# Price as a matter of choice and nonstochastic randomness 

Yaroslav Ivanenko<br>Banque de France ${ }^{1}$


#### Abstract

The problem of valuation of a European option is considered as a problem of choice, with price being a decision. A version of indifference valuation relation is proposed that includes statistical regularities of nonstochastic randomness. Expressions for bid and ask prices are obtained as well. Classical relations (forward contract value and Black-Scholes formula) are obtained as particular cases. We show that in the general case of nonstochastic randomness the minimal expected profit of uncovered European option position is always negative. A version of delta hedge is proposed.


Key words Derivatives Valuation, Decision System, Nonstochastic Randomness, Statistical Regularity

## 1. Introduction

Traditionally, beginning from von Neumann and Savage, decision theory has provided the language of discourse to many topics in mathematical economics (see, for instance, MasCollel et al (1995)) as well as in mathematical finance (see, for instance, Markowitz (1952)). With the time being, the synthesis of decision theory and economics, with its focus on uncertainty, produced many alternative decision theories (see, Machina (2004)). These decision theories, relying on economic applications and in order to correctly represent preferences on actions of economic agents, required existence of set functions (measures) that were difficult to interpret from traditional statistical standpoint (see, for instance, Shmeidler (1989)). On the other hand, the synthesis of decision theory and mathematical finance, with its focus on arbitrage and probabilistic modeling, proved to be of somewhat another nature (Harrison and Kreps (1979), Back and Pliska (1991), Carmona (2008). Namely, relying on the well developed theory of stochastic processes, mathematical finance has almost become an application of this sophisticated mathematical discipline, hard to explain to a non practitioner. A bridge between these two branches of application of decision theory does not seem to have been made yet.

[^0]The present article explores one of the ways in which the decision theory presented in Ivanenko (2010) can be applied to some problems of mathematical finance. In particular, this version of decision theory allows one to interpret the price of a derivative contract as a matter of choice, or a decision. In its own turn, this interpretation allows one to consider the absence of arbitrage opportunity as a particular case of the more general phenomenon of uncertainty, precisely defined and provided with existence theorem in the abovementioned book. In this optics, the absence of arbitrage opportunities becomes more a matter, or a criterion of choice than anything else.

To consider price (of a financial contract, of an insurance contract, etc.) as an object of choice, or a decision, seems a very natural attitude. We learn about this attitude once, for example, we start working on a trading floor. A limit price is a decision of the market participant - a buyer or a seller. A choice of the price can have a huge impact on profit and loss in case of important swap transactions. Interest rates have been traditionally regarded as decisions.

In financial markets, exposure to a financial contract is only possible through a mutual agreement of two market participants, a buyer and a seller - two decision makers. Once they agree upon the price, the transaction is done, the price thereupon is known (it may be published on a computer screen, newspaper page, etc). The fact that the transaction has happened may be interpreted in the following way: we say that the counterparties have agreed to make the same decision, to choose the same price. This phenomenon will be described in the present article as a problem of choice.

One of the consequences of this description will be the reference to the so called statistical regularities of nonstochastic randomness, a new mathematical formalism that was developed in order to describe statistically unstable random phenomena, and systematically presented for the first time in the abovementioned book as well. That it is important to distinguish between stochastic and nonstochastic randomness was stressed already by Kolmogorov (1986). Perhaps, this formalism, since it operates with families of finitelyadditive probability measures, could be profitable both to decision theory and mathematical finance researchers.

Indeed, the abovementioned extension of the definition of arbitrage may lead one to the necessity to accept the families of finitely-additive probability measures, a step which is necessary to take in order to deal with the well known paradoxes of the fundamental theorems of asset pricing. For example, in Back and Pliska (1992) it was shown that countable -
additive measures are insufficient in order to represent no arbitrage situations when the state space is infinite. On the other hand, the part of the mathematical finance literature which is dedicated to dynamic hedging (see Black and Scholes (1973)) and the whole corpus of related research, which becomes possible only if a model of random evolution of a financial variable has been specified. Traditionally these models have been represented by stochastic processes. However, it becomes clear that the constraint of stochasticity is too restrictive in order to represent the reality of the markets (see, for example, Mandelbrot (2006)). Some authors, recognizing this limitation, relax that constraint and come very close to nonstochastic modeling of random evolution of financial variables, as in Epstein et al (2000). Another branch of research on this topic, the indifference pricing framework, as summarized in Carmona (2008), takes the road of expected utility maximization and pursuits the search of optimal decisions under diverse constraints. Unlike standard indifference pricing framework, the approach proposed in this article does not adopt the ideas of optimization. Moreover, unlike usual utility maximization technique, the approach proposed here does not require the use of convex utility functions: the convexity responsible for risk or uncertainty aversion is in the shape of the decision criterion. One can see parallels of this approach and the arbitrage pricing theory proposed in Ross (1976, 1977). It seems as well that the approach proposed here is a natural framework for static replication arguments: see Derman and Taleb (2008).

The new terminology, notions and theorems encountered in this article are due to Ivanenko (2010), which is the main reference of this article, and could be consulted there. In particular, we refer to Theorems 4.2 and 4.8 , concerning existence of the statistical regularity of nonstochastic randomness, to Theorem 5.2 concerning existence of statistical regularity in decision systems, to Definition 2.4 and Theorems 2.2 and 2.3 concerning existence of uncertainty in decision systems.

In what follows we present the theoretical framework (Section 2.1-2.2) that essentially is an adaptation of valuation problematic to decision theory and vice versa, and then show (Section 2.3) how some classical results of mathematical finance (forward contract value and Black-Scholes formula) are obtained as particular cases. In Section 2.4 an expression for the generalized delta is proposed for statistical regularities of general form.

