# Psychophysical Interpretation for Utility Measures 

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#### Abstract

The paper explores utility measures by combining experiments with mathematical derivations in psychophysics paradigm. The analysis on ultimatum game experiment reveals an evidence for utility threshold and thus supports Bernoulli's utility logarithmic law. Both experimental results and theoretical derivations show that the logarithmic law is suitable for the description of commodity choice and the power law for risk choice. The further mathematical demonstration indicates the logarithmic law for utility scaling to be a Klein-Rubin utility function, a utility function well defined in microeconomics. Based on this, the experimental utility measure is connected with the econometric model Linear Expenditure System, and presents an experimental procedure for testing the utility maximization hypothesis, which will remove a long unsettled perplexity in a fundamental stone of economics since Gossen proposed it in 1854.


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Keywords Psychophysics; ultimatum game; utility function; logarithmic law; power law

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## 1 Introduction

The paper seeks to achieve four goals: (i) reveal correspondences between sensation threshold in psychophysics and responder's acceptance threshold in the ultimatum game; (ii) experimentally and mathematically verify that utility is an observable psychological magnitude that agrees with the form of sensation scales in psychophysics; (iii) derive the quantitative attributes of commodity choice and risk choice; iv) present an experimental procedure to test utility maximization hypothesis.

The first highlight of this paper is the shape of utility function. It is based on other researchers' three earlier studies: Bernoulli's derivation for the logarithmic function of utility (Bernoulli, 1738), Kahneman and Tversky's power function for money-risk utility estimation proposed in their experimental analysis on the cumulative prospect theory (Kahneman and Tversky, 1992), and Ekmen's mathematical inference of deriving the power law from the logarithmic law in psychophysics (Ekmen, 1964).

To deal with St. Petersburg paradox in a gamble (e.g. see Stigler, 1950), Swiss mathematician D. Bernoulli assumed in 1738 that the increase of one's subjective economic utility $d U$ is directly proportional to the increment of wealth $d x$ and inversely proportional to one's whole wealth $x$, or $d U=k d x / x$. By integrating on both sides, he derived a logarithmic law $U=k \ln x+C$ for the utility (Bernoulli, 1738). More than one hundred years later, this result agreed with Fechner's logarithmic law for the sensory response of physical stimulus. This was once regarded as an evidence of utility judgment to be a kind of psychophysical response by many researchers including Marshall and Fechner (Stigler, 1950). Evidently, Bernoulli's above derivation foundation is incomplete. To support his hypothesis, an empirical Weber fraction $\Delta x / x$ in utility judgment is required. It can only be resolved by empirical or experimental studies. Bernoulli did not continue and finally finish empirical analyses for his hypothesis.

Differing from the suggestion of Bernoulli's logarithmic function, Kahneman and Tversky in their experimental analysis for the cumulative prospect theory found that it is appropriate if a power function is used to describe subjects' money-risk utility estimating data (Kahneman and Tversky, 1992). It is only a direct experimental observation outcome but not a complete discussion, lacking theoretical support, and attracts little attention in existing literatures.

Responding to the debate on the alternatives of Fechner's logarithmic law and Stevens' power law in psychophysics, Ekmen supposed the logarithmic law commonly applicable in psychophysical judgments, then theoretically introduced the logarithmic descriptions simultaneously into both the numeric judgment and physical stimulus judgment in Stevens' magnitude estimate approach (e.g., Stevens, 1959; Stevens and Guirao, 1963), and finally obtained a mathematical relation in which the power law could be derived from the logarithmic law (Ekmen, 1964). Nevertheless, Ekmen only discussed his formulations theoretically without any experimental test. It left an attractive inspiration to us rather than a final resolution to the debate.

Though the above three analyses all are incomplete as conclusive discussions either in theory or experiment, they at least give us such an impression that there may exist two alternatives for the shape of utility functions, one logarithmic and another power, and in Ekmen's analytical framework, the two alternatives may be logically consistent but not conflicting.

This paper will remedy the defects among the above three studies and integrate them into a logically consistent description, namely, reveal an experimental basis for Bernoulli's hypothesis, clear theoretical elements for Kahneman and Tversky's power function in money-risk utility estimation, and
introduce experimental decomposition to test Ekmen's formulations.
Combining psychophysics approach with the evidence from the ultimatum game experiment, Section 2 will reveal an experimental basis for Bernoulli's hypothesis, and discuss the commodity choice and risk choice in a new framework of utility scaling. In this analysis, the ultimatum game experiment is given a new sense of utility threshold measurement, a typical psychophysics interpretation. In this section, a mathematical demonstration will identify the logarithmic law with Klein-Rubin utility function (Klein \& Rubin, 1947) in a new experimental scheme, and finish the key preparation for setting up an experimental procedure to test the utility maximization hypothesis.

In Section 3, by discriminating two kinds of estimates, single estimate and double estimate, in experimental utility measures, commodity choice and risk choice will be logically related in a utility scaling framework, and an experimental test to Ekmen' formulations will be naturally introduced into the discussions.

In the other hand, there are two traditional paths to treat the issue of utility measures. One is the individual utility measure, and another the average utility measure. The former is usually followed by researchers who probe the features of utility concept itself, often seen in most of experimental studies (e.g., Fisher, 1892; Mosteller and Nogee, 1951; MacCrimmon and Toda, 1969; Kahneman and Tversky, 1979; Starmer, 2004), and the latter is usually followed by those who use econometric models to describe realistic markets, often seen in applied studies of market empirical data (e.g., Klein and Rubin, 1947; Stone, 1954; Liuch, 1973; Houthakker, 1960; Theil, 1965; Deaton and Muellbauer, 1980). The two paths separately stretch in relatively independent two fields of experimental and econometric studies, and cause isolation between them.

There were some debates on the average utility measures. The pioneer of experimental utility measurement, psychologist Thurstone firstly used average utility measure to determine the experimental indifference curve (Thurstone, 1931), but was criticized by other researchers (e.g., Wallis and Friedman, 1942). The criticism seems continuing today. Pooling all individuals' data and then fitting the model to the pooled data is regarded may misrepresenting the true utility functions for individuals. For example, one can approximate a power function by combining exponential functions of different rates, furthermore, even if the form of the utility function is the same for every subject, the subjects may differ in parameters. Such a worry may be reasonable if one uses an average result to describe individual judgment. However, if the average result is limited to apply to describing group behaviors, for example, market empirical data, the above worry will be unnecessary. Even if the individuals' utilities follow exponential functions of different rates but the best asymptote for their average result is power function, we should use the power function but not the exponential function to depict the overall effect of all individuals in a market or an experiment, in which both exponential and power functions may be true but respectively for either individual or group. In this case, we should also forbid using an individual utility function to describe group behavior. Whether or not a utility report has economics sense depends on whether or not it presents a regular and stable result, as a regular and stable result will contribute a meaningful description in economics.

