Midterm Exam Microeconomics: 2021

Victor H. Aguiar

This version: 2021

Problem 1. (30 points) Let X be a nonempty and finite set of alternatives. Consider the following bounded rational consumer. Given a menu $A \subseteq X$, the consumer searches sequentially through the menu according to a fixed search order on X. At each step in the search process, the consumer compares the utility of the item that she sees to a fixed threshold. If the utility of the item is above the threshold then the consumer stops the search and picks the item. Formally, let $S \subseteq X \times X$ be a linear order on X, let $u: X \to \mathbb{R}$ be a utility function, and $u^* \in \mathbb{R}$ be a threshold. The choice correspondence of this bounded rational consumer is given by

$$c_S(A) = \{b \in A | u(b) \ge u^*; \text{ there exists no } aSb; u(a) \ge u^* \}$$

for all $A \subseteq X$, where the subscript S denotes the dependence of the choice correspondence of the linear order that models the search process.

- (a) Show that the choice correspondence c_S defined above is in fact a choice function.
- (b) Show that the choice correspondence c_S satisfies WARP (define WARP using the notes and textbook).
- (c) Assume you observe a dataset $(2^X \setminus \emptyset, c)$ (such that $c: 2^X \setminus \emptyset \to 2^X \setminus \emptyset$ is an observed choice function -possibly different from c_{BR} -) that satisfies WARP. Show that there is a bounded rational consumer characterized by a triple (S, u, u^*) a linear order on X, a utility function and a thresholds as in (a) such that $c(A) = c_{BR}(A)$ for all $A \in 2^X \setminus \emptyset$.
- d) Assume now that there is a distribution over search orders. Denote the collection of all search orders/linear orders on $X \times X$ by S. The distribution over search orders is $\pi \in \Delta(S)$. The probability of choice of item ain menu A such is given by $\rho_A(a)$ where $a \in A$ and $\rho_A \in \Delta(A)$. Let the stochastic version of the bounded rational model above generate ρ_A such that:

$$\rho_A(a) = \sum_{S \in \mathcal{S}} \pi(S) 1(a = c_S(A)).$$

Show that if you have two menus $A \subseteq B$ and $a \in A \cap B$, it has to be that $\rho_B(a) \leq \rho_B(a)$ (i.e., regularity).

Problem 2. (30 points) Let X be a nonempty and finite set of alternatives. Consider the following stochastic bounded rational consumer called "stochastic limited consideration (SLC)".

Assume there is a default alternative $o \notin X$, that is always present in any given menu. Let $X^* = X \cup \{o\}$. Also let the collection of menus that contain o be defined as \mathcal{A} , such that $A \in \mathcal{A}$ if $A \subseteq X^*$ and $o \in A$. Given a menu $A \in \mathcal{A}$, the probability of this consumer of choosing an alternative a in menu A, denoted by $\rho_A(a)$, where $\rho_A \in \Delta(A)$, is given by the following decision algorithm: First the consumer is endowed with a strict preference relation \succ defined over X^* with the restriction that for any $x \in X$, it must be that $x \succ o$ (i.e., the default is the worst item). The consumer is also endowed with a distribution over mental categories. Formally, mental categories, \mathcal{D} , is the collection all subsets of X^* . At each decision trial, the consumer draws a mental category from \mathcal{D} with probability $m \in \Delta(\mathcal{D})$ with the restriction that m(D) = 0 if $o \notin D$ (i.e., the probability of drawing a category that does not have the default is zero). Then the consumer forms a "consideration set" by taking the intersection of menu A and D such that the consideration set is equal to $D \cap A$. Then the consumer maximizes her preferences \succ on the consideration set $D \cap A$. Finally, she chooses the item that maximizes her preferences on $D \cap A$. The resulting probability of this decision algorithm is:

$$\rho_A(a) = \sum_{D \in \mathcal{D}} m(D) 1(a \succ b \forall b \in (D \cap A), b \neq a),$$

where $1(\cdot) = 1$ when the argument is true and zero otherwise.