## 2. Price as decision and nonstochastic randomness

### 2.1. Indifference as absence of arbitrage opportunities: one underlying

Let $f(\theta)$ be a pay-off of a financial contract, depending on the value of uncontrolled (unknown) parameter $\theta$. For example, $f(\theta)=\left(\theta-\theta^{*}\right)^{+}$, where $\theta$ is the price of the underlying financial variable at the maturity date $T, \theta^{*}$ is the strike. Let at the moment $t=0$ we observe the transaction of the pay-off $f(\theta)$ at a certain price $u$. Let the buyer finances the purchase with a loan, so that by the time $T$ she would owe the bank the sum $u e^{r T}$. Let the seller places the proceeds of the sale in the bank, so that by the time $T$ the bank will owe him the sum $u e^{r T}$. We shall write that by the moment $T$ the profit and loss of the buyer and the seller will be, respectively,

$$
\begin{equation*}
L_{b}(\theta, u)=-u e^{r T}+f(\theta) \tag{1}
\end{equation*}
$$

and

$$
\begin{equation*}
L_{s}(\theta, u)=+u e^{r T}-f(\theta) . \tag{2}
\end{equation*}
$$

Our goal is to interpret the observed price $u$ of the financial contract as a decision of the buyer and of the seller. Let $U=\{u: u \in \mathbb{R}\}$ be the set of decisions, or actions, that in this case are prices. Let $\Theta$ be the set of values of the random parameter $\theta$. Note that the values of the above profit and loss functions $L$ represent outcomes with the natural preference relation on them: the bigger the profit, the better. We can thus construct two sets, $Z_{b}=\left(\Theta, U, L_{b}(.,).\right)$ and $Z_{s}=\left(\Theta, U, L_{s}(\ldots)\right)$, called the decision scheme of the buyer and, respectively, of the seller.

Following the framework of decision systems we admit that the buyer and the seller belong to the class $\Pi_{1}$ of decision makers, that is those whose preference relation on actions is represented by means of the criterion $L_{Z}^{*}: U \rightarrow \mathbb{R}$ that has the form

$$
\begin{equation*}
L_{Z}^{*}(u)=\max _{p \in \boldsymbol{P}} \int L(\theta, u) p(d \theta), \tag{3}
\end{equation*}
$$

where $Z=(\Theta, U, L(.,)$.$) is the decision scheme, U$ - is the set of decisions $u, \Theta$ - is the set of values of the uncontrolled (random) parameter $\theta, L: \Theta \times U \rightarrow \mathbb{R}$ - is a loss function, and where $\boldsymbol{P}$ is a so called statistical regularity in the form of a closed set of finitely-additive probability measures on $\Theta$. Note that the sets $\Theta$ and $U$ are arbitrary and the function $L$ is bounded. These circumstances will allow us to extend the set of decisions in a way that suits
most the goal of the article. Any statistical regularity $\boldsymbol{P}$ describes a statistically unstable random phenomenon, called nonstochastically random ${ }^{2}$. If $L$ has the meaning of a utility function, or as in (2.1)-(2.2) of a profit and loss function, the criterion (2.3) of maximal expected losses becomes the criterion of minimal expected utility ${ }^{3}$

$$
L_{Z}^{*}(u)=\min _{p \in P} \int L(\theta, u) p(d \theta)
$$

In what follows we shall use the form ( $3^{\prime}$ ) of the criterion and call it the criterion of minimal expected profits.

We admit further that the buyer and the seller not only belong to the same class $\Pi_{l}$ of decision makers, but share as well the same view on the statistical regularity $\boldsymbol{P}$ of the behavior of the uncontrolled random parameter $\theta \epsilon \Theta$. In our case this means that the price $u$ of the transaction is that decision $u$ that makes the buyer and the seller indifferent between the profit and loss profiles $L_{b}(\theta, u)$ of the buyer and $L_{s}(\theta, u)$ of the seller. In other words, we can interpret the fact of the transaction at the price $u$ as equality ${ }^{4}$

$$
\begin{equation*}
L_{Z_{b}}^{*}(u)=L_{Z_{s}}^{*}(u), \tag{4}
\end{equation*}
$$

or, substituting (3')

$$
\begin{equation*}
\min _{p \in P} \int L_{b}(\theta, u) p(d \theta)=\min _{p \in P} \int L_{s}(\theta, u) p(d \theta) \tag{5}
\end{equation*}
$$

Taking into account (1)-(2), and due to boundedness of $\int f(\theta) p(d \theta)$, we obtain

$$
\begin{equation*}
-u e^{r T}+\min _{p \in P} \int f(\theta) p(d \theta)=u e^{r T}-\max _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta) \tag{6}
\end{equation*}
$$

Thus the observed price $u$ may be interpreted as the average of the following type

[^1]\[

$$
\begin{equation*}
u=e^{-r T} \frac{\min _{p \in P} \int f(\theta) p(d \theta)+\max _{p \in P} \int f(\theta) p(d \theta)}{2} \tag{7}
\end{equation*}
$$

\]

where $\boldsymbol{P}$ is a statistical regularity on $\Theta$. The formula (7) means that market negotiates, or agrees upon, a particular choice of the statistical regularity $\boldsymbol{P}$. A stochastic analogue of this is known as implied distribution.