This paper is an experimental study but follows the path of average utility measure, agreeing with the psychophysics and econometrics paradigm. Namely, it will determine parameters in a utility function by using averages of subjects' performances in experiments. Another highlights in this paper is Klein-Rubin utility function (Klein and Rubin, 1947) based on average measures, which has been demonstrated in Section 2 as a psychophysical function for utility and is given a behavioral economics sense. This will provide a realizable framework to estimate both the econometric model Linear

Expenditure System (Stone, 1954) and the utility function in psychophysical paradigm simultaneously in a set of experimental data, and further, to form a new way to experimentally test utility maximization hypothesis, which is so fundamental and important but has never been tested in experiments since Gossen proposed it in 1854.

Section 4 summarizes and discusses the findings in this paper, including an experimental procedure for testing the utility maximization hypothesis.

## 2 Experimental analyses to the utility threshold in ultimatum game

### 2.1 Weber fraction in the ultimatum game

Bernoulli's discussion in 1738 means Fechner's assumption of just noticeable difference (JND) (Fechner, 1860) unnecessary, and implies another mathematical path of deriving the logarithmic law directly from empirical Weber fraction (Masin, Zudini, and Antonelli, 2009). That is, as long as observations of utility scaling behavior supports a Weber fraction, using Bernoulli's mathematical derivation, the logarithmic law will be possibly derived as an empirical law.

In psychophysics, Weber's law states that in the moderate stimulus intensity, the amount of change needed for the difference threshold $\Delta I$ is a constant fraction $K$ of the initially presented physical stimulus intensity $I$, that is, $\Delta I / I=K$. Following Weber's discovery, Fechner related a least sensation variation $\Delta S$ to $K$, and obtained a relationship between the subjective sensation scaling $S$ and the objective stimulus intensity $I$. Fechner's result $S=c \ln I+C$ can also be derived similarly by the derivations used by Bernoulli (1738), where, $c$ and $C$ are two constants.

In a typical ultimatum game experiment (e.g., see Güth, 1995), there are two players, one plays proposer and the other responder, to distribute a "stake" (usually certain amount of money) between them. Both sides know the money stake $q$. At first, the proposer controls $q$. And then, the proposer proposes to the responder an ultimatum division of the stake, in which the proposer offers to the responder a money amount $q_{o}$ cutting out from $q$, and retains $q-q_{o}$ to himself. Then the responder determines to accept or reject it. If the responder accepts, he gains $q_{o}$, and the proposer gains $q-q_{0}$. If the responder rejects, both sides obtain nothing. Usually for the proportion $q_{o} / q$ there is a proportional rejection threshold $q_{o r} / q$. Once the proportional offer is below such a threshold, the responder would rather probably reject the proposer's proposal. After investigating the results from sixteen research groups, Camerer found that half of proposers' offers were rejected by responders below $q_{o} / q=0.2$ (Camerer, 2003). This proportional rejection threshold is similar to that Weber fraction reveals in psychophysics.

As far as their corresponding relations, components of the subjective measure in ultimatum game are completely identical to those leading us to derive Fechner's logarithmic law from sensation response of physical stimulus. Those components include: (i) standard stimulus, compared stimulus, absolute difference threshold, and they give the evidence to construct $\Delta S=\mathrm{c} \Delta I / I$; and (ii) a constant Weber's fraction, which supports an integral on both sides of $\Delta S=\mathrm{c} \Delta I / I$ and finally derives the logarithmic law by Bernoulli's mathematical path.

In one of the difference threshold measurements for physical stimulus, a standard stimulus $I$ is given, then a compared stimulus $I_{c}$ is presented. The intensity of $I$ keep constant, while the intensity of $I_{c}$ is adjusted. The difference between $I_{c}$ and $I$ is gradually diminished and approaches zero. And the subject is asked to compare the difference between $I_{c}$ and $I$. In the adjusting course of $I_{c}$, there statistically is a critical intensity level $I_{c c} \neq I$, at which the subject reports the difference beginning to
vanish between $I_{c}$ and $I ; \Delta I=\left|I-I_{c c}\right|$ is the absolute difference threshold, and Weber's fraction $\Delta I / I$ is determined too (e.g. see Luce and Krumhansl, 1988).

In an ultimatum game, denote the proposer's offer $q_{o r}$ as the responder's rejection threshold, below which a responder rather probably rejects proposer's proposal, while, above which rather probably accepts it. In the ultimatum game experiment, a responder always faces to three "stimuli": (i) money stake $q$; (ii) offer $q_{o}$; and (iii) the part a proposer retains for himself $q-q_{o}$. Here, in $q_{o}$ and $q-q_{o}$ only one is independent. The responder's utility estimates in an ultimatum game are interpreted as follows: A responder's utility estimate to the offer is a comprehensive result involving various judgments, including the money amount, unfairness, and so on. In the responder's utility estimate, if $q_{o}<q_{o r}$, the proposer's retained part $q-q_{o}$ will be regarded unfair, the responder's loss from unfairness will exceed his gain from $q_{o}$, the utility is null, and the offer will be rejected; if $q_{o}>q_{o r}$, the proposer's retained part $q-q_{o}$ will be regarded acceptable, the responder's gain from $q_{o}$ will exceed his loss from unfairness, the utility is positive, and the offer will be accepted. Comparing with the measurement to difference threshold of physical stimulus, $q$ is a standard stimulus, $q-q_{o}$ a compared stimulus, and $q_{o r}$ the absolute difference threshold. If $q_{o}<q_{o r}$, the responder feels no utility difference between $q$ and $q-q_{o}$ (for the responder, it is equivalent to proposer occupying all money stake). It corresponds the difference vanishing between $I_{c}$ and $I$ in the psychophysical measurement. Reversely, if $q_{o}>q_{o r}$, the responder feels the difference. It corresponds to a difference between $I_{c}$ and $I$ being detected in the psychophysical measurement. Here, a one-to-one correspondence exists between sensation threshold in psychophysics and responder's acceptance threshold in ultimatum game. The above explanation is called "utility scaling interpretation" for the ultimatum game experiment.

In fact, Fechner's logarithmic law in psychophysics depends on the primary relationship

$$
\Delta S=c \frac{\Delta I}{I} .
$$

Let $S$ indicate a responder's subjective utility estimate in the ultimatum game experiment, then a similar relationship

$$
\Delta S=c \frac{q_{o r}}{q}
$$

holds in the utility scaling interpretation, where $q_{o r}$ identifies with $\Delta I$. Naturally, the ratio rejection threshold $q_{o r} / q$ corresponds to Weber's fraction $\Delta I / I$.

The above analogue has revealed the structural correspondence between $q_{o r} / q$ and Weber's fraction. The further problem is whether or not such a ratio $q_{o r} / q$ is stable in a specific ultimatum game. In 1994, Cameron had worked out an experimental study on the simple one-period ultimatum game with raising stakes (Cameron, 1999), from which we are able to derive a test to the stability of rejection threshold at different monetary levels.

### 2.2 Examination to Bernoulli's hypothesis

Cameron conducted experiments in Indonesia by using real Indonesia money, the Rupiah, as stakes at three levels of $5,000,40,000$, and 200,000 , about $0.075,0.6$, and 3.0 times the average monthly expenditure of a participant. this is a maximal span between low and high stakes that had been used in ultimatum game experiments. Three groups of subjects respectively participated in 5,000-, 40,000-, and 200,000-experiments. Cameron tested their homogeneity by simple $t$-tests and no significant difference was detected among them. Their results can be compared with each other. Only taking into account unrepeated games, there are three 5,000-experiments, one 40,000 -experiment, and one

200,000-experiment provided in Cameron's study. As the statistical test in Cameron's paper shows no significant difference among three 5,000-experiments, their data will be pooled. We will concentrate on the responders' switching points from acceptance to rejection in ultimatum games with different stakes. It is sufficient for us to discuss the results of $q_{o} / q$ below $50 \%$.

