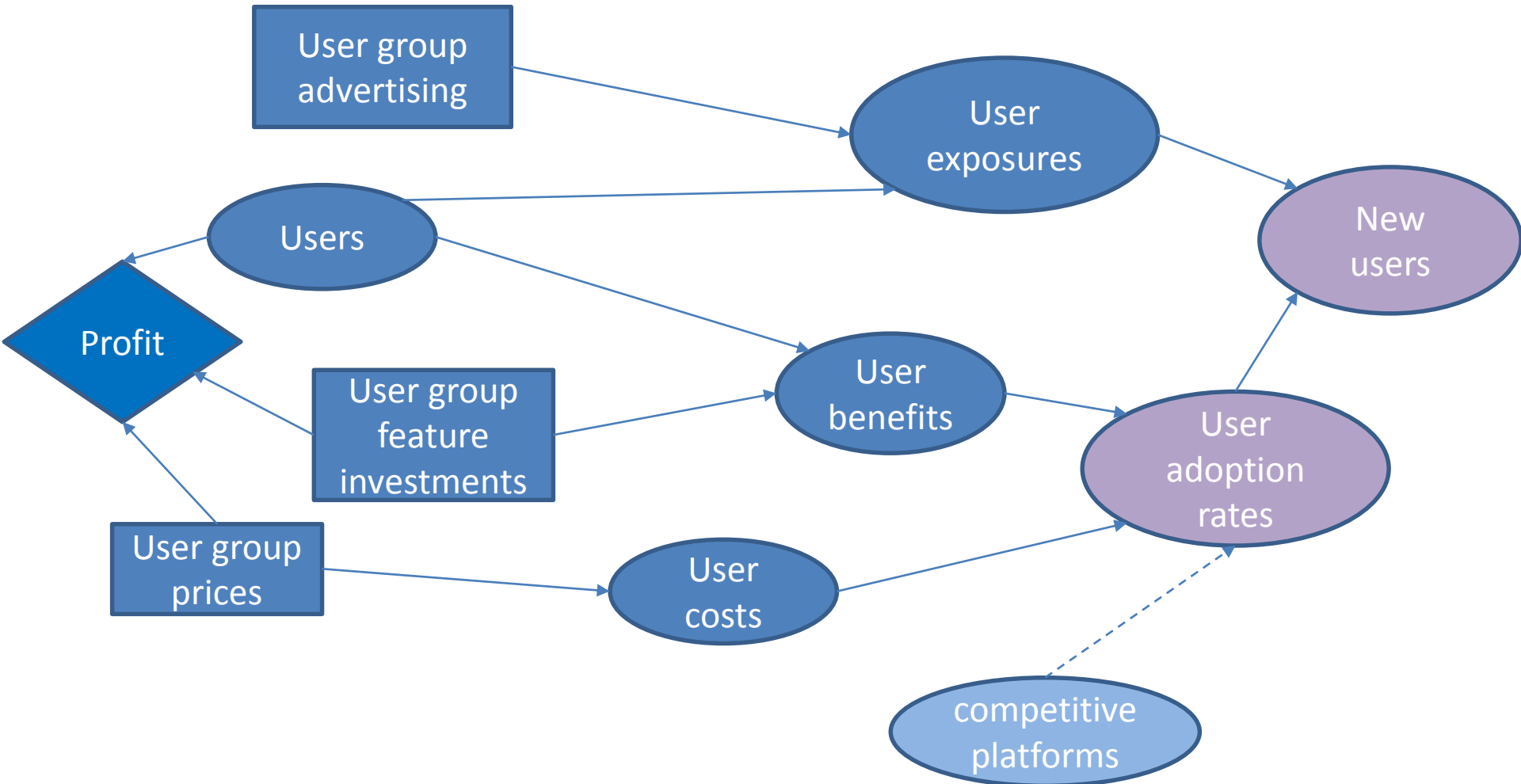
Influence diagrams do NOT have cycles



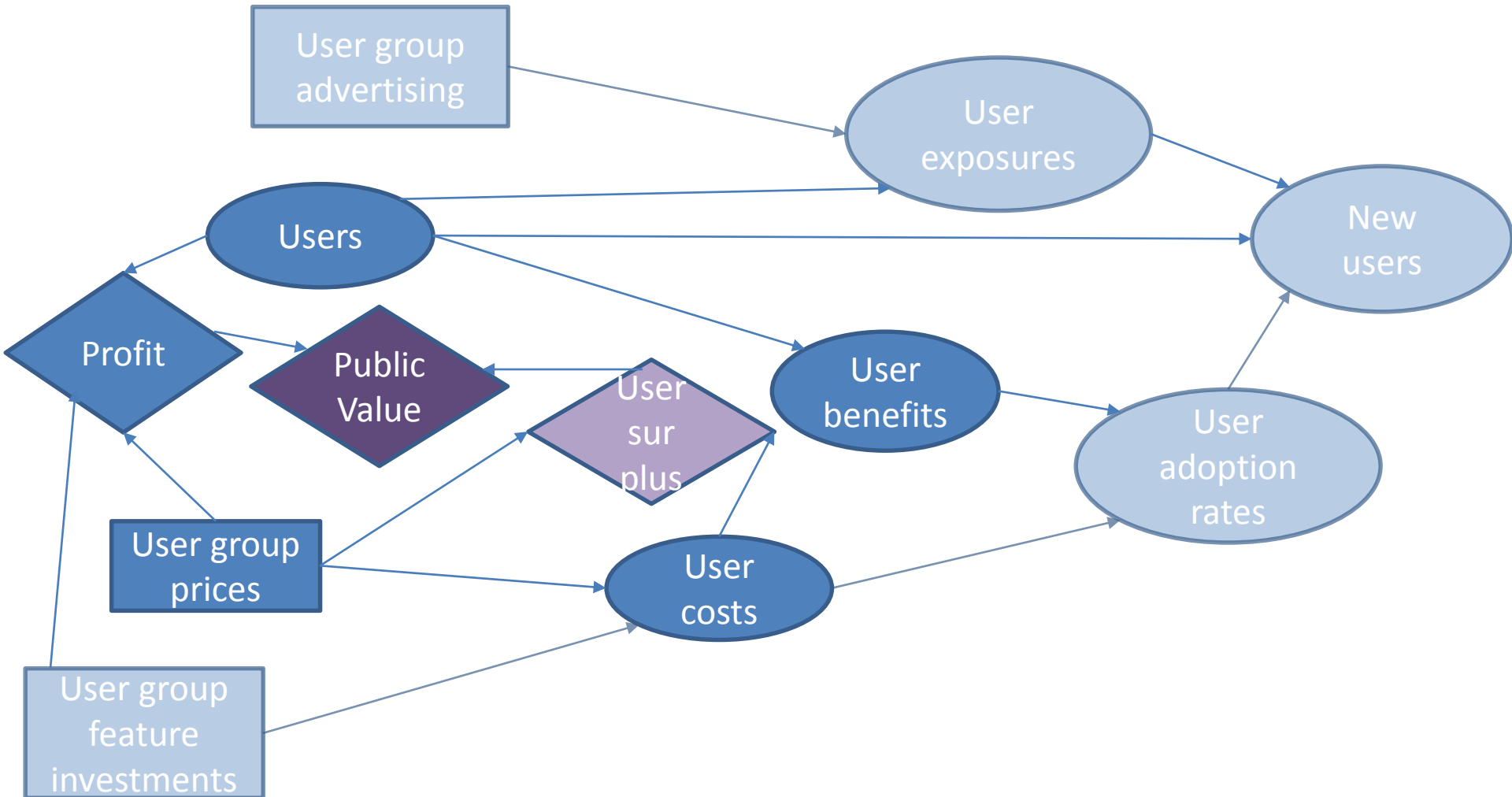
Solution



Dynamic model



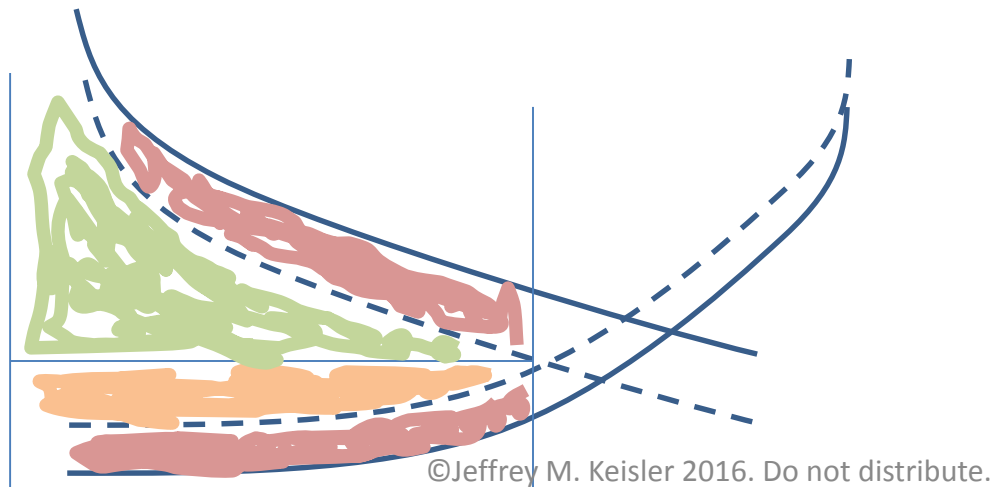
Public perspective