Remark though that if one requires that $L_{Z_{b}}^{*}(u)=0$ and $L_{Z_{s}}^{*}(u)=0$ then one obtains, respectively, that

$$
u_{b}=e^{-r T} \min _{p \in P} \int f(\theta) p(d \theta)
$$

that can be interpreted as the bid price and

$$
\begin{equation*}
u_{s}=e^{-r T} \max _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta) \tag{7"}
\end{equation*}
$$

that can be interpreted as the ask price. The indifference price (7) can be interpreted thus as the mid price.

In the case of complete uncertainty, when $\boldsymbol{P}$ is a the set of all finitely-additive probability measures on $\Theta$, including all Dirac delta distributions, the formula (7) becomes

$$
\begin{equation*}
u=e^{-r T} \frac{\min _{\theta \in \boldsymbol{\Theta}} f(\theta)+\max _{\theta \in \boldsymbol{\Theta}} f(\theta)}{2} \tag{8}
\end{equation*}
$$

In the case of stochastic character of the statistical regularity, that is when the set $\boldsymbol{P}$ contains a single stochastic, that is countably-additive probability measure $p$, the formula (7) becomes

$$
\begin{equation*}
u=e^{-r T} \quad \int f(\theta) p(d \theta) \tag{9}
\end{equation*}
$$

Because of the equality (4) the price (7) may be considered as another version of the indifference price: see Carmona (2008). In our case, however, a decision maker is indifferent between the role of the buyer and the role of the seller. Here indifference means that she has no grounds to prefer one operation to another. We say that this inability to prefer, or indifference, is synonymic to the absence of arbitrage opportunities. By the same token, a
situation where there is no indifference, or where preference can be established, is synonymic of arbitrage opportunity. ${ }^{5}$

In statistical literature, for example in De Groot (1970), the criterion of the type (3) is called risk. Therefore our indifference price may be interpreted as that decision that makes equal the risks of the counterparties. This interpretation leads to a parallel with the arbitrage pricing theory, developed in Ross $(1976,1978)$.

### 2.2. Arbitrary portfolio

Since we admit that decision makers belong to one class of decision makers, it is reasonable to rewrite the above for the case of a single decision maker as well as for the case of multiple securities. Let us first consider only one underlying. Let the random parameter $\theta$ be the price of the underlying security by the future time moment $T$, and let $\Theta$ be the set of its possible values. Let $F=\{\Theta \rightarrow \mathbb{R}\}$ be a set of all real bounded functions on $\Theta$ that we interpret as pay-offs of financial instruments depending on the random parameter $\theta \in \Theta$. Let $q \in Q=\{q: q \in \mathbb{R},|q|<\infty\}$ represent the sense and the amount of the operation ( $q>0$ for purchase, $q<0$ for sale, $q=0$ for no operation. Let $U=\{u: u \in \mathbb{R},|u|<\infty\}$ be the set of prices. Then let $D=F \times Q \times U$ be the set of decisions. This means that a decision maker chooses three elements: a contract, her role and the price. Supposing that the decision maker finances her purchases via a bank account and places the proceeds of sales there as well, the profit and loss function $L: \Theta \times D \rightarrow \mathbb{R}$ will be

$$
\begin{equation*}
L(\theta, d)=q\left(-u e^{r T}+f(\theta)\right), d=(q, u, f(.)) \in D, f \in F . \tag{10}
\end{equation*}
$$

In this way one constructs the matrix decision scheme $Z=(\Theta, D, L(\ldots)$.$) , an essential element$ of decision system. Since the representation (3) is valid for arbitrary sets $D$ and $\Theta$, and for bounded function $L$ one can extend the above setting (which is a model of a one period investment in a singular market) to the case of an arbitrary portfolio. Indeed, let there exist a collection of $j=1, \ldots, M$ underlying securities. Namely let $\theta=\left\{\theta_{1}, \ldots, \theta_{M}\right\}$ be their prices by the time moment $T$, and let $\Theta=\left\{\Theta_{1}, \ldots, \Theta_{M}\right\}$ be the collection of sets of their possible values. Without loss of generality we can consider $\Theta$ as the set of the states of Nature. Let $D^{(\infty)}=$

[^2]$\cup_{n=1}^{\infty} D^{n}, D^{n}=\underbrace{D \times D \ldots D}_{n}$, where $D$ is the elementary decision set from (10). Then the profit and loss function $L: \Theta \times D^{(\infty)} \rightarrow \mathbb{R}$ has the form
$$
L(\theta, d)=\sum_{i=1}^{N_{d}} q_{i}\left(-u_{i} e^{r T}+f_{i}(\theta)\right)
$$
where $d=\left(d_{1}, \ldots, d_{N_{d}}\right) \in D^{(\infty)}, d_{i}=\left(f_{i}(),. u_{i}, q_{i}\right) \in D, i=1, . ., N_{d}, \theta \in \Theta$. In this case the matrix decision scheme takes the form
\[

$$
\begin{equation*}
Z=\left(\Theta, D^{(\infty)}, L(\ldots)\right) \tag{11}
\end{equation*}
$$

\]

Any decision scheme

$$
Z^{\prime}=\left(\Theta^{\prime} \subseteq \Theta, D^{\prime} \subseteq D^{(\infty)}, L(., .)\right)
$$

is called a market. When for each $d_{i} \in d, i=1, . ., N_{d}, u_{i}$ are fixed, we say that we deal with an investment decision problem; when $f_{i} \in d$ are fixed we say that we deal with a valuation decision problem; when neither is fixed we may say that we deal with a market decision problem. The set of values of the profit and loss function (10') comprises the set of outcomes with the natural preference relation on it: the bigger the profit, the better. Due to the nature of financial markets, all the elements introduced above are considered as bounded: $M<$ $\infty, \forall d N_{d}<\infty, \forall j=1, \ldots, M \inf \Theta_{j}<\infty, \sup \Theta_{j}<\infty$.