Table 1. Data of rejection and offer in Cameron's experiment
$\mathrm{n}=101$ pairs in three 5,000 -experiments; $\mathrm{n}=35$ pairs in 40,000 -experiment; $\mathrm{n}=37$ pairs in 200,000-experiment

|  |  | Proposer's proportional offer interval $\left(p_{o} / p\right)$ |  |  |  |  |  |
| :--- | :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| Stake |  | $0-0.09$ | $0.1-0.19$ | $0.2-0.29$ | $0.3-0.39$ | $0.4-0.49$ | $0.5-0.59$ |
| 5,000 | Rejection rate | $100 \%$ | $50 \%$ | $55 \%$ | $20 \%$ | $19 \%$ | $3 \%$ |
|  | Number of proposals | 4 | 2 | 11 | 5 | 31 | 40 |
| 40,000 | Rejection rate |  |  | $25 \%$ | $29 \%$ | $0 \%$ | $0 \%$ |
|  | Number of proposals |  |  | 4 | 7 | 3 | 20 |
| 200,000 | Rejection rate | $100 \%$ | $100 \%$ | $50 \%$ | $0 \%$ | $0 \%$ | $0 \%$ |
|  | Number of proposals | 1 | 1 | 2 | 3 | 11 | 16 |



Fig. 1. Proportional rejection thresholds of 5,000 -experiment and 200,000-experimrnt. There is little difference between them.

In Cameron's 5,000- and 200,000-experiments, proposers offered $q_{d} / q$ from $0 \%$ through $50 \%$ stakes; the switching points will be shown in these experiments directly. In 40,000 -experiment, however, no offers were given by proposers below $20 \%$ stake, and no switching point can be observed in this case. Hence, only the results of 5,000- and 200,000-experiments can be compared. Cameron's experimental data are rearranged in Table 1, which is estimated (after discarding invalid data) from "FIGUREs 1, 2, and 3" in Cameron's paper (1999). In Table 1, the percentage of responder's rejections to proposer's proposals is described in six proportional offer intervals " $0 \sim 0.09$ ", " $0.1 \sim 0.19$ ", " $0.2 \sim 0.29$ ", " $0.3 \sim 0.39$ ", " $0.4 \sim 0.49$ ", and " $0.5 \sim 0.59$ "; "Number of proposals" indicates the number of proposers who propose to responders in a specific proportional offer interval; and "Rejection rate" indicates the percentage of proposer's offers rejected by responders in a specific proportional offer interval. For instance, in the bottom-right in Table 1, " 16 " indicates sixteen proposers making
proportional offers within the interval $0.5 \sim 0.59$, and above " 16 ", " $6 \%$ " indicates six percent of the sixteen offers being rejected by responders. Results of 5,000- and 200,000-experiments in Table 1 are plotted in Fig. 1. The rejection threshold is defined to be the value of proportional offer that is rejected exactly 50 percent of the time, similar to that in a Yes-No detection procedure in psychophysics (e.g. Gescheider, 1976). In Fig. 1, the solid line describes 5,000-experiment, and the dashed line 200,000-experiment. They are the results of linear regression on data between proportional offers of 0.5 and 1.0. Fig. 1 shows that the rejection threshold is 0.245 for 5,000 -experiment and 0.25 for 200,000 -experiment. They are so close that they can be viewed as identical. That is, Cameron's experiment supports Bernoulli's hypothesis, and the logarithmic law could exist in the utility scaling.

Although the above discussion involves only small data set in its key point and cannot be regarded as ultimate evidence, doubtlessly, it first delivers a direct positive proof for Bernoulli's hypothesis since 1738. At least, with this proof, Bernoulli's utility logarithmic law represents not only purely a theoretical imagination but also an attractive possible reality. The further conclusion, of course, will be determined by a combination with experimental tests in Section 3.

List and Cherry (2000), Slonim and Roth (1998), and Roth, et al. (1991) also compared frequencies of responders' rejection behavior for low and high stakes in ultimatum games. List and Cherry used stakes $\$ 20$ and $\$ 400$, Slonim and Roth used Slovak Crown as stakes at three levels Sk60, Sk300, and Sk1500, and Roth et al. used $\$ 10$ and $\$ 30$ in their first round experiment for Pittsburgh subjects. Their first-round tests can be viewed as a simple one-period ultimatum game, equivalent to the type of Cameron's experiment. Also investigating affections from low and high stakes in simple one-period games, Straube and Murnighan (1995) used stakes $\$ 5$ and $\$ 100$, and Hoffman, MacCabe, and Smith (1996) \$10 and \$100. Their outcomes show only little different effects on subjects' rejection behavior for low and high stakes in the simple one-period ultimatum game. The stability of ratio rejection threshold in simple one-period ultimatum game is a general phenomenon.

In addition, in repeated game studies, List and Cherry (2000) and Slonim and Roth (1998) found significant difference between low and high stakes in the rejection threshold of experienced participants (e.g. in the tenth round). The learning effect accounts for the behavioral change from the simple one-period ultimatum game to repeated ultimatum game: an experienced subject understands how to strategically take more benefit or diminish loss from his opponent (Roth, et al., 1991). It therefore does not deny Bernoulli's hypothesis. Repeated games describe strategic manipulations rather than intuitive performances. To reveal intuitive features, studying subjects' performance in the simple one-period cases is more appropriate.

### 2.3 Discussions on the logarithmic law of utility

To overcome the failure of Weber's law in very low stimulus intensity, a revised form of Weber's fraction $\Delta V(I+a)=K$ was proposed in psychophysics, where $a$ is a constant parameter (see Gescheider, 1997). Accordingly, the revised Fechner's law should be as follows:

$$
\begin{equation*}
S=c \ln (I+a)+C \tag{1}
\end{equation*}
$$

To tally with the conventions in economics, (1) is rewritten as

$$
\begin{equation*}
S=c \ln (q+a)+C, \tag{2}
\end{equation*}
$$

where $q$ is the economic quantity, and $c, a$, and $C$ are constants determined by experiment. Evidently, compared with the logarithmic law in psychophysics, the subjective scaling $S$ in (2) has a meaning of intuition scaling of economic quantity $q$. It is similar to the sensation scaling of physical stimulus. As there is no particular modality for utility judgment, the word "intuition" is used here to replace
"sensation". However, whether it is sensation or intuition, stimulus-response is their common attribute.
The sensation measurement in psychophysics and the utility scaling in ultimatum game are identified in their components and quantitative stabilities, and present a similar stimulus-response pattern that reveals an experimental foundation for Bernoulli's hypothesis. The logarithmic law should be generally suitable for the utility scaling.

### 2.3.1 Utility implication of the logarithmic law

If a logarithmic law exists in the above mentioned utility scaling, does it identify with a utility function well defined in microeconomics?

