

# **Minimalism and Model-Building: An assured model of the exchanges among consumers, retailers, and manufacturers**

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**December 2007**

# Agenda

- **purposes:**  
*assurance* = verification and validation
- **issues in ABM**
- **assuring ABMs**
- **minimalism, or beyond parsimony**
- ***Supermarket ABM 2.0***
- **verification**
- **validation**

## Simulation and Sufficiency

**Simulations might attempt, inter alia:**

- **to *explain* a phenomenon;**
- **to *predict* the outcome of a phenomenon; or**
- **to *explore* a phenomenon, to play, in order to understand the interactions of elements of the structure that produce the phenomenon.**

**Simulation derives sufficient conditions for the emergence of a phenomenon.**

**Ironically, sufficient conditions are necessary (but not sufficient) for explanation.**

## Consider historical market data:

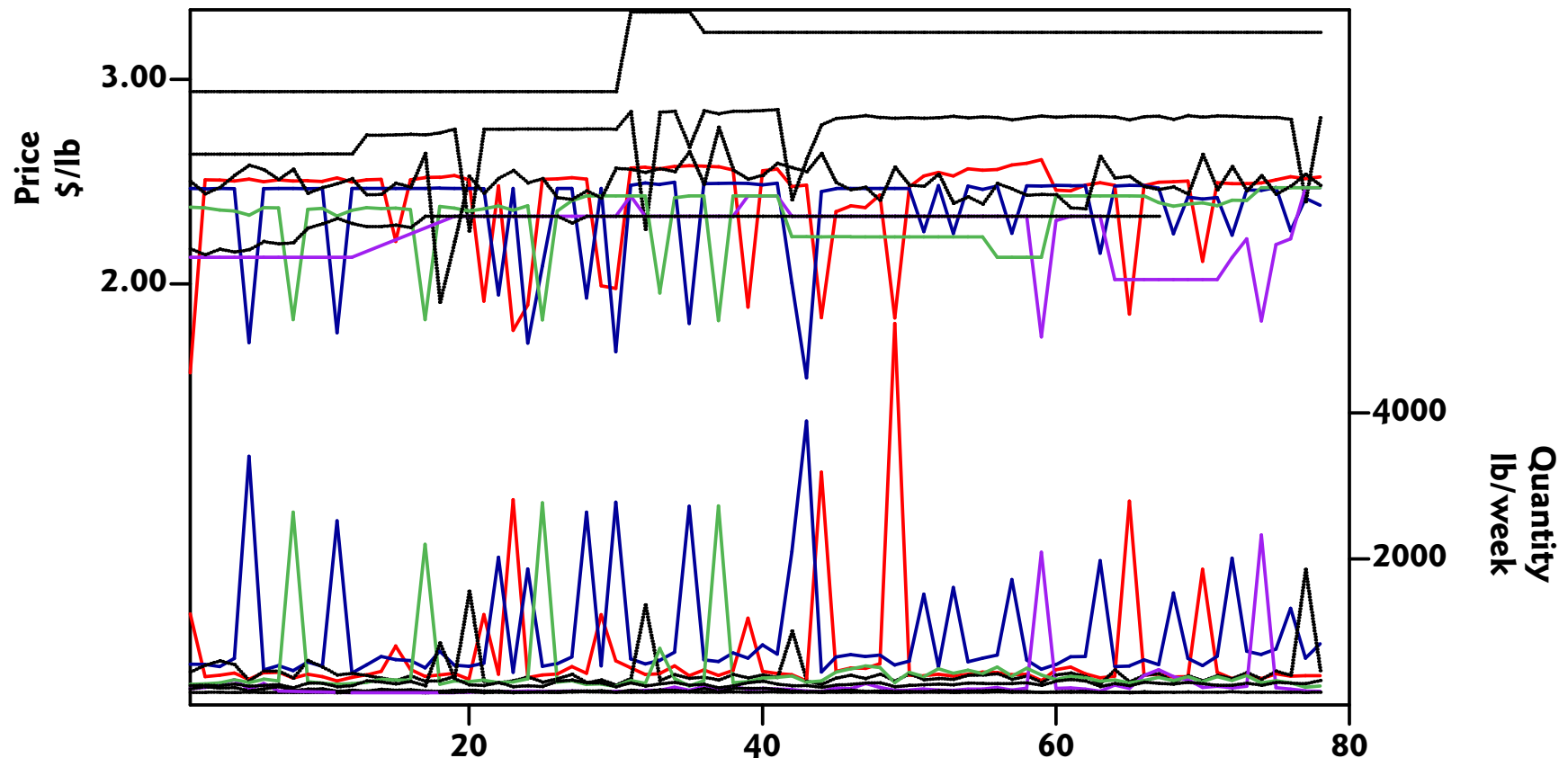


Figure 1: Weekly Prices and Sales (Source: Midgley et al. 1997)  
(Coloured lines: **Folgers**, **Maxwell House**, **Hills Bros**, **CFON**)