In standard economic framework, the relation between the dimension of the set $\Theta$ of values of the uncontrolled parameter $\theta$ and of the set $D^{(\infty)}$ of decisions is known to take only two forms: complete or incomplete markets. We draw attention of the reader that the accent on this distinction seems to be somewhat artificial and is related, probably, to the fact, that finitely-additive probability measures were ruled out of the picture. Indeed, let the decision maker belong to the class $\Pi_{1}$. Then, according to Theorem 1 from Ivanenko (1986) or Theorem 5.2 from Ivanenko (2010), whatever the set of decision $D$ and whatever the set $\Theta$ of values of the unknown parameter, her preference relation on actions is represented by means of the criterion ( $3^{\prime}$ ) that now is written as

$$
\begin{equation*}
L_{Z}^{*}(d)=\min _{p \in P} \int L(\theta, d) p(d \theta) \tag{12}
\end{equation*}
$$

where $\boldsymbol{P}$ is a statistical regularity on $\Theta$ in the form of a closed set of finitely-additive probability measures on $\Theta$. The meaning of this theorem is the following (see details in Apendix). For any decision scheme $Z=(\Theta, D, L(.,)$.$) , and any decision maker from the$ class $\Pi_{l}$ one can find a statistical regularity $P \in \mathbf{P}(\boldsymbol{\Theta})$ such that $L_{Z}^{*}(d)=\min _{p \in \boldsymbol{P}} \int L(\theta, d) p(d \theta)$, where $\mathbf{P}(\boldsymbol{\Theta})$ is the set of all statistical regularities on $\boldsymbol{\Theta}$, and vice versa, for any $P \in \mathbf{P}(\boldsymbol{\Theta})$ and any $Z$, if $L_{Z}^{*}(d)=\min _{p \in P} \int L(\theta, d) p(d \theta)$ then decision makers belongs to the class $\Pi_{1}$. If for two decisions $d_{1}, d_{2} \in D^{(\infty)}$ the decision maker has

$$
\begin{equation*}
L_{Z}^{*}\left(d_{1}\right)=L_{Z}^{*}\left(d_{2}\right), \tag{13}
\end{equation*}
$$

then she has no grounds to prefer one decision to another, she, in other words, is indifferent with respect to the choice between them. Again, inability to prefer one decision to another, or indifference, is synonymic to the absence of arbitrage opportunities.

That one can consider the notions of indifference and the absence of arbitrage opportunities as synonyms seems to have deep roots. Indeed, when we deal with point payoffs, that is when $L\left(\theta, d_{1}\right)=\varphi\left(d_{1}\right), L\left(\theta, d_{2}\right)=\varphi\left(d_{2}\right), \forall \theta \in \Theta$, the requirement that $\varphi\left(d_{1}\right)=$ $\varphi\left(d_{2}\right)$, is known to be called a no arbitrage condition, see for example Ross (1976). At the same time, the situation when $L\left(\theta, d_{1}\right)=L\left(\theta, d_{2}\right), \forall \theta \in \Theta$, is known in mathematical finance as replication. When, on the other hand, $L\left(\theta, d_{1}\right) \geq L\left(\theta, d_{2}\right), \forall \theta \in \Theta$, and $\exists \theta \in$ $\Theta L\left(\theta, d_{1}\right)>L\left(\theta, d_{2}\right)$, that is when there exists a dominating decision, then the classical arbitrage portfolio can be formed. Thus these two phenomena, of no arbitrage and indifference, are closely related to the more general phenomenon called uncertainty, which takes place when the preference relation on outcomes can be projected onto more than one preference relations on decisions (see Ivanenko (2010) Chapter 2. Moreover, due to Theorems 2.2-2.3 and Theorem 5.2 from that book, condition (12) of indifference can be considered as a sufficient condition of existence of uncertainty in the decision system $S=(Z, P)$. Therefore we say that indifference price (7) is the choice that brings uncertainty into, or rules out arbitrage opportunities from, decision system $S=(Z, P)$. Note that this extension of the definition of arbitrage leads one to the necessity to accept the families of finitely-additive probability measures, a step which is necessary to take in order to deal with the well known paradoxes of the fundamental theorems of asset pricing (see, for instance Back and Pliska (1991)). As far as the author of this article knows, the above parallels have not been considered systematically.

Remark though, that when $d_{1} \in D^{1}, q=1$, that is for the buyer, one can, requiring $L_{Z}^{*}\left(d_{1}\right)=0$, conclude that

$$
\begin{equation*}
u=e^{-r T} \min _{p \in P} \int f(\theta) p(d \theta) \tag{14'}
\end{equation*}
$$

and interpret it as the bid price. Respectively when $q=-1$, that is for the seller,

$$
u=e^{-r T} \max _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta)
$$

and interpret it as the ask price.

In what follows we consider valuation decision problems for $M=1$. Now the question is what statistical regularity $\boldsymbol{P}$ will the decision makers use? We show in the next section how the argument of indifference leads to the analogue of a risk neutral measure, but in the case of nonstochastic randomness. In what follows relations (3')-(5) and (10)-(13) are our basic tool.