In econometrics, Linear Expenditure System (LES) (Stone, 1954) and Extended Linear Expenditure System (ELES) (Liuch, 1973) are two successful econometric models in empirical studies. They were mathematically derived by using Klein-Rubin utility function in consumer theory (e.g., see Liuch, 1973). Klein-Rubin utility function is

$$
U=\sum b_{i} \ln \left(q_{i}-r_{i}\right)
$$

Where, counting $q_{i}$ and $r_{i}$ by the monetary value, $b_{i}=\frac{q_{i}-r_{i}}{\sum\left(q_{i}-r_{i}\right)}, 0<b_{i} \leq 1, \sum b_{i}=1, q_{i}$ is the consuming quantity of commodity $i$, and $r_{i}$ the consumer's necessary consuming quantity of commodity $i$, a constant.

Measuring the commodity quantity by its monetary value, any commodity quantity $q_{i}, i=1,2,3, \ldots n$, can always be added up to a sum $Q$, or $Q=\sum q_{i}$. It can be demonstrated mathematically that the logarithmic law for $Q$ represents a Klein-Rubin utility function. Namely,

$$
c \ln (Q+a)+C=\sum b_{i} \ln \left(q_{i}-r_{i}\right)
$$

It is demonstrated as follows:
Taking differentials on the both sides of $Q=\sum q_{i}$,

$$
d Q=\sum d q_{i}
$$

$$
d(Q+a)=\sum d\left(q_{i}-r_{i}\right)
$$

where $a$ is a constant and $0 \leq r_{i} \leq q_{i}$ is a constant for $q_{i}$. Identically transform the equation

$$
\begin{align*}
& \frac{Q+a}{Q+a} d(Q+a)=\sum \frac{q_{i}-r_{i}}{q_{i}-r_{i}} d\left(q_{i}-r_{i}\right), \\
& (Q+a) d \ln (Q+a)=\sum\left(q_{i}-r_{i}\right) d \ln \left(q_{i}-r_{i}\right) \tag{3}
\end{align*}
$$

Set $\frac{\sum\left(q_{i}-r_{i}\right)}{Q+a}=\lambda$, then $\sum\left(q_{i}-r_{i}\right)=\lambda(Q+a)$. Denote $b_{i}=\frac{q_{i}-r_{i}}{\sum\left(q_{i}-r_{i}\right)}=\frac{q_{i}-r_{i}}{\lambda(Q+a)}$, which naturally satisfies $0<b_{i} \leq 1$ and $\sum b_{i}=1$. (3) is rewritten as

$$
\begin{equation*}
d \ln (Q+a)=\lambda \sum \frac{q_{i}-r_{i}}{\lambda(Q+a)} d \ln \left(q_{i}-r_{i}\right)=\lambda \sum b_{i} d \ln \left(q_{i}-r_{i}\right) \tag{4}
\end{equation*}
$$

When LES and ELES are derived from Lagrange's method, $b_{i}$ is a ratio, independent from $q_{i}$ (e.g., Liuch, 1973), thus, independent from $q_{i}-r_{i}$, and further from $\ln \left(q_{i}-r_{i}\right)$. Integrating on both sides in (4) under the condition suitable for the derivations of LES and ELES, finally get

$$
\begin{align*}
& \frac{1}{\lambda} \ln (Q+a)+C=\sum b_{i} \ln \left(q_{i}-r_{i}\right), \\
& c \ln (Q+a)+C=\sum b_{i} \ln \left(q_{i}-r_{i}\right), \tag{5}
\end{align*}
$$

where $c=1 / \lambda$.
In (3), the left-hand side is the logarithmic law for $Q$, and the right-hand side is a Klein-Rubin utility function. $c \ln (Q+a)+C$, the logarithmic law for $Q=\sum q_{i}$, is a Klein-Rubin utility function. Furthermore, measuring an economic quantity by its monetary value, an economic quantity can always theoretically be decomposed into $n$ parts $q_{i}, i=1,2,3, \cdots, n$. We are always able to construct $Q=\sum q_{i}$ for any economic quantity theoretically, and thus derive (5) for them. This means that the intuitive scaling value for an economic quantity, if it holds as a logarithmic law, always represents a Klein-Rubin utility function. Experiments indeed approve the utility scaling value to hold as logarithmic law (see Section 3, later). It is therefore concluded that such a logarithmic law obtained from psychophysical paradigm is of a utility description in economics. Klein-Rubin utility function implies the relationship between subjective and objective commodity quantities. In other words, the essential of utility maximization in LES and ELES is the maximization of subjective commodity quantities. Revealing the relationship between subjective and objective quantities is just the basic task of psychophysics. Psychophysical paradigm will play an important role in economic analysis.

In the above demonstration, no special restriction is made to the type of $q_{i}$. The demonstration holds for all economic quantities provided they are additive. Therefore, it is generally suitable for money amount, commodity quantity, risk degree, and so on.

We now interpret Klein-Rubin utility function clearly from the viewpoints of behavioral economics and the utility scaling interpretation: in this utility function, $\left(q_{i}-r_{i}\right)$ indicates the measures for economic quantities $q_{i}$ relative to the reference point $r_{i}$ (in LES and ELES, it is interpreted as the necessary consuming quantities to maintain one's basic life), and $\sum b_{i} \ln \left(q_{i}-r_{i}\right)$ is a linear combination of utility scaling values (logarithmic laws) for economic quantity $q_{i}$ compared to $r_{i}$.

A subjective scaling value of economic quantity is generally a utility value, and evidently, it is the cardinal utility. An ultimatum game experiment is essentially to measure the intuition threshold of utility.

There are two parallel processes of a responder's intuitions in an ultimatum game experiment. One is the objective observation of the number that tells the responder the objective amounts of money, and another, the subjective utility scaling that tells the responder the subjective utility obtained from proposer's offers. When $q_{o}$ is below $q_{o r}$, the former is still bigger than null but the latter diminishes to nothing. The utility scaling interpretation suits the latter. The responder's rejection behavior in the ultimatum game is a behavioral presentation of the utility scaling. Two independent processes are not mutually exclusive. The noticeable difference between $q$ and $q-q_{o}$ in objective observation of the number does not affect the vanishing of difference between $q$ and $q-q_{o}$ in utility scaling below the absolute difference threshold $q_{o r}$. As the two processes are always mixed together, the utility implication of a responder's intuition scaling of money amounts in an ultimatum game might be ignored for too long a time.

### 2.3.2 Single estimate and double estimate

The above demonstration relies on $Q=\sum q_{i}$. This relation as a perceptible object means that
subjects clearly know the sizes of $Q$ and $q_{i}$ and only the utility is required to estimate. It will be called "single estimate" below. Therefore, the demonstration contains that the logarithmic law as a utility function applies to the single estimate. Commodity choice is usually a single estimate.