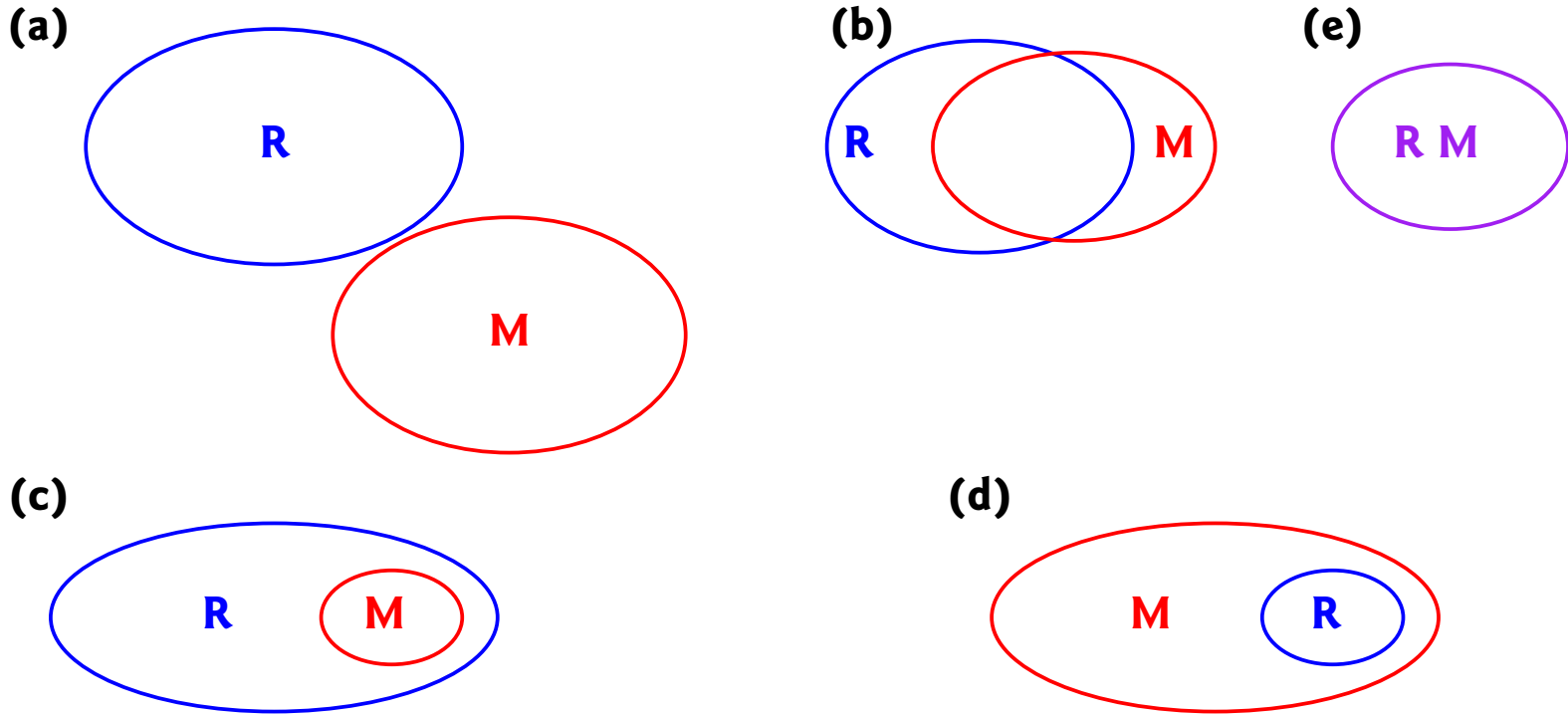
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## Simulations with Agent-Based Models

**We have been pursuing a research programme of simulation of oligopolistic market behaviour, first, of a ground coffee market, and, today, of a market in laundry detergent.**

**We are searching for simple sufficient conditions to produce observed behaviour, which requires model *assurance*: verification and validation.**

**With a metric in performance space, we can derive a graphical description of five possibilities for validation: model behaviour  $M$  and real-world phenomenon  $R$ .**



**Validity Relationships for prediction (after Haefner (2005)).**

**M = model behaviour**

**R = historical reality**

- (a) the model is (pretty) useless**
- (b) the model is both incomplete and inaccurate**
- (c) the model is accurate but incomplete**
- (d) the model is complete but inaccurate**
- (e) the model is complete and accurate (in your dreams!)**

## Issues in ABM

- **ABMs have several benefits, but they are computer *simulations*.**
  - ∴ **to liberate their potential, three key issues must be addressed:**
    - **Does the software implement the desired model?**
    - **Is the system always well behaved?**
    - **Can the models be empirically validated?**
- **Recently, we proposed a procedure to address these issues, called *model assurance*.**  
**see Midgley, Marks, Kunchamwar (2007)**

# Minimalism

## Parsimony:

- Other things equal, the simplest solution tends to be the best one (from *Wikipedia*: “Occam’s razor”, and see Einstein too)

## Minimalism:

- What is the simplest solution that explains 80% of the observed phenomenon?

## Why?

- Current science tends to have many subtleties and nuances, pushing explanation from 80% to 90%,  
but Incorporating these into ABMs results in models that are very difficult to assure,  
so It is better to start minimalist and then to add complications later, one-by-one.