### 2.3. European options, forward contract and put-call parity: conditions on $P$.

The pay-off of a European call option is $f(\theta)=\left(\theta-\theta^{*}\right)^{+}$, where $\theta$ - is the underlying stock price at the maturity date $T, \theta^{*}$ - is the strike,. Hence, using (2.7) we have

$$
\begin{equation*}
u_{c}=e^{-r T} \frac{\min _{p \in P} \int\left(\theta-\theta^{*}\right)^{+} p(d \theta)+\max _{p \in P} \int\left(\theta-\theta^{*}\right)^{+} p(d \theta)}{2} \tag{15}
\end{equation*}
$$

The pay-off of the corresponding European put is $f(\theta)=\left(\theta^{*}-\theta\right)^{+}$. Hence

$$
\begin{equation*}
u_{p}=e^{-r T} \frac{\min _{p \in P} \int\left(\theta^{*}-\theta\right)^{+} p(d \theta)+\max _{p \in P} \int\left(\theta^{*}-\theta\right)^{+} p(d \theta)}{2} . \tag{16}
\end{equation*}
$$

For the corresponding forward contract on the underlying stock $f(\theta)=\theta-\theta^{*}$. Hence

$$
\begin{align*}
u_{f} & =e^{-r T} \frac{\min _{p \in \boldsymbol{P}} \int\left(\theta-\theta^{*}\right) p(d \theta)+\max _{p \in \boldsymbol{P}} \int\left(\theta-\theta^{*}\right) p(d \theta)}{2}= \\
& =e^{-r T}\left[\frac{\min _{p \in \boldsymbol{P}} \int \theta p(d \theta)+\max _{p \in \boldsymbol{P}} \int \theta p(d \theta)}{2}-\theta^{*}\right] . \tag{17}
\end{align*}
$$

What meaning could we attribute to the first term in the brackets? Let us try to define as decision the forward price $\theta_{F}$ for the stock supposing that its current spot price is $\theta_{0}$. On one
hand, $\theta_{F}$ must be such as to guarantee indifference between the long and the short forward positions, i.e. when, respectively,

$$
\begin{aligned}
& L_{l}\left(\theta, \theta_{F}\right)=-\theta_{0} e^{r T}+\theta_{F} \\
& L_{S}\left(\theta, \theta_{F}\right)=+\theta_{0} e^{r T}-\theta_{F}
\end{aligned}
$$

then one must have

$$
\begin{equation*}
\min _{p \in P} \int L_{l}\left(\theta, \theta_{F}\right) p(d \theta)=\min _{p \in P} \int L_{s}\left(\theta, \theta_{F}\right) p(d \theta) \tag{18}
\end{equation*}
$$

Since $L_{l}\left(\theta, \theta_{F}\right)$ and $L_{s}\left(\theta, \theta_{F}\right)$ are point pay-offs (they do not depend on $\left.\theta\right)$, then

$$
\begin{equation*}
\theta_{F}=\theta_{0} e^{r T} \tag{19}
\end{equation*}
$$

which is the classical forward price for a stock whose current price is $\theta_{0}$ (see Hull (2005)). On the other hand, in the case of non-deliverable forward, the indifference between long and short positions, the pay-offs of which in this case are, respectively,

$$
\begin{aligned}
& L_{l}\left(\theta, \theta_{F}\right)=-\theta_{F}+\theta \\
& L_{s}\left(\theta, \theta_{F}\right)=+\theta_{F}-\theta
\end{aligned}
$$

implies

$$
\begin{equation*}
\min _{p \in P} \int L_{l}\left(\theta, \theta_{F}\right) p(d \theta)=\min _{p \in P} \int L_{s}\left(\theta, \theta_{F}\right) p(d \theta) \tag{20}
\end{equation*}
$$

where again $\theta$ is the stock price at the date $T$ and $\theta_{F}$ is the decision with respect to the forward price. Whence

$$
\begin{equation*}
\theta_{F}=\frac{\min _{p \in \boldsymbol{P}} \int \theta p(d \theta)+\max _{p \in \boldsymbol{P}} \int \theta p(d \theta)}{2} . \tag{21}
\end{equation*}
$$

That is the average of the type (21), where $\theta$ - is the stock price at the future time moment $T$, $\Theta$ - is the set of its possible values, and $\boldsymbol{P}$ - is a statistical regularity on $\Theta$, has the meaning of the forward price as well. At the same time, from (19), we must have

$$
\begin{equation*}
\theta_{0}=e^{-r T} \frac{\min _{p \in \boldsymbol{P}} \int \theta p(d \theta)+\max _{p \in \boldsymbol{P}} \int \theta p(d \theta)}{2} \tag{22}
\end{equation*}
$$

Provided the price $\theta_{0}$ is known, equation (22) becomes a condition on the "fair" statistical regularity $\boldsymbol{P}$ on $\Theta$.

Note that we would obtain (22) considering as well the following decision problem. Let today's stock price $\theta_{0}$ be a decision in the situation where a decision maker chooses between long and short stock positions held to a certain time horizon and financed with a bank account. Namely, let the profit and loss function of the long and short positions be respectively

$$
L_{l}\left(\theta, \theta_{0}\right)=\theta-\theta_{0} e^{r T}
$$

and

$$
L_{s}\left(\theta, \theta_{0}\right)=\theta_{0} e^{r T}-\theta
$$

Being indifferent in this case means

$$
\begin{equation*}
\min _{p \in \boldsymbol{P}} \int\left(\theta-\theta_{0} e^{r T}\right) p(d \theta)=\min _{p \in \boldsymbol{P}} \int\left(\theta_{0} e^{r T}-\theta\right) p(d \theta) \tag{23}
\end{equation*}
$$

or

$$
\begin{equation*}
\theta_{0}=e^{-r T} \frac{\min _{p \in P} \int \theta p(d \theta)+\max _{p \in P} \int \theta p(d \theta)}{2} \tag{24}
\end{equation*}
$$

One can thus conclude that the choice of the spot price $\theta_{0}$ becomes a condition on $\boldsymbol{P}$ similar to (22).