If subjects are not informed the meaningful sizes of $Q$ and $q_{i}$, both quantity and utility need to be estimated, it will be called "double estimate". In the cases involving risk decision, one only possibly performs through the double estimate. For example, a kind of Allais' paradox used in Kahneman and Tversky's experiment is presented as the following (Kahneman and Tversky, 1979):

PROBLEM 1: Choose between
A: 4000 with probability 0.80 , or B: 3000 with certainty.
PROBLEM 2: Choose between
C: 4000 with probability 0.20 , or $\mathrm{D}: 3000$ with probability 0.25 .
Similar choices were also used by Kahneman and Tversky's experiment conducted for the cumulative prospect theory (Kahneman and Tversky, 1992). It is a money-risk estimate. Money-risk estimate is always a double estimate for the following reasons: 1) Clearly telling subjects the probability of a risk is unlikely to change their intuitive assessment into the precise quantitative calculation for an issue such as assessing worth of a risk, that is, it is still un-meaningful for estimating a quantity; and 2) there usually is no objective standard of market price for a risk in a trifling item such as a risk of losing a chance to get 3000 . In this case, one has to estimate subjective utilities of two factors simultaneously.

Kahneman and Tversky proposed that money-risk estimate follows the power law. In the light of the above analysis, the double estimate should follow power law. The next section will answer this question from theoretical and experimental analyses.

## 3 Experimental tests of single- and double-estimate utilities

After a pilot study (see Part 1 in Supplemental Files), 192 graduate student volunteers were recruited from Jinan University as subjects participating in experiments. Five simple experiments and one Latin square experiment (see Bailey, 1996) were carried out among those subjects. The Latin square experiment is usually used to reduce order effects revealed in simple experimental measures so that high-quality regression curves could be obtained.

Five simple experiments (with 80 subjects) tested the differences between subjects with and without economics-study backgrounds. They found no significant difference between the two types of subjects in their experimental performances, namely, subjects' behaviors observed in valid data have no reliance on the specialized knowledge in economics. These results and other interested outcomes of the five simple experiments are presented in Parts 2-4 of Supplemental Files. To save the pages, the coming text mainly discusses the Latin square experiment, briefly L-experiment, in detail. Detailed Latin square designs are presented in Part 5 of Supplemental Files.

### 3.1 Method

### 3.1.1 Participants

The L-experiment requires 60 subjects who deliver complete valid data in all measures. The subjects were divided into three groups, labeled Groups I $\sim$ III, and each of Group I $\sim$ III contains five subgroups, labeled Subgroups i $\sim$ v. Some subjects were bored by experimental measures and their measures had to be stopped midway and were therefore incomplete. There were totally 112 subjects (ages 22~25, majored in management or economics) randomly selected to enter the L-experiment, and

52 subjects failed to deliver complete valid data. In the final experimental results, each of Groups I~III contains 20 valid subjects. Every subgroup has 4 subjects (male 2, female 2). Experimental tests to single- and double-estimate utilities were performed by curve fits to the averages of the 60 valid subjects' data. To attract subjects, all participants were provided a free banquet or paid.

### 3.1.2 Measures

The test for each of Groups I~III consists of three measurements, which investigates subjects’ utility estimates respectively under three laboratory conditions of "quantity-price double estimate", "quantity single estimate", and "price single estimate".

Under the condition of quantity-price double estimate, labeled Meas. 1, subjects experience a sequence of consumptions, in which consuming quantities are unknown to them, and the prices are without existing recognized standards in practical markets. Subjects are asked to report money amounts they are willing to pay by estimating the consuming quantities through subjective assessments. This condition makes the utility scaling similar to one of Stevens' magnitude estimates in psychophysics, called cross-modality matching, a double estimate between the sensation scalings, for example, of handgrip and other physical stimuli such as loudness, brightness, visual length, vibration, electric shock, and so on (e.g., Stevens, 1959; Stevens and Guirao, 1963).

Under the condition of quantity single estimate, labeled Meas. 2, money amounts are clearly assigned to subjects for their consumption, and they are asked to estimate consuming quantities they think "should be" to match the money amounts.

Under the condition of price single estimate, labeled Meas. 3, subjects are clearly told about quantities consumed and asked to estimate the prices they are willing to pay.

Meas. 2 and 3 are decompositions from Meas. 1.
Electrical-power massage is used as consumption in the experiment. It is conducted through a portable electrical-power massage machine, which is handy for the experimenter to control. The duration of the electrical-power massage is the consuming quantity and the money a subject is willing to pay for such duration is the price chosen. Such an exact length of time is easily treated to be unknown to the subject. And, very importantly, till date in China, it is quite seldom to see commercial services of electrical-power massage, and the stimulus of electrical-power massage has no recognized standard market price. These satisfy conditions required by the measurement of double-estimate utility. In experiments, subjects were not required to pay their estimates. It was a hypothetical test.

Taking Subgroup i of Group I as an example, Meas. 1, 2, and 3 were specified as follows:
Meas. 1: A subject was given five stimuli of the massage on the waist of durations 8, 16, 24, 32, and 40 seconds, sequentially; the subject was not informed about durations of the stimuli. After one stimulus was completely presented, the subject was immediately asked to report orally a money mount he was willing to pay for the stimulus just experienced. And so on for next stimulus. When all five stimuli were given and five money amounts were reported, Meas. 1 concluded. With the same subject, the experiment proceeded to Meas. 2.

Meas. 2: The experimenter posed the following question to the subject: If you are asked to pay $¥ 1.0$, what is duration of the massage you will demand? The experimenter began the massage on the subject’s waist. The subject estimated the time for the money amount $¥ 1.0$ and said an "OK" immediately when he felt that an appropriate length of time had passed. The experimenter recorded the time with a stopwatch. And so on for $¥ 2.0, ¥ 3.0, ¥ 4.0$, and $¥ 5.0$ sequentially. When all five durations were reported, Meas. 2 concluded. With the same subject, the experiment proceeded to Meas. 3.

Meas. 3: The experimenter posed the following question to the subject: If you are given an electrical-power massage for 8 seconds, how much will you be willing to pay? The experimenter began the massage on the subject's waist for 8 seconds to acquaint him with such a time. After the subject reported on an answer sheet and the experimenter collected the answer sheet, the experimenter posed the second question to the subject: If you are given an electrical-power massage for 16 seconds, how much will you be willing to pay? And the subject reported on another answer sheet. If the subject asked, 16 seconds of the massage could be experienced by him (in fact, no subject did so in experiments). And so on until 24, 32, and 40 seconds were presented to the subject. Finally, the subject reported five money amounts respectively for $8,16,24,32$, and 40 seconds on five separate answer sheets. Meas. 1, 2, and 3 all concluded.

Massage durations and money amounts assigned in L-experiment are shown in Table 2.

Table 2. Stimulus levels of Meas. 1~3

| Meas. 1 | $8^{\prime \prime}$ | $16^{\prime \prime}$ | $24^{\prime \prime}$ | $32^{\prime \prime}$ | $40^{\prime \prime}$ |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Meas. 2 | $¥ 1.0$ | $¥ 2.0$ | $¥ 3.0$ | $¥ 4.0$ | $¥ 5.0$ |
| Meas. 3 | $8^{\prime \prime}$ | $16^{\prime \prime}$ | $24^{\prime \prime}$ | $32^{\prime \prime}$ | $40^{\prime \prime}$ |

There are two kinds of order effects taken into account in Latin square designs. The first is the order effect between Meas. 1~3, and the second between five stimuli assigned in each of Meas. 1~3.