## ***Supermarket ABM 2.0***

**Three types of agent: Consumers (several subtypes), Retailers (two stores), and Manufacturers (eight brands).**

**The overriding philosophy is to keep the model simple by keeping the agents as simple as necessary, but no simpler.**

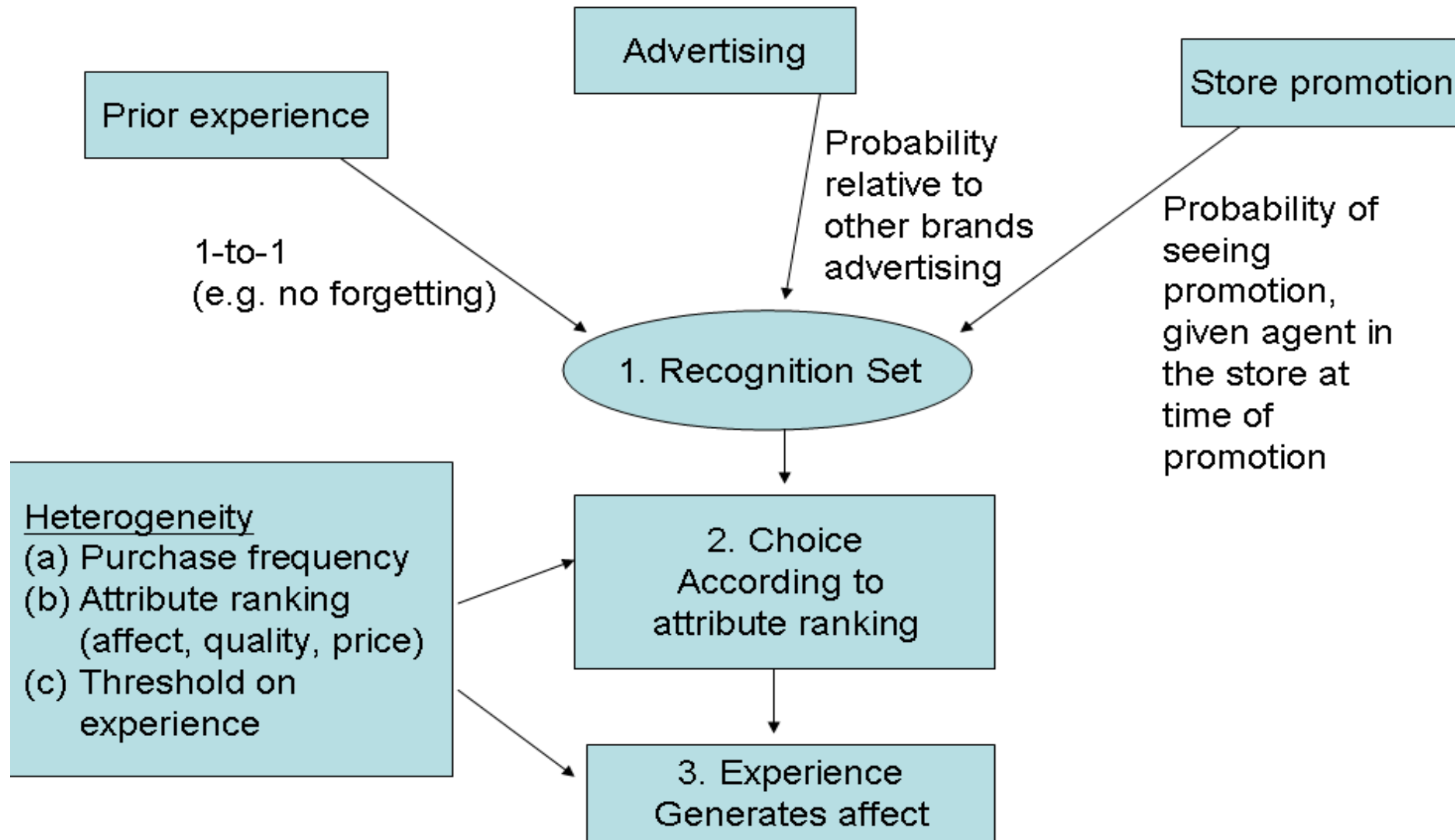
**Consumer agents vary by: purchase frequency, attribute (affect, quality, price) rankings, and threshold on experience.**

**Each drawn from one of four distributions.**

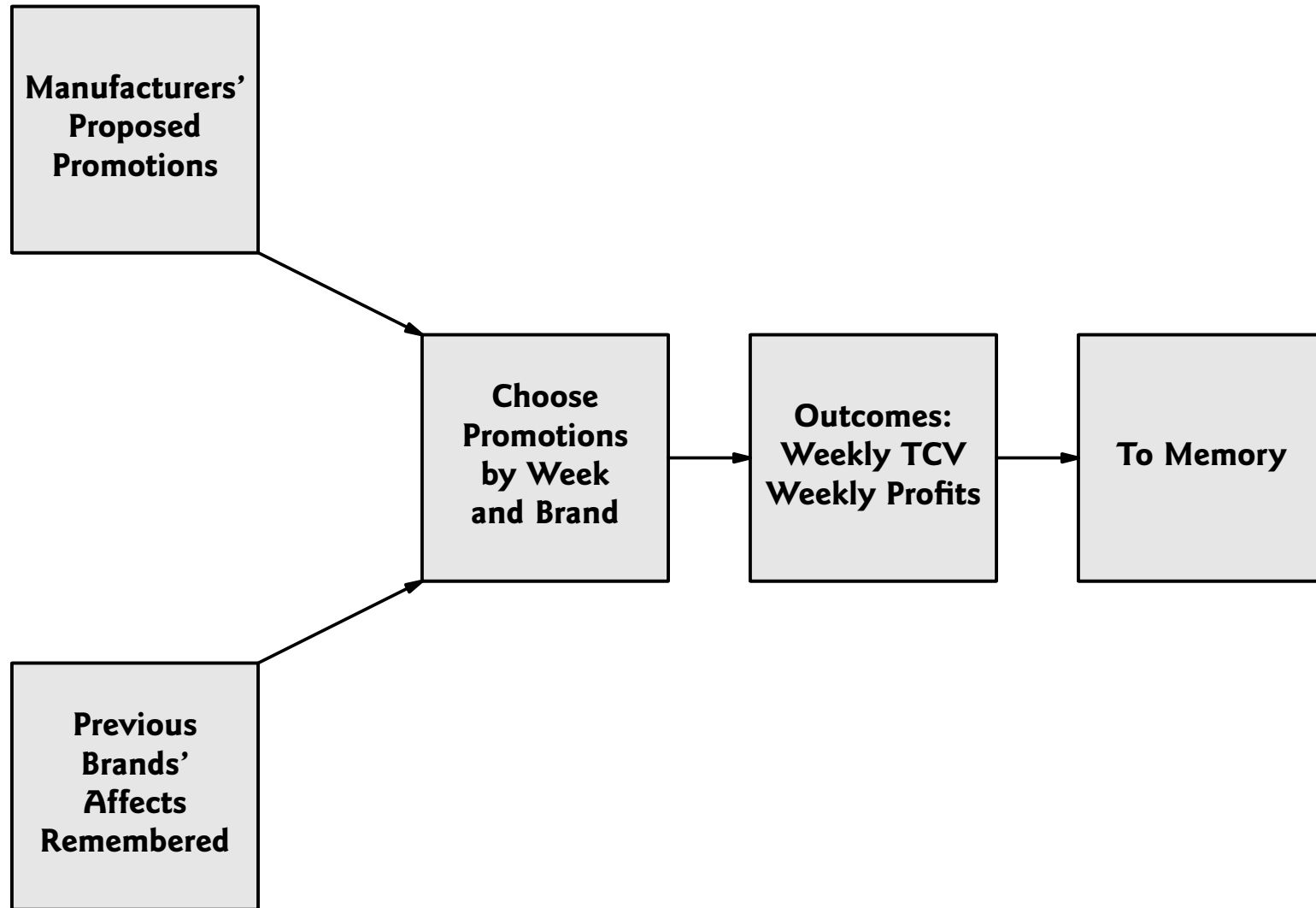
**Retailer agents have several assigned parameters, including number of consumers (exogenous).**