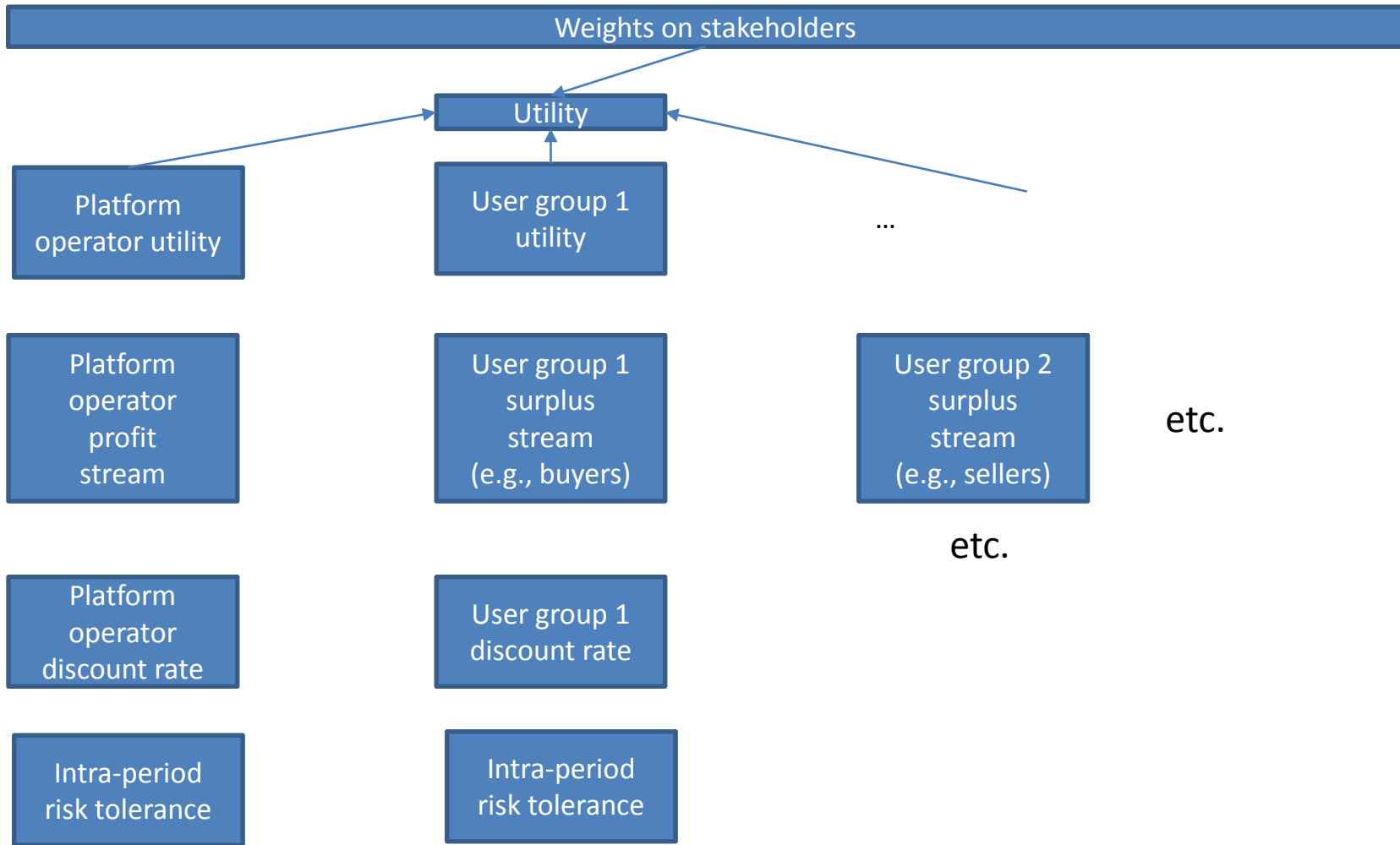


Spreadsheet example

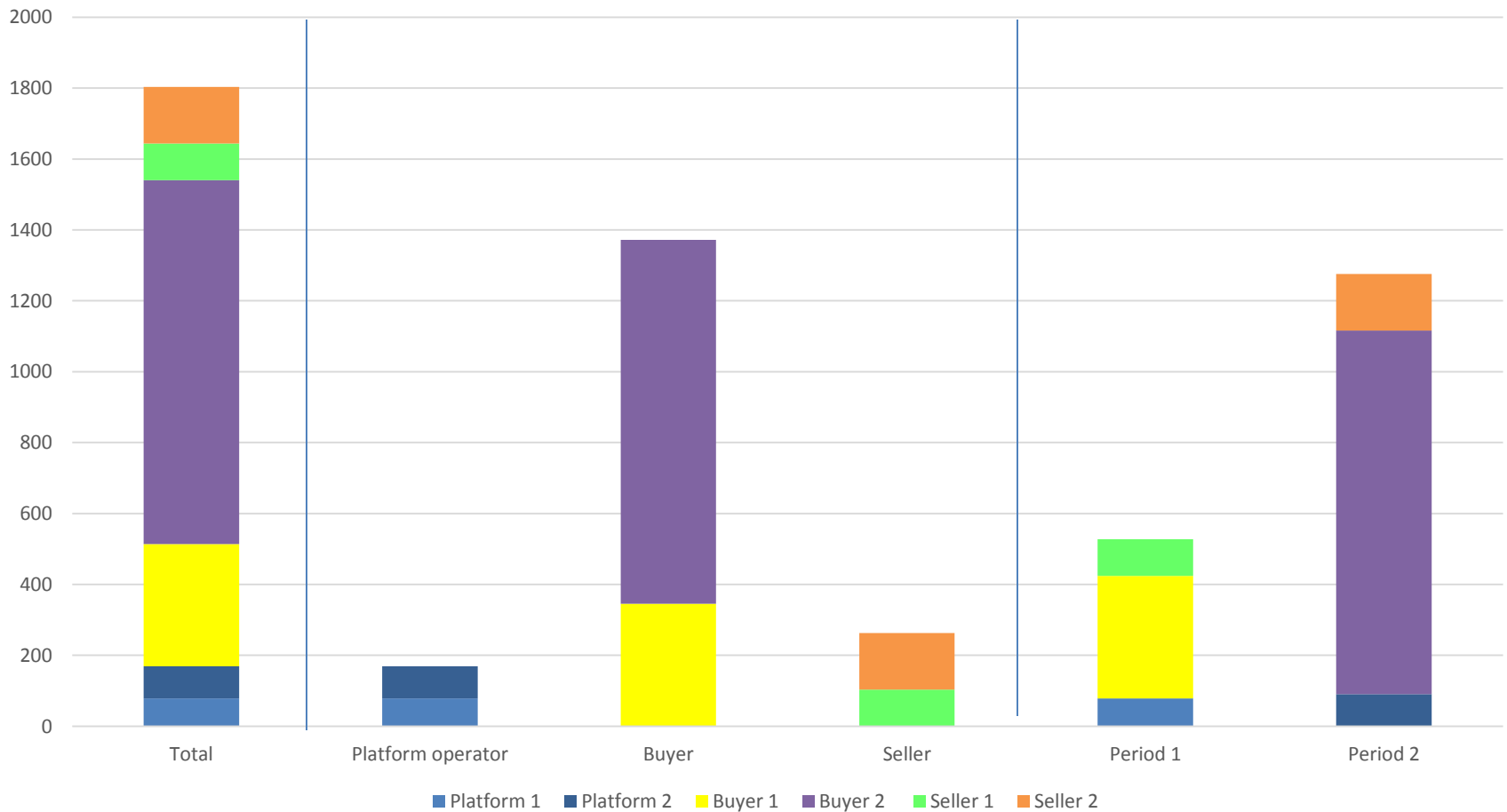
Extension to government problem

- Balancing interests in backing plans
- Economic analysis computes buyer surplus, seller surplus, platform operator profit, etc.
- Discount over time
- Can use MCDA / MAU for multi-stakeholder view





Breakdown of platform benefits



Conclusion

- Decision analysis can use function valued variables
- Structuring models requires some novel ways
- Allowing incorporation of common micro-economic modeling methods
- Enabling insights about complicated problems like platform ecosystem design

Jeff.Keisler@umb.edu