Meas. 1~3 were carried on sequentially one by one in an independent room and spent about 10~15 minutes in total for a subject's complete program. During every measurement, only one subject was allowed into the room.

### 3.2 Ekman relations between single and double estimates

Meas. 2 and 3 are single estimate. They follow the logarithmic law $S=c \ln (q+a)+C$. Thus, for Meas. 2,

$$
\begin{equation*}
S_{2}=c_{2} \ln \left(m+a_{2}\right)+C_{2} ; \tag{6}
\end{equation*}
$$

and for Meas. 3,

$$
\begin{equation*}
S_{3}=c_{3} \ln \left(q+a_{3}\right)+C_{3} ; \tag{7}
\end{equation*}
$$

where, $S_{2}$ denotes the utility scaling value of money measured by consuming quantity (the massage time), $m$ the money amount assigned in Meas. $2 ; S_{3}$ denotes the utility scaling value of consuming quantity (the massage time) measured by money amount, $q$ the duration of the massage assigned in Meas. 3. In the coordinate with horizontal scale of massage duration and vertical scale of money amount, (6) will presents a convex curve for Meas. 2 and (7) a concave curve for Meas. 3. They will be tested by experiments.

Meas. 1 is a double estimate. Subjects have to estimate the utility of money in terms of quantity consumed, at the same time, estimate the utility of consuming quantity in terms of money, all through their intuitions. In this estimation a subject makes effort to find a subjective scaling value $S_{2}$ to match with another subjective scaling value $S_{3}$, that is,

$$
\begin{equation*}
S_{2}=b S_{3} . \tag{8}
\end{equation*}
$$

Where $b$, named "preference coefficient", is a proportion reflecting subjects' preference to money or consuming in Meas. 1; here $b$ increases as the money preference increases.

Substitute (6) and (7) in (8)

$$
\begin{align*}
& c_{2} \ln \left(m+a_{2}\right)+C_{2}=b c_{3} \ln \left(q+a_{3}\right)+b C_{3}, \\
& \begin{aligned}
& \ln \left(m+a_{2}\right)=\frac{b c_{3}}{c_{2}} \ln \left(q+a_{3}\right)+\frac{b C_{3}-C_{2}}{c_{2}}=\ln \left(q+a_{3}\right)^{\frac{b c_{3}}{c_{2}}}+\ln \exp \left(\frac{b C_{3}-C_{2}}{c_{2}}\right) \\
&=\ln \left[\left(q+a_{3}\right)^{\frac{b c_{3}}{c_{2}}} \exp \left(\frac{b C_{3}-C_{2}}{c_{2}}\right)\right]=\ln \left[\beta\left(q+a_{3}\right)^{\alpha}\right], \\
& m=\beta\left(q+a_{3}\right)^{\alpha}-a_{2} .
\end{aligned}
\end{align*}
$$

In (9), $\alpha$ and $\beta$ follow

$$
\begin{align*}
& \alpha=\frac{b c_{3}}{c_{2}}  \tag{10}\\
& \beta=\exp \left(\frac{b C_{3}-C_{2}}{c_{2}}\right) \tag{11}
\end{align*}
$$

Equation (9) shows that the relation between money amount ( $m$ ) and consuming quantity ( $q$ ) follows Stevens' power law in the double estimate of Meas. 1. The above inferences from (6) and (7) to (9) are similar to those of Ekman deriving Stevens' power law from Fechner's logarithmic law (Ekman, 1964). Below, (10) and (11) are together called Ekman relations. They will be tested in experiments as assistant evidences to examine the power law.

### 3.3 Results

Original data and their treatments are presented in Part 5 of Supplemental Files. Before the original data were used in analysis, a structural normalization had been made on them to equalize the structural contributions from every subject's data. This structural normalization completely reserves the overall quantitative levels and characters of all subjects' data (see Part 3 of Supplemental Files).

Fitting the power law (9) in means of structural-normalized data of Meas. 1 and the logarithmic laws (6) and (7) in means of structural-normalized data of Meas. 2 and 3, experimental values of $a_{2}, c_{2}$, $C_{2}, a_{3}, c_{3}, C_{3}, \alpha$, and $\beta$ are acquired as shown in Table 3.

Table 3. Observed parameters in Latin square experiment

| $a_{2}$ | $c_{2}$ | $C_{2}$ | $a_{3}$ | $c_{3}$ | $C_{3}$ | $\alpha$ | $\beta$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 10 | 74.6 | -169.8 | 20 | 10.4 | -32.2 | 0.596 | 1.88 |

Using the results of Table 3, the power law and logarithmic laws for Meas. 1~3 in L-experiment are derived as the following

$$
\begin{aligned}
& m=1.88(q+20)^{0.596}-10 ; \\
& S_{2}=74.6 \ln (m+10)-169.8 ; \\
& S_{3}=10.4 \ln (q+20)-32.2 .
\end{aligned}
$$

Fig. 2 plots the above curve estimations of Meas. 1~3. The data of Meas. 1 well approve the power law with $\mathrm{R}^{2}=0.99$, and the data of Meas. 2 and 3 well approve the logarithmic laws with $\mathrm{R}^{2}=1.0$ for Meas. 2 and $\mathrm{R}^{2}=0.99$ for Meas. 3, respectively. Just as the above theoretical expectations, the logarithmic law for Meas. 2 is convex, while, the logarithmic law for Meas. 3 concave. The similar
results are also observed in five simple experiments (see Part 4 in Supplemental Files). The L-experiment and five simple experiments present similar utility curve groups, revealing the robustness of experimental results.


Fig. 2. Curve regressions for Meas. $1 \sim 3$ in Latin square experiment. The data of Meas. 1 well approve the power law with $\mathrm{R}^{2}=0.99$, and the data of Meas. 2 and 3 well approve the logarithmic laws with $\mathrm{R}^{2}=1.0$ for Meas. 2 and $\mathrm{R}^{2}=0.99$ for Meas. 3, respectively. Though they are with identical stimulus intensities and presentation orders, the double estimate Meas. 1 and the single estimate Meas. 3 deliver two significantly different utility curves. And just as theoretical expectations, the logarithmic law for Meas. 2 is convex, while, the logarithmic law for Meas. 3 concave.

In L-experiment, massage durations and presentation orders of Meas. 1 and 3 are the same, and the distinction between Meas. 1 and 3 is only that the former is a double estimate but the latter a single estimate. To examine the distinction between single and double estimates, and taking into account that fitted data for each measurement are small sample ( $n=5$ ), Wilcoxon test is selected as an examination tool. In L-experiment, the Wilcoxon test for Meas. 1 and 3 presents $\mathrm{Z}=2.023, \mathrm{p}=0.043$. Their difference is significant. For five simple experiments, labeled Exps. I~V, Wilcoxon tests are also performed for Meas. 1 and 3 in every experiment. Table 4 collects all those testing results.