**Manufacturer agents have several parameters drawn from one of eight distributions.**

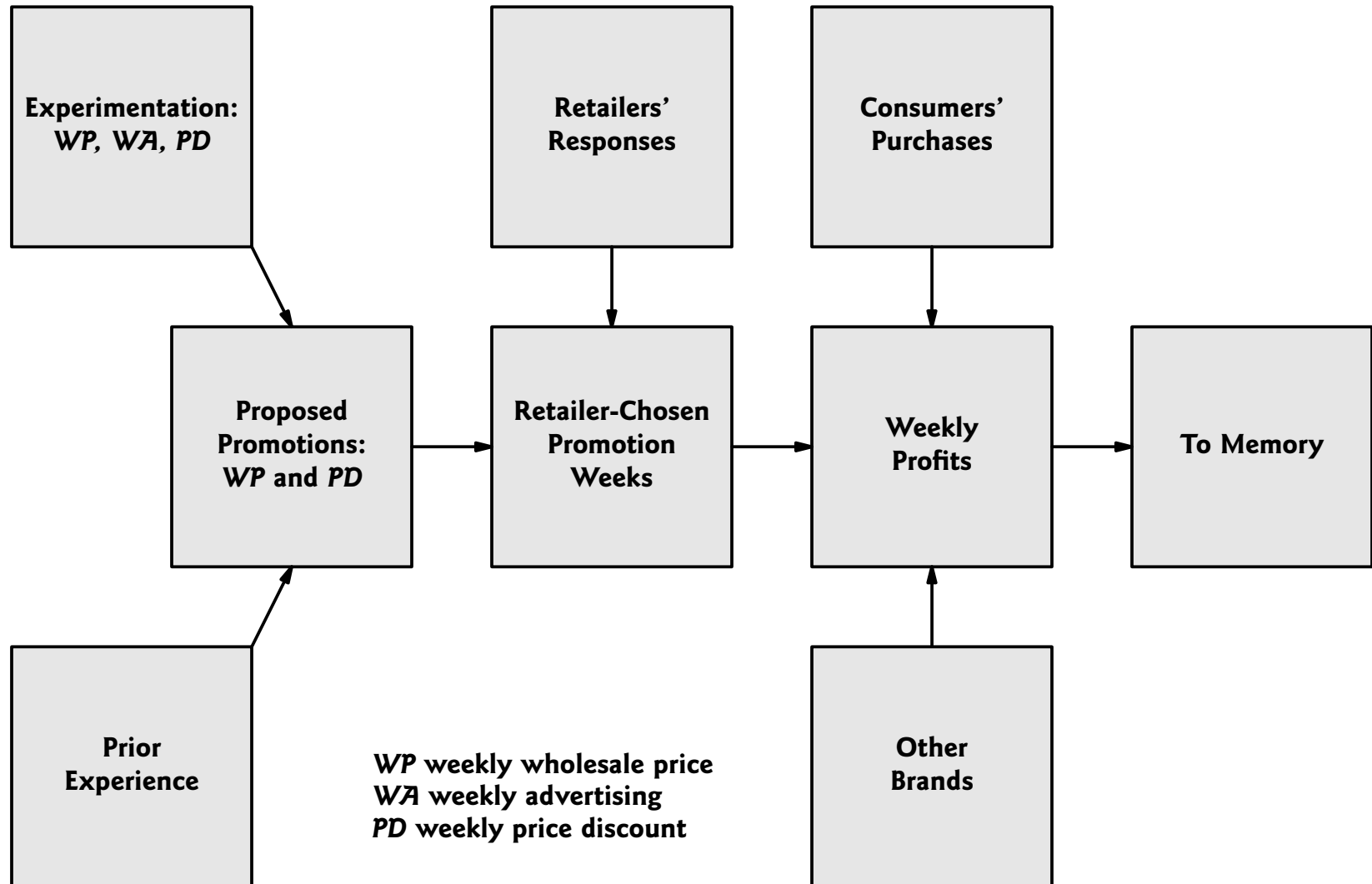
# Consumer agent



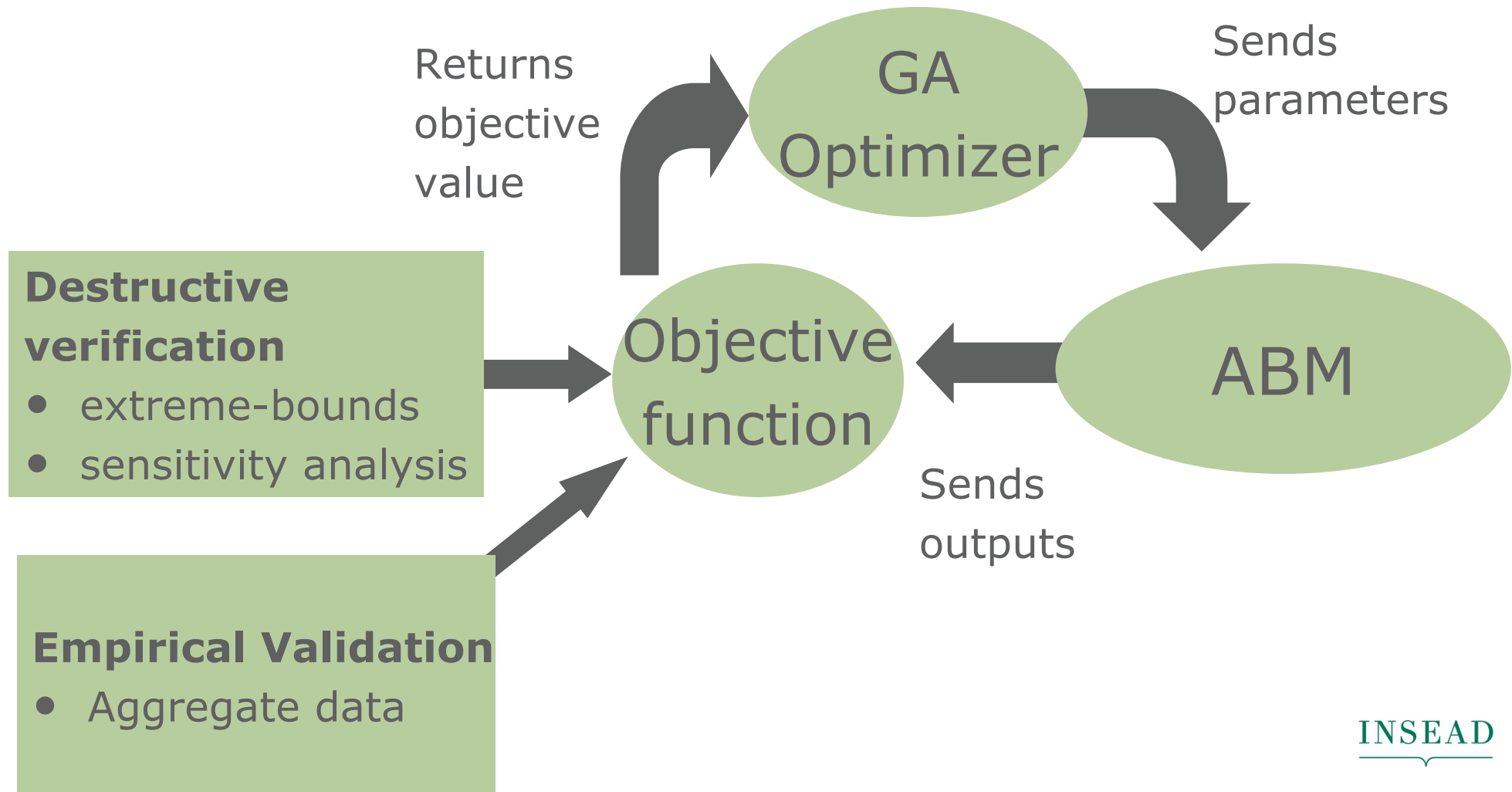
# Retailers



# Manufacturers



# Work at the macro-level: Embed the ABM in an Automated Nonlinear Testing System