Substituting (22), or (24), in (17) one obtains

$$
\begin{equation*}
u_{f}=e^{-r T}\left(\theta_{0} e^{r T}-\theta^{*}\right)=\theta_{0}-\theta^{*} e^{-r T} \tag{25}
\end{equation*}
$$

where $\theta_{0}$ - is the current stock price and $\theta^{*}$ - the strike. This quantity is usually called the value of the forward contract: see, for instance, Hull (2005). In other words, the problem of choice of the forward contact value is equivalent to the problem of choice of the spot price.

Now it is easy to retrieve the classical put-call parity relation. Consider the following decision scheme. On one hand, a decision maker can buy the call at the price $u_{c}$ (15), and simultaneously sell the put for the price $u_{p}(16)$. On the other hand, the decision maker can
take a forward position, the value of which is (25). The profit and loss from holding long callshort put position is

$$
\begin{align*}
L_{c p}\left(\theta,-u_{c}+u_{p}\right) & =\left(-u_{c}+u_{p}\right) e^{r T}+\left(\theta-\theta^{*}\right)^{+}-\left(\theta^{*}-\theta\right)^{+}= \\
& =\left(-u_{c}+u_{p}\right) e^{r T}+\left(\theta-\theta^{*}\right) \tag{26}
\end{align*}
$$

And the profit and loss from holding a long forward position is

$$
\begin{equation*}
L_{f}\left(\theta, u_{f}\right)=-u_{f} e^{r T}+\left(\theta-\theta^{*}\right) \tag{27}
\end{equation*}
$$

According to (12) and (13), being indifferent between these payoffs means

$$
\begin{equation*}
\min _{p \in P} \int L_{c p}\left(\theta,-u_{c}+u_{p}\right) p(d \theta)=\min _{p \in P} \int L_{f}\left(\theta, u_{f}\right) p(d \theta) \tag{28}
\end{equation*}
$$

or, taking into account (26) and (27),

$$
\begin{equation*}
u_{c}-u_{p}=u_{f} \tag{29}
\end{equation*}
$$

Substituting (15, 16, 25) in (29) we have:

$$
\begin{align*}
& \frac{\min _{p \in P} \int\left(\theta-\theta^{*}\right)^{+} p(d \theta)+\max _{p \in P} \int\left(\theta-\theta^{*}\right)^{+} p(d \theta)}{2} \\
& -\frac{\min _{p \in P} \int\left(\theta^{*}-\theta\right)^{+} p(d \theta)+\max _{p \in P} \int\left(\theta^{*}-\theta\right)^{+} p(d \theta)}{2} \\
& =  \tag{30}\\
& =\theta_{0} e^{r T}-\theta^{*}
\end{align*}
$$

Provided the choice of the current stock price $\theta_{0}$ has been made by the market participants, this condition on statistical regularity $\boldsymbol{P}(\boldsymbol{\Theta})$ is the nonstochastic analogue of the no-arbitrage condition on the pricing measure of the usual stochastic case. Indeed, in stochastic case we would have from (30), as well as from (22) and (24),

$$
\begin{equation*}
\int \theta p(d \theta)=\theta_{0} e^{r T} \tag{31}
\end{equation*}
$$

a condition that locks in the expectation of the stock price and yields the Black-Scholes pricing formula for a log-normal $p(\theta)$ with a volatility parameter $\sigma$.

### 2.4. Generalized delta

In the general case of statistical regularity $\boldsymbol{P}$ the minimal expected profits (3') of the uncovered European option position with pay-off $f(\theta)$ are equal for the buyer and the seller, and are always negative. Substituting (7) back into (5) one has

$$
\begin{gather*}
L_{Z_{s}}^{*}\left(u^{*}\right)=L_{Z_{b}}^{*}\left(u^{*}\right)=\min _{p \in P} \int L_{b}\left(\theta, u^{*}\right) p\left(d \theta_{T}\right)=\min _{p \in P} \int f(\theta) p(d \theta)-u^{*} e^{r T}= \\
=\min _{p \in P} \int f(\theta) p(d \theta)-\frac{\min _{p \in P} \int f(\theta) p(d \theta)+\max _{p \in P} \int f(\theta) p(d \theta)}{2}= \\
=\frac{\min _{p \in P} \int f(\theta) p(d \theta)-\max _{p \in P} \int f(\theta) p(d \theta)}{2} \leq 0 . \tag{32}
\end{gather*}
$$

When statistical regularity $\boldsymbol{P}$ is stochastic, the minimal expected pay-off of the uncovered option position is zero. Relation (32) shows that the use of stochastic probability measures may be the reason why the markets systematically underestimate the risks of financial transactions.

It is important to note that it is not yet clear if in the case of statistical regularities of the general form the dynamic hedging framework is theoretically possible. First of all, we simply do not know yet how to describe the time evolution of statistical regularity. Second, from the valuation view point, the delta hedge construction is not necessary in order to determine the price of the call option: the price is given by (15) and (24). Third, from risk management perspective the knowledge of the delta is important, but even having obtained the pricing relations, the derivation of the option price sensitivity to the change of the underlying price is not trivial.