Table 4. Wilcoxon tests on Meas. 1 and 3 in L-experiment and Exps. I $\sim \mathrm{V}$ ( $\mathrm{n}=5$ in each)

| L-exp. | Exp. I | Exp. II | Exp. III | Exp. IV | Exp. V |
| :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathrm{Z}=2.023$ | $\mathrm{Z}=2.041$ | $\mathrm{Z}=1.095$ | $\mathrm{Z}=2.023$ | $\mathrm{Z}=2.023$ | $\mathrm{Z}=2.023$ |
| $\mathrm{p}=0.043$ | $\mathrm{p}=0.041$ | $\mathrm{p}=\mathbf{0 . 2 7 3}$ | $\mathrm{p}=0.043$ | $\mathrm{p}=0.043$ | $\mathrm{p}=0.043$ |

Table 4 shows the difference between Meas. 1 and 3 significant ( $\mathrm{p}<0.05$ ) in five of six experiments,
and only Exp. II ( $p=0.273$ ) is insignificant. That is, single and double estimates are generally different, they deliver two distinguished subjective measures.

Substituting the experimental values of $c_{2}, C_{2}, c_{3}$, and $C_{3}$ and taking $b=4.0$ in (10) and (11), obtain the theoretical values of $\alpha$ and $\beta$ in L-experiment

$$
\begin{aligned}
& \alpha=\frac{b c_{3}}{c_{2}}=\frac{4.0 \times 10.4}{74.6} \approx 0.558 \\
& \beta=\exp \left(\frac{b C_{3}-C_{2}}{c_{2}}\right)=\exp \left(\frac{-4.0 \times 32.2+169.8}{74.6}\right) \approx e^{0.550} \approx 1.73 .
\end{aligned}
$$

Similarly theoretical values of $\alpha$ and $\beta$ in five simple experiments are also obtained. Table 5 presents all those results. The theoretical values of $\alpha$ and $\beta$ agree with their experimental values with an average relative error [(Experimental-Theoretical) /Experimental] of $6.5 \%$ for L-experiment and five simple experiments. Ekman relations (10) and (11) hold at this error level. They reveal relationships between single- and double-estimate utility scales.

In addition, Table 5 shows a systematic deviation between theoretical and experimental values of $\alpha$ and $\beta$, where the theoretical values are identically less than the experimental ones. If the experimental values multiply a revision coefficient 0.92 , their agreements would be greatly improved as shown in Table 6.

Table 5. Experimental and theoretical values of $\alpha$ and $\beta$

|  | $b$ | $\alpha$ |  | $\beta$ |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | Theory | Experiment | Theory | Experiment |
| L-exp. | 4.0 | 0.558 | 0.596 | 1.73 | 1.88 |
| Exp. I | 2.6 | 0.641 | 0.704 | 0.72 | 0.77 |
| Exp. II | 2.5 | 0.730 | 0.786 | 0.50 | 0.51 |
| Exp. III | 3.1 | 0.903 | 0.950 | 0.25 | 0.26 |
| Exp. IV | 2.4 | 0.716 | 0.785 | 0.52 | 0.54 |
| Exp. V | 2.7 | 0.602 | 0.651 | 0.84 | 0.97 |

Table 6. Revised experimental values of $\alpha$ and $\beta$

|  | $\alpha$ |  |  | $\beta$ |  |
| :--- | :---: | :---: | :---: | :---: | :---: |
|  | Theoretical | Revised experimental |  | Theoretical | Revised experimental |
| L-exp. | 0.558 | $0.596 \times 0.92 \approx 0.548$ |  | 1.73 | $1.88 \times 0.92 \approx 1.73$ |
| Exp. I | 0.641 | $0.704 \times 0.92 \approx 0.648$ |  | 0.72 | $0.77 \times 0.92 \approx 0.71$ |
| Exp. II | 0.730 | $0.786 \times 0.92 \approx 0.723$ |  | 0.50 | $0.51 \times 0.92 \approx 0.47$ |
| Exp. III | 0.903 | $0.950 \times 0.92 \approx 0.874$ |  | 0.25 | $0.26 \times 0.92 \approx 0.24$ |
| Exp. IV | 0.716 | $0.785 \times 0.92 \approx 0.722$ |  | 0.52 | $0.54 \times 0.92 \approx 0.50$ |
| Exp. IV | 0.602 | $0.651 \times 0.92 \approx 0.599$ |  | 0.84 | $0.97 \times 0.92 \approx 0.89$ |

The deviation may be resulted from constant errors occurring in measurements. For example, in the estimates of massage durations in Meas. 2, whatever subjects' immediate oral reports by saying "OK" or experimenters' manual records by pressing a stop watch tend to enlarge the massage durations but never to shorten them. It might be evident one of causes for constant errors. Table 6 indicates the revision coefficient for those constant errors is about 0.92 . To prevent those errors, using real payoff

## 4. Conclusions and discussions

### 4.1 Syntheses for evidences

Utility scaling can be divided into two types of single and double estimates by their different scaling processes. Double estimate Meas. 1 and single estimate Meas. 3 in an experiment are completely the same as consumptions, but present significantly different utility scaling curves. The category of single and double estimates is necessary in the utility scaling.

The evidence from ultimatum game experiment justifies an empirical basis of Bernoulli's hypothesis for deriving a utility logarithmic law. The mathematical demonstration confirms that a logarithmic law, if it exists in economic single estimate, has a sense of Klein-Rubin utility function, in other words, it does agree with a usual utility concept in economics. Finally, experimental outcomes of single estimates Meas. 2 and 3 under distinguished circumstances approve the logarithmic law. To meet with evidences from the above three aspects, that is, Bernoulli's hypothesis and its supporting evidence from the ultimatum game, the mathematical demonstration for utility implication, and the experimental observation of utility scaling, in a logical framework, the best conclusion is just that single-estimate utility judgment follows the logarithmic law. In fact, a number of important utility functions broadly used in empirical studies which describe commodity choice behaviors, the typical single estimates, consist of the logarithmic components (e.g., Klein and Rubin, 1947; Stone, 1954; Liuch, 1973; Houthakker, 1960; Theil, 1965; Deaton and Muellbauer, 1980). The logarithmic law of single estimate had broad empirical evidences long ago.

Money-risk estimate is usually a double estimate. The subjects' estimate described in Kahneman and Tversky's experiment conducted for cumulative prospect theory is a typical example of money-risk estimate. They obtained the experimental curves for those money-risk estimates and found that it is appropriate if a power function is used to describe them (Kahneman and Tversky, 1992). It supports the power law of double estimate in the utility scaling. In addition, experimental tests to Ekman relations deliver a positive result supporting the power law in Meas. 1, derived from logarithmic laws in Meas. 2 and 3. And finally, utility-scaling experimental outcomes approve the power law in Meas. 1, a double estimate. Synthesizing the evidences from the three aspects, that is, experiments of cumulative prospect theory, tests of Ekman relations, and outcomes from Meas. 1, in a logical framework, the best conclusion is just that double-estimate utility judgment follows the power law. Furthermore, experimental Ekman relations mean that in utility scaling, the logarithmic law is a fundamental law and the power law is only a corollary derived from such a fundamental law.

The above conclusions mean that at least at the experimental levels of money amounts and consumption quantities, commodity choice follows logarithmic law, while, risk choice follows power law. They were derived from synthesized evidences rather than from single compelling one, that is, not from single analysis on ultimatum game, experiment for cumulative prospect theory, mathematical demonstration, or experimental curve regression, but from all of them.