# Verification

## Two steps:

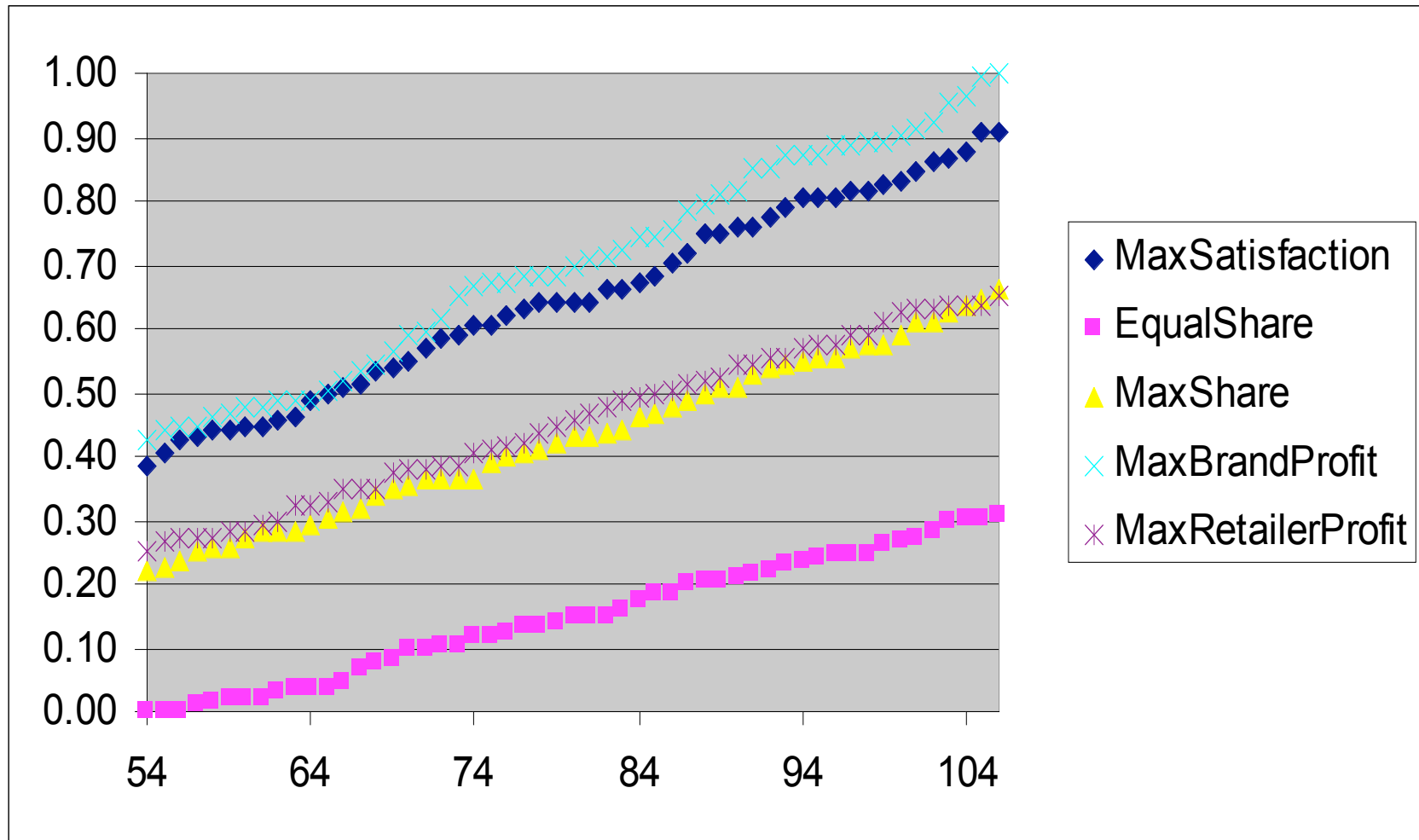
- I. **Code checked against specification by two independent judges (PhD students in computer science).**
  - **Good news: some redundant and unused code, some interpretation issues, but relatively minor.**
  - **Bad news: time consuming.**