Nevertheless we can define a static hedge, or, as one may call it, a generalized delta. Indeed let us compose a portfolio of a European option and of a position in the underlying and require that its minimal expected pay-off (12) were not negative. Represent this portfolio construction as a problem of choice (11). Let $M=1, \Theta=\Theta_{1}, d=\left(d_{1}, d_{2}\right) \in D \times D=D^{2} \in$ $D^{(\infty)}, d_{1}=\left(f(), 1,. u^{*}\right), d_{2}=\left(g(),. \delta, \theta_{0}\right)$, where $f(\theta)$ is as above, $g(\theta)=\theta$ in order to represent a position on the underlying itself, $u^{*}$ is chosen as (7), $\theta_{0}$ is chosen as (22) and $\delta \in Q$. Then, according to (10'),

$$
\begin{equation*}
L(\theta, d)=\sum_{i=1}^{2} q_{i}\left(-u_{i} e^{r T}+f_{i}(\theta)\right)=-u^{*} e^{r T}+f(\theta)+\delta\left(-\theta_{0} e^{r T}+\theta\right) \tag{33}
\end{equation*}
$$

and the decision scheme is given as $Z=\left(\Theta, D^{2}, L(),\right)$. The criterion (12) now is

$$
\begin{gathered}
L_{Z}^{*}(d)=\min _{p \in P} \int L(\theta, d) p(d \theta)= \\
=\min _{p \in P}\left(\left(\int f(\theta) p(d \theta)-u^{*} e^{r T}\right)+\delta\left(-\theta_{0} e^{r T}+\int \theta p(d \theta)\right)\right) \geq \\
\geq \min _{p \in P}\left(\int f(\theta) p(d \theta)-u^{*} e^{r T}\right)+\delta \min _{p \in P}\left(-\theta_{0} e^{r T}+\int \theta p(d \theta)\right)= \\
=\min _{p \in P} \int f(\theta) p(d \theta)-u^{*} e^{r T}+\delta\left(-\theta_{0} e^{r T}+\min _{p \in P} \int \theta p(d \theta)\right)= \\
=\frac{\min _{p \in P} \int f(\theta) p(d \theta)-\max _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta)}{2}+\delta \frac{\min _{p \in P} \int \theta p(d \theta)-\max _{p \in P} \int \theta p(d \theta)}{2}
\end{gathered}
$$

Requiring the last sum be equal zero, one obtains

$$
\begin{equation*}
\delta=-\frac{\max _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta)-\min _{p \in \boldsymbol{P}} \int f(\theta) p(d \theta)}{\max _{p \in \boldsymbol{P}} \int \theta p(d \theta)-\min _{p \in \boldsymbol{P}} \int \theta p(d \theta)} \tag{34}
\end{equation*}
$$

This quantity of the underlying makes the minimal expected pay-off of the portfolio $d=\left(d_{1}, d_{2}\right) \in D \times D \in D^{(\infty)}$ non-negative. Are there conditions that make (34) converge to the standard Black-Scholes delta, $\delta \rightarrow \delta_{B S}$ ? What does (34) converge to when $P=\mu, \forall p \in$ $P$, that is when random phenomena is $\mu$-stochastic in the sense of the Definition 4.8 from the book Ivanenko (2010)? When $P=\mu$, does $\delta=\frac{\mu(\partial f)}{\mu(\partial \theta)}$ ? These questions are important because statistical regularities of the general form correspond to statistically unstable random phenomena, and hence, one is tempted to say, more real, than stochastic processes. This is yet another argument in order to question the dynamic hedging framework as it is done in Derman and Taleb (2008).

## 3. Discussion

The theoretical elements presented in this article are an adaptation of the general decision theory described in Ivanenko (2010) to the problem of valuation of derivative contracts, a traditional problem of mathematical finance. In its essence this adaptation is an interpretation of observed phenomenon, namely of an observed transaction price. However, financial decision makers may belong to different classes of decision makers, not necessarily to the
class $\Pi_{l}$, and thus may use very different criteria for their actions, if any. They may have as well very different views on the type of behavior of random variables. And transactions may still take place. The benefit from being a member of this class is obvious: one is prepared, in a manner of speaking, from the onset to statistically unstable random outcomes. The representatives of the class $\Pi_{0}$, that is those who follow the guidance of the expected utility criterion, are devoid of this benefit.

It seems that considering pricing problem as a problem of choice, or a decision, is a natural framework for static replication argument, as presented in Derman and Taleb (2008), and allows for a coherent introduction of the concept of nonstochastic randomness in mathematical finance.

In conclusion we would like to stress that due to Ivanenko $(1990,2010)$ the families of finitely-additive probabilities, that appear in decision theory (see Ivanenko (1986), Gilboa (1989)) but that have never been accepted in mathematical finance, have finally acquired their statistical meaning: they describe statistically unstable, or nonstochastic, random phenomena. Therefore it seems reasonable to suggest that further exploration of statistical regularities of nonstochastic randomness, besides being a new research topic, may happen to be a road away from current underestimating of risks of financial transactions as well as one more argument in favor of those who disbelieve stochastic character of financial variables.

## Appendix

## On the definition of the class $\Pi_{1}$

We argue that decision makers belonging to the class $\Pi_{1}$ are those who would systematically prefer neutral or fully diversified investment strategy to a directional one. Namely, Condition 3 of the following definition, called in Ivanenko (1986, 2010) the guaranteed result principle generalized for mass events and reflecting uncertainty aversion of the decision maker, can be interpreted as the diversification argument, encouraging an investor to pursue neutral strategies. Below we reproduce the axiomatic definition of the class $\Pi_{l}$ and its characterization theorem from Ivanenko (1986, 2010).