### 4.2 Psychological attributes of utility judgment

Traditional concept of utility is vaguely addressed as an intuitive interpretation, for example, "the satisfaction derived from consuming commodities" (Henderson and Quant, 1971). However, such a utility concept has never been seriously applied in economic theoretical or empirical studies. To give
this concept an operational sense, a market investigation for consumers' satisfaction seems feasible. Nonetheless, existing recognized utility functions used in empirical studies, such as Klein-Rubin utility function, all are not obtained from satisfaction investigations. In fact, useful utility functions in empirical studies all are proposed for commodity quantity measures but not for satisfaction measures. The traditional utility concept is an oral decorate artifact rather than a meaningful definition, without any direct contribution to existing economic descriptions such as utility maximization, empirical analysis on consumers' commodity choices, and so on. It may mix economics with marketing. This concept should be clarified from more detailed analysis on its psychological attributes and economics requirement.

As commodity consumptions, subjects’ judgments in Meas. 1~3 have no difference, and the satisfactions derived from them should be the same. However, they deliver significantly different utility descriptions. The results of Meas. 1 and Meas. 3 are two different utility curves subject to different functions; Meas. 2 follows a convex utility curve, but Meas. 3 a concave one. They are distinguished from each other only by their different estimate patterns. Thus, the judgment of utility relies on the psychological presentation patterns rather than a purely satisfaction derived from consuming commodities. Utility is a typical psychological magnitude that is differently yielded from various perceptual processes. Perceptual pattern is one of important determinants in it.

The utility curves of Meas. 2 and 3 were changed from convex to concave in experiments. It is certainly not caused by the variation of subjects' preference but by the variation of stimulus presentation pattern. In Meas. 2, the money amounts were informed to subjects who were asked to estimate utilities of those money amounts in terms of massage durations. In reverse, in Meas. 3, the massage durations were informed to subjects who were asked to estimate utilities of those massage durations in terms of money amounts. Such a change of perceptual process caused the variations from convex to concave. In Meas. 1 and 3, two different perceptual processes, double and single estimates, also caused different two kinds of utility curves reported. All these changes are attributed to the variations of perceptual patterns.

Either in existing theoretical or empirical applications of utility analyses, for instance, in the theoretical analysis on utility maximization or econometric applications in empirical data, the utility function is always used to describe how consumers to distribute their quantity choices among varieties of commodities but not to indicate the degrees of consumers' satisfactions. The psychophysical interpretation for utility measure seeks to reveal the relationship between subjective and objective commodity quantities. It agrees with economics requirements of utility analysis. Utility theory should emphasize more on the analysis of perceptual quantity.

In the light of discussions in Part 2.3.1, Klein-Rubin utility function implies the relationship between subjective and objective commodity quantities. In other words, the essential of utility maximization, for example in LES and ELES, is the maximization of subjective commodity quantities. Therefore, "utility" should be operationally defined as "subjective quantity of commodity or evaluation"-a measurable object and an applicable theoretical concept. It will enable us to positively test the utility maximization hypothesis in a framework of theoretical cardinal utility maximization and empirical econometric model.

### 4.3 An experimental procedure for testing utility maximization

The cardinal utility maximization has been proposed for more than one hundred and fifty years (Gossen, 1854). Such an important hypothesis had been regarded for a long time un-testable but
replaceable by the ordinal utility maximization (e.g. Hicks, 1939), which seems measurable at a first glance. However, any experimental analysis on ordinal utility maximization must rely on the observable indifference curve. As seen in early experimental studies, no ideal indifference curves with precise shape had ever been obtained. Those experimental studies fell in fact into predicaments in which no ideal theoretical properties were found in objective observations unless the properties were introduced as a prior knowledge in the data arrangements (Thurstone, 1931; Roussears and Hart, 1951) or as a prior reminder to instill into subjects by a training session before they participated in the experiment (MacCrimmon and Toda, 1969). Till date, it has been found impossible to test the utility maximization experimentally within the ordinal utility framework. Ordinal utility may be another idle dream for classical economics. In this paper, cardinal utility had been revealed testable. Combining with the experimental outcome in the single estimate, the interpretation of Klein-Rubin utility function for the logarithmic law implies a possibility of experimentally examining the cardinal utility maximization hypothesis.

Klein-Rubin utility function $U=\sum b_{i} \ln \left(q_{i}-r_{i}\right), \quad b_{i}=\frac{p_{i}\left(q_{i}-r_{i}\right)}{\sum p_{i}\left(q_{i}-r_{i}\right)}$, is measurable provided $b_{i}$, $q_{i}$, and $r_{i}$ can be determined in an experiment. Traditional application of LES uses empirical data of $q_{i}$ to determine $b_{i}$ and $r_{i}$. It can be transplanted to experimental approach. The change in this transplantation is only that empirical $q_{i}$ is replaced by experimental $q_{i}$. Nevertheless, to avoid the logic circulation, LES is impossible to independently test the Klein-Rubin utility maximization hypothesis because LES is just derived from such a utility maximization. However, now there is another way to determine Klein-Rubin utility function experimentally by utility scaling approach: assigning a fixed budget constraint and prices $p_{1}, p_{2}, \ldots, p_{n}$ in an $n$-commodity bundle for all subjects in an experiment, we can determine quantities $q_{1}, q_{2}, \ldots, q_{n}$ by observing subjects how to distribute their choices in such an $n$-commodity bundle, derive constants $r_{1}, r_{2}, \ldots, r_{n}$ by measuring the logarithmic laws of the utilities of separate commodities (similar to Meas. 3), and combining data of $q_{i}$ and $r_{i}$ further estimate $b_{1}, b_{2}, \ldots$ $b_{n}$. In this way, we are able to construct the experimental Klein-Rubin utility function by the linear combination of the logarithmic laws experimentally in an $n$-commodity bundle. Meanwhile, as just mentioned, solving LES for $b_{i}$ and $r_{i}$ in data of the experimental $q_{i}$, we also obtain a Klein-Rubin utility function implied in LES for this $n$-commodity bundle. Denote $U_{E s t}$ as the Klein-Rubin utility function experimentally estimated from the linear combination of logarithmic laws, and $U_{L E S}$ the Klein-Rubin utility function implied in LES derived from experimental data of $q_{i}$. The utility maximization will be experimentally tested by comparing $U_{\text {Est }}$ with $U_{L E S}$ in a commodity bundle.

LES is the result of maximizing Klein-Rubin utility function, thus, $U_{L E S}$ can be viewed as the utility function theoretically required by utility maximization for this commodity bundle. $U_{\text {Est }}$ is the intuitive utility estimate for such a commodity bundle, thus, directly indicates an outcome of behavioral observations. If the comparison reveals an agreement between $U_{L E S}$ and $U_{E s t}$, the theoretic utility maximization implied in LES agrees with the behavioral utility evaluation in subjects' intuitive judgment and, therefore, the experiment supports the utility maximization hypothesis; otherwise, not.

It will be an important and inspiring progress that removes a long unsettled perplexity in a fundamental stone of economics.

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#### Abstract

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