## **2. Destructive Testing**

**Using Miller's 1997 ANT system:**

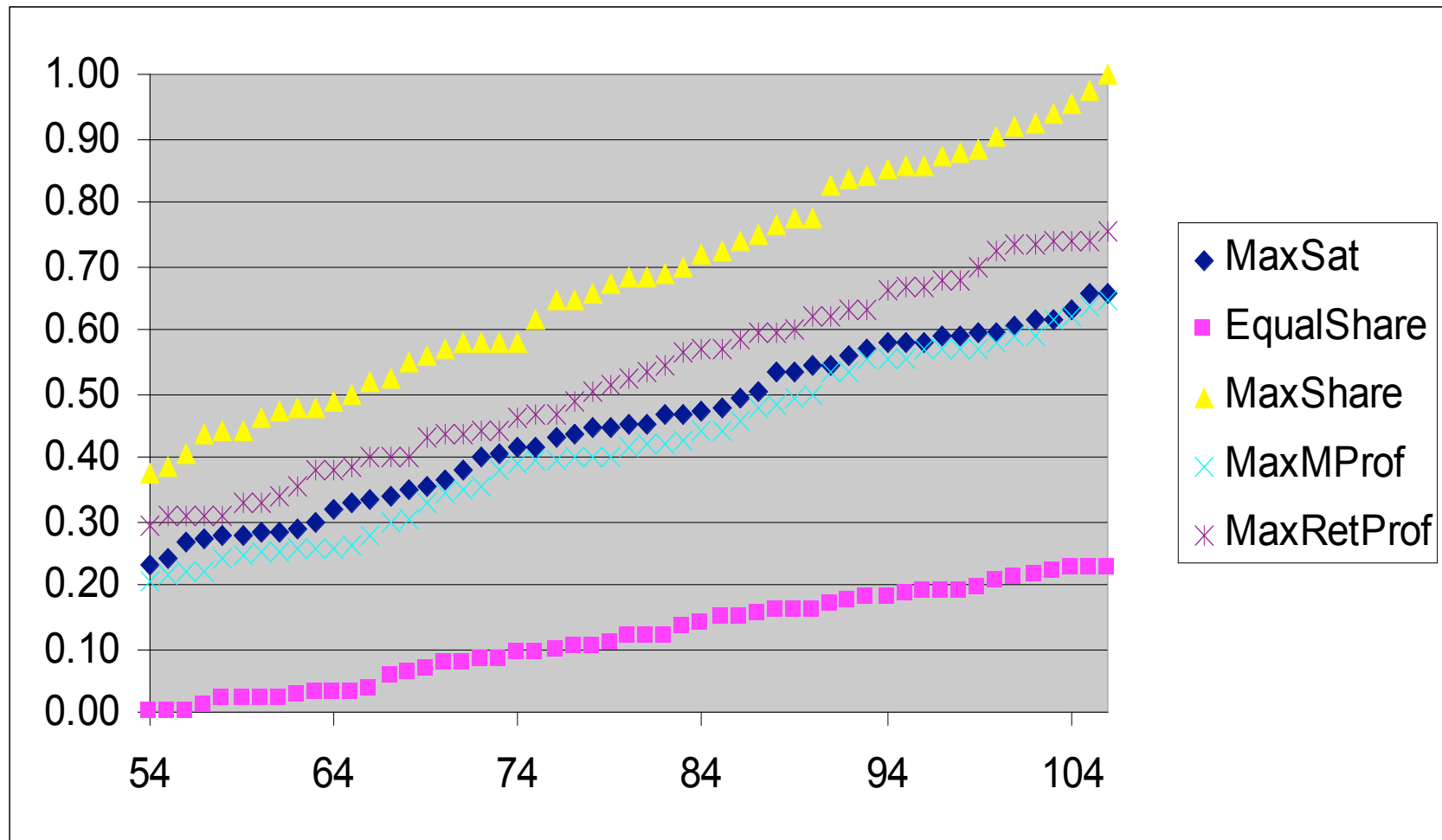
- 5 extreme objectives tested for (consumer satisfaction, equal market shares, manufacturer profit, monopoly, retailer volume)**
- Consumer agents already micro-optimized (to reduce the number of parameters and hence the degrees of freedom)**
- GA used to optimize remaining parameters.**
- Good news: difficult to drive the model to extremes.**
- Bad news: noisy optimization, hard to get convergence.**