- A complete stochastic dataset, ρ , is a collection of ρ_A for all menus $A \in \mathcal{A}$.
- We say that a is stochastically revealed preferred to b (i.e., aRb) when $p(b, A \cup \{a\}) < p(b, A)$.
- We say R is a acyclic if there is no integer $n \geq 2$ and $a_1, \dots, a_n \in X$ such that $a_i \succ a_{i+1}$ for $i = 1, \dots, n-1$ and $a_n \succ a_1$.
- We say that a complete stochastic dataset, ρ , is regular if for any pair of menus $A, B \in \mathcal{A}$ such that $A \subseteq B$ it must be that $\rho_B(a) \leq p_A(a)$ for any $a \in A \cap B$.
- a) Show that a complete stochastic dataset ρ that is generated by a SLC, as described above, implies that the stochastic revealed preference relation R is acyclic.
- b) Show that a complete stochastic dataset ρ that is generated by a SLC, implies that ρ is regular.
- c) Let ρ be generated by a SLC with the additional restriction that $m(X^*)=1$ and zero otherwise. This means that this consumer always considers all alternatives.

Show that the choice correspondence defined by:

$$c(A) = \{ a \in A : \rho_A(a) = 1 \},\$$

is a choice function (i.e., single-valued), and that c(A) satisfies SARP (first define SARP as seen in class and in the problem sets).

d) Show that a complete stochastic dataset ρ that is generated by a SLC, satisfies the ASRP (Axiom of Stochastic Revealed Preference) as defined in the notes. (Hint: use the equivalence theorem between the ASRP and random utility and define ASRP using the notes.).

Problem 3. (Modeling question). (30 points) Modeling the "Endowment Effect". I am going to present to you a summary of an experiment showing a behavioral bias called the Endowment Effect. The task is to write a decision model algorithm to explain the Endowment Effect.

The Endowment Effect Experiment.

Kahneman, Knetch and Thaler [1990], did an experiment with the following features:

- 44 subjects
- 22 subjects given mugs
- The other 22 subjects given nothing
- Subjects who owned mugs asked to announce the price at which they would be prepared to sell mug
- Subjects who did not own mug announced price at which they are prepared to buy mug
- Experimenter figured out prices at which supply of mugs equals demand.

If our subjects are rational consumers then the prediction in this market is:

- Prediction for rationality: As mugs are distributed randomly, we should expect half the mugs (11) to get traded.
- Explanation: Consider the group of mug lovers (i.e. those that have valuation above the median), of which there are 22.

- Half of these should have mugs, and half should not.
- The 11 mug haters that have mugs should trade with the 11 mug lovers that do not.

The Experiment Outcome however differs from the Prediction, and we call this the Endowment Effect.

- In 4 sessions, the number of trades was 4,1,2 and 2 (respectively per session).
- Median seller valued mug at \$5.25
- Median buyer valued mug at \$2.75
- Willingness to pay/willingness to accept gap
- Subject's preferences seem to be affected by whether or not their reference point was owning the mug.
- 1. Explain why the Endowment Effect is not consistent with a traditional model of rationality. Hint: The answer has less to do with the idea of revealed preferences, but more with the idea of stability of preferences. Write down the model of trade formally. You can assume a parametric distribution of preferences, make as simple as possible (uniform distribution).
- 2. Propose a decision algorithm that explains the endowment effect given the trading scheme of the experiment.
- 3. Calibrate your model parameters to reproduce the experiment outcome.

Grading Criterion

- 1. Work group is not allowed, if I find out that several of you have essentially the same model, you will receive the total grade of this question divided by the number of students that share the same response. Same goes for the answers to the other two question. I expect that each person has their own individual answers.
- 2. Every question indicates how much points they are worth out of 90.
- 3. The last question will be graded on the following 5 items: Logical consistency of the decision algorithm. Clarity of the explanation of the decision algorithm. Degree of success in explaining the endowment effect (calibration). Clarity in the explanation of the endowment effect using the decision algorithm. Creativity and degree to which you have included topics that we covered in class.