# Consumer satisfaction for the leading brand under 5 criteria

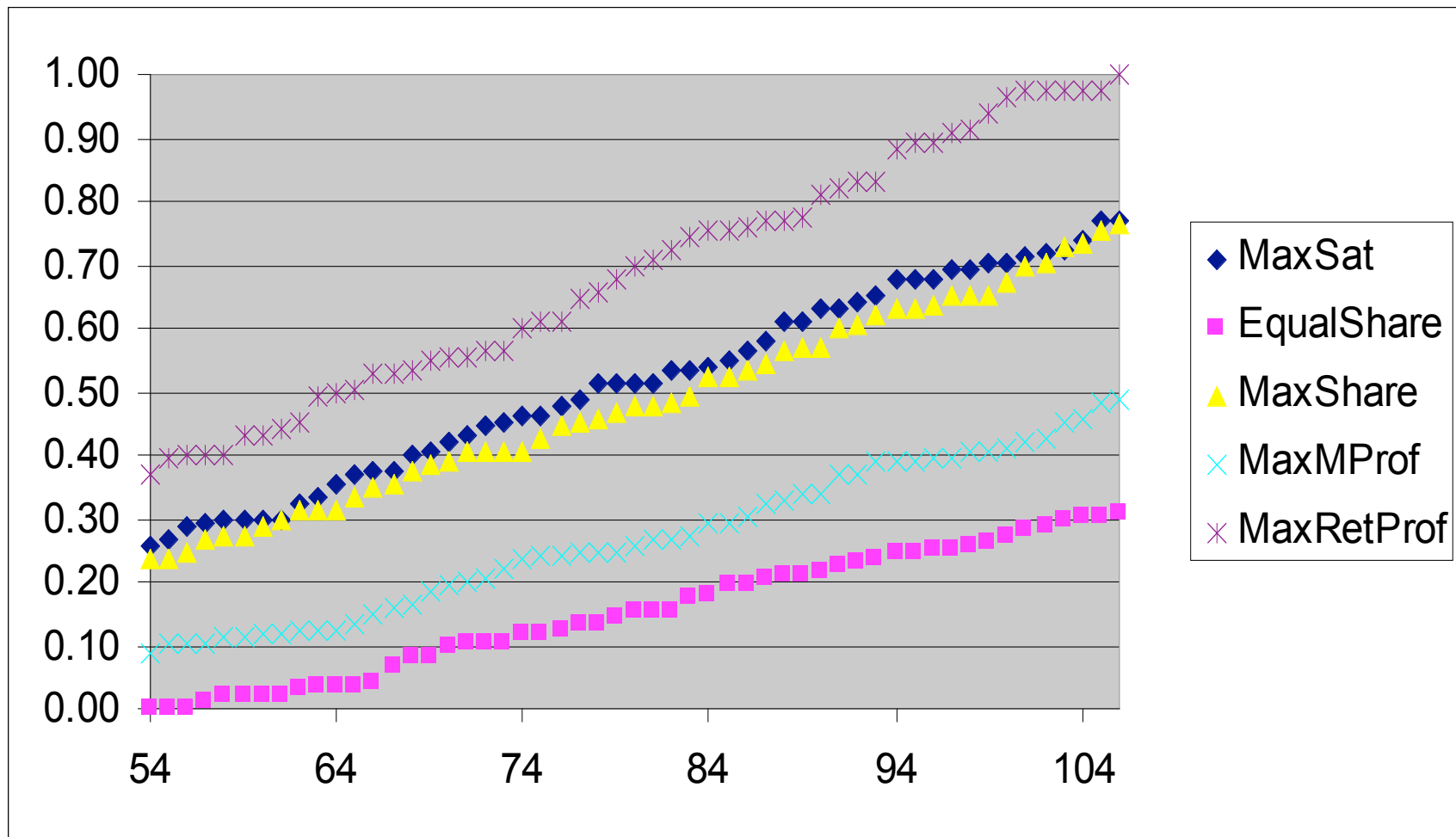




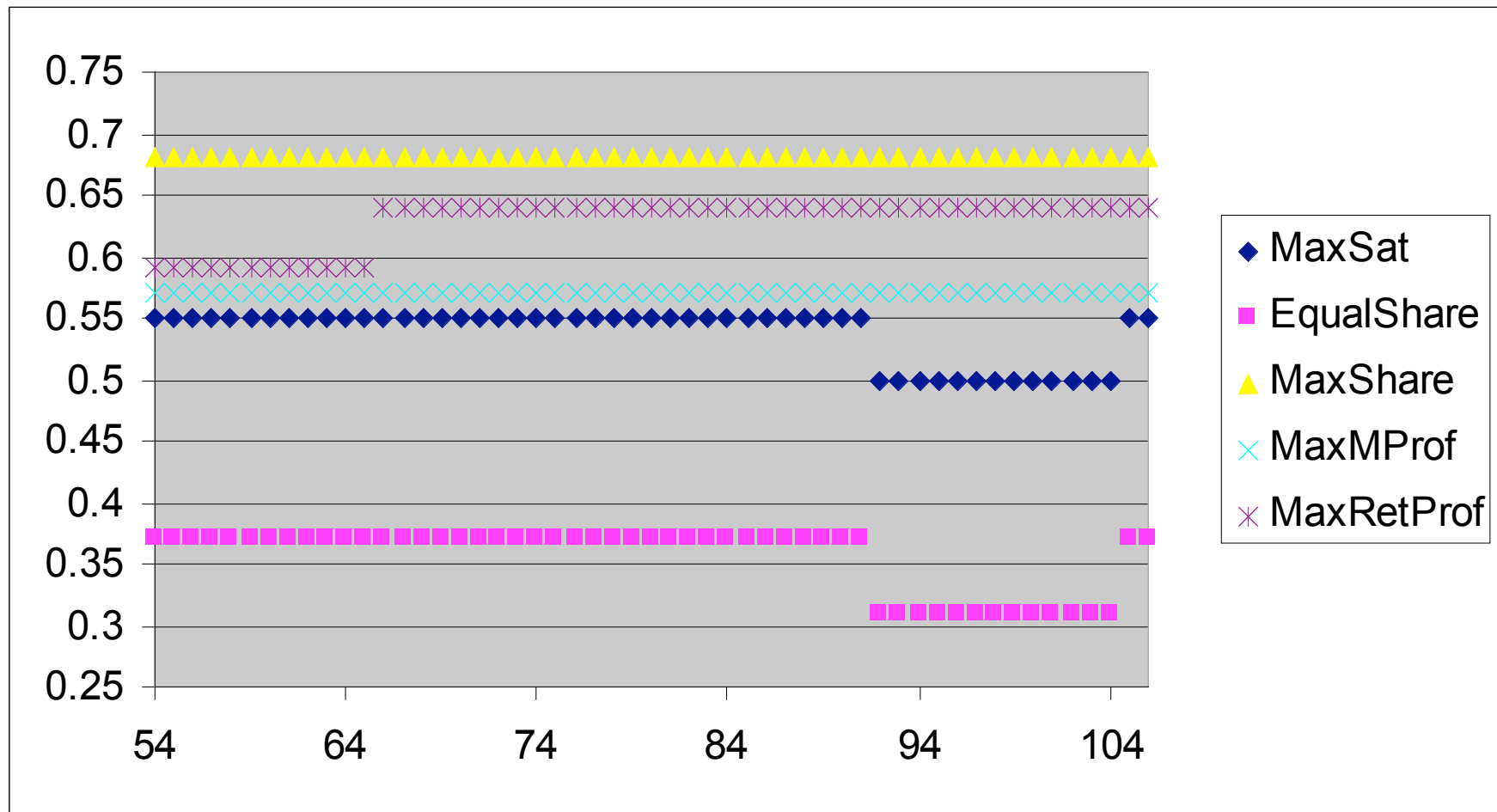
# Profits for the leading brand under 5 criteria



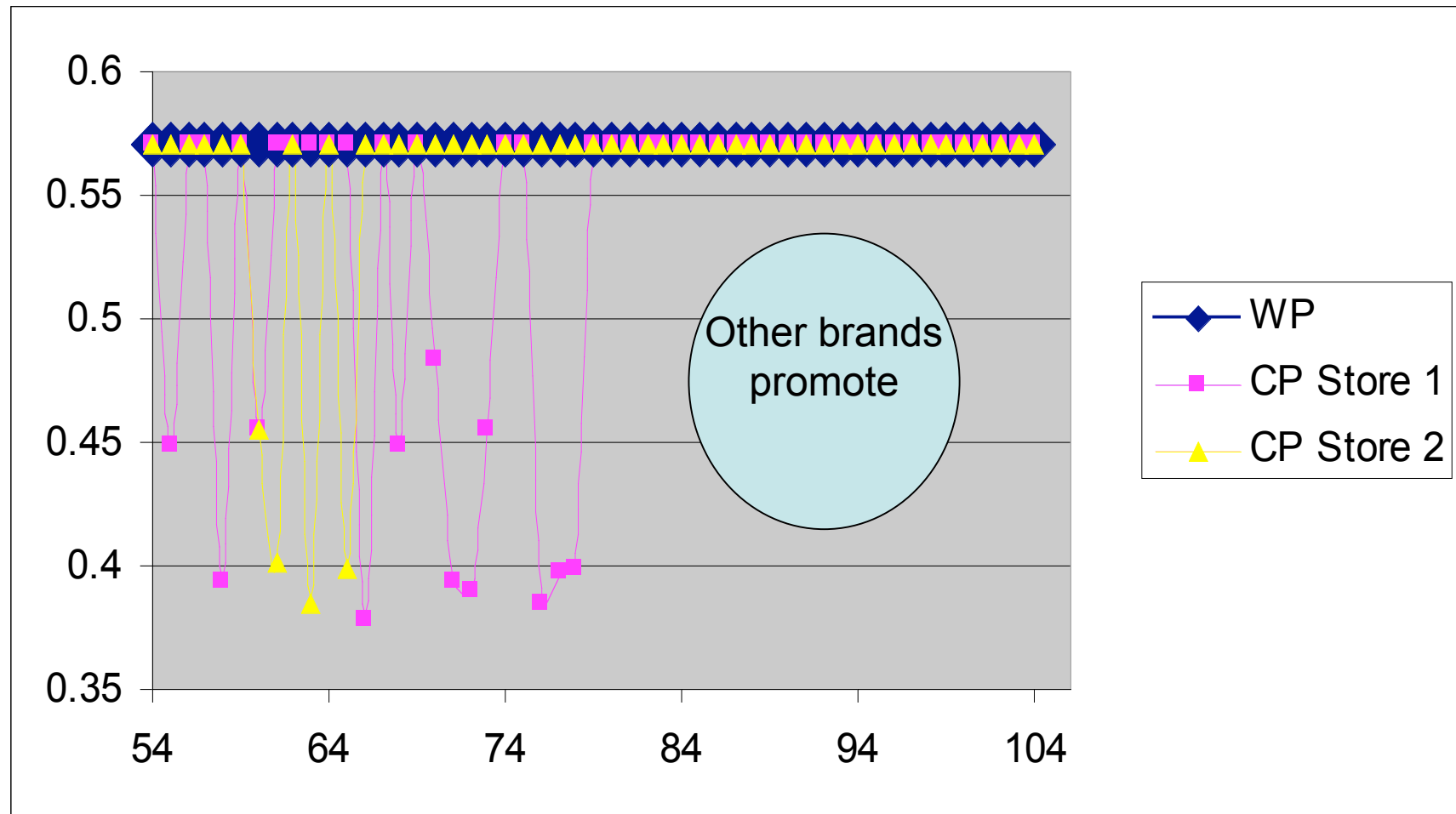
# Profits for the leading retailer under 5 criteria



# Wholesale price for leading brand under 5 criteria



# Consumer price in the two stores for leading brand under MaxMprof



# Observations and Insights from Destructive Testing?

## Modelling weaknesses revealed?

- I. **Rising Consumer Satisfaction with post-burn-in ticks 54–100, while prices level and purchases are constrained. Why?**
  - **No satiation or even diminishing marginal utility modelled.**
    - $\therefore$  Affect unlimited  $\rightarrow$  positive feedback.**
    - $\therefore$  Satisfaction unlimited.**
  - **No budget constraints on consumer agents.**

**Fitting consumer behaviour to historical data:  $\therefore$  constrained.**

## Insights continued

- 2. Consumer Satisfaction from brand X is higher after GA optimization to max MaxBrandProfits than to max MaxConsumerSatisfaction.**

**i.e. Perverse behaviour of model.**

- Let's review after validation → better parameter and variable values.**

**Is this behaviour of the model an artifact of unrealistic numbers into the model?**

## Validation

- **Have panel data of a sample of households.**
- **And check-out scanner data for two stores and eight brands.**
- **Develop 4 stereotypical agent types for Consumers (deals with heterogeneity).**
- **Micro-calibrate these agents from the panel data and analysis.**
- **Fit the remaining parameters by minimizing differences (MSE) between 10 outputs from simulation and actual store data (store and brand revenues).**
- **Again using the GA to optimize.**

## Issues in Assurance

- **Increasing measures in destructive testing.**
- **Satisfaction highest with MaxBrandProfit, not MaxConsumerSatisfaction.**
- **Scaling between the model and the universe.**
- **GA convergence in validation (difficult).**
- **Can we somehow weight consumer agents so that we don't need 10,000, which may be impossible to optimize?**
- **Mapping real-world prices onto a simpler frame.**
- **Need for a additional “light buyer” agent**



## References

- [1] **Haefner J.W. (2005), *Modeling Biological Systems: Principles and Applications*, New York: Springer, 2nd ed.**
- [2] **Marks, R.E. (2007) Validating Simulation Models: A General Framework and Four Applied Examples. *Computational Economics*. 30(3): 265–290, October. <http://www.agsm.edu.au/~bobm/papers/sl.pdf>**
- [3] **Midgley D.F., Marks R.E., and Kunchamwar D. (2007) The Building and Assurance of Agent-Based Models: An Example and Challenge to the Field, *Journal of Business Research*, Special Issue: Complexities in Markets, 60: 884–893. <http://www.agsm.edu.au/~bobm/papers/Midgley-Marks-Kunchamwar.pdf>**
- [4] **Miller J., Active nonlinear tests (ANTs) of complex simulations models. *Manage Sci* 1998, 44(6):820–30.**