Definition A1 Let $\mathbb{Z}$ be the class of all ordered triples of the form $Z=(\Theta, U, L)$, where $\Theta, U$ are arbitrary nonempty sets and $L: \Theta \times U \rightarrow \mathbb{R}$ is a real bounded function. The triple $Z$ is called a decision scheme. We denote by $\mathbb{Z}(\Theta)$ the subclass of all decision schemes of the form $Z=(\Theta, . .$.$) , where the set \Theta$ is fixed.

Definition A2 We define a criterion choice rule to be any mapping $\pi$, defined on $\mathbb{Z}(\Theta)$ and associating to every scheme $Z=(\Theta, U, L)$ some real function $L_{Z}^{*}(\cdot)$, a criterion, determined on $U$. We denote the class of all criterion choice rules by $\Pi(\Theta)$ and include in the subclass $\Pi_{1}(\Theta) \subset \Pi(\Theta)$ all criterion choice rules that satisfy the following three conditions:

C1. If $Z_{i}=\left(\Theta, U_{i}, L_{i}\right) \in \mathbb{Z}(\Theta), i=1,2, U_{1} \subset U_{2}$, and $L_{1}(\theta, u)=L_{2}(\theta, u) \forall u \in U_{1}, \forall \theta \in$ $\theta$, then $L_{Z_{1}}^{*}(u)=L_{Z_{2}}^{*}(u) \forall u \in U_{1}$.

C2. If $Z=(\Theta, U, L) \in \mathbb{Z}(\Theta), u_{1}, u_{2} \in U$, then if $L\left(\theta, u_{1}\right) \leq L\left(\theta, u_{2}\right), \forall \theta \in \Theta$, then $L_{Z}^{*}\left(u_{1}\right) \leq L_{Z}^{*}\left(u_{2}\right)$, and if $a, b \in \mathbb{R}, a \geq 0$ and $L\left(\theta, u_{1}\right)=a L\left(\theta, u_{2}\right)+b, \forall \theta \in \Theta$, then $L_{Z}^{*}\left(u_{1}\right)=a L_{Z}^{*}\left(u_{2}\right)+b$.

C3. If $Z=(\Theta, U, L) \in \mathbb{Z}(\Theta), u_{1}, u_{2}, u_{3} \in U$ and $L\left(\theta, u_{1}\right)+L\left(\theta, u_{2}\right)=2 L\left(\theta, u_{3}\right) \forall \theta \in$ $\theta$, then $2 L_{Z}^{*}\left(u_{3}\right) \leq L_{Z}^{*}\left(u_{1}\right)+L_{Z}^{*}\left(u_{2}\right)$.

The next Theorem (which is a simplified version of Theorem 1 from Ivanenko (1986) or Theorem 5.2 from Ivanenko (2010) establishes the link between the properties of $L_{Z}^{*}(\cdot)$ and its structure.

Theorem Criterion $L_{Z}^{*}(\cdot)$ possesses the properties C1-C3 if and only if it has the following structure

$$
\begin{equation*}
L_{Z}^{*}(u)=\max _{p \in P} \int L(\theta, u) p(d \theta) \tag{A1}
\end{equation*}
$$

where $P$ is a statistical regularity on $\Theta$ in the form of a closed family of finitely-additive probability measures.

Note that in the above definition and theorem the sets $U, \Theta$ are arbitrary nonempty sets and the loss function $L: U \times \Theta \rightarrow \mathbb{R}$ is bounded.

If instead of the loss function $L$ one considers profit and loss function $\hat{L}=-L$, as we do in this article, then condition C3 of the above Definition becomes

C3'. If $Z=(\Theta, U, \hat{L}) \in \mathbb{Z}(\Theta), u_{1}, u_{2}, u_{3} \in U$ and $\hat{L}\left(\theta, u_{1}\right)+\hat{L}\left(\theta, u_{2}\right)=2 \hat{L}\left(\theta, u_{3}\right) \forall \theta \in$ $\Theta$, then $2 L_{Z}^{*}\left(u_{3}\right) \geq L_{Z}^{*}\left(u_{1}\right)+L_{Z}^{*}\left(u_{2}\right)$.

The criterion (A1) of maximal expected losses then becomes the criterion of minimal expected utility (see Ivanenko (1986), Gilboa (1989))

$$
L_{Z}^{*}(u)=\min _{p \in \boldsymbol{P}} \int \hat{L}(\theta, u) p(d \theta)
$$

Below we demonstrate that if a decision maker uses criterion (A1'), then Condition 3 can be interpreted as the diversification argument (see as well Ivanenko (2010), page 111).

Let $F=\{\Theta \rightarrow \mathbb{R}\}$ be the set of all bounded real functions on $\Theta$ that we interpret as payoffs of financial instruments, depending on a random parameter $\theta \in \Theta^{6}$. Let $U=\{u: u \in \mathbb{R}\}$ be the set of prices of the pay-offs $F$. Let $q \in Q=\{-1,0,+1\}$ represent the sense of the operation, -1 for a sale, +1 for a purchase and 0 for the absence of operation. Then let $D=F \times Q \times U$ be the set of decisions. This means that a decision maker chooses three elements: a contract, her role and the price. Supposing that the decision maker finances her purchases via a bank account and places the proceeds of sales there as well, the profit and loss function will be

$$
\begin{equation*}
L(\theta, d)=q\left(-u e^{r T}+f(\theta)\right), d=(q, u, f(.)) \in D, f \in F . \tag{A2}
\end{equation*}
$$

In this way one constructs in the matrix decision scheme $Z=(\Theta, D, L(\ldots))$, an essential element of decision system.

Let

$$
\begin{equation*}
d_{1}=\left(+1 ; f_{1}(.) ; u_{1}\right), \quad L\left(\theta, d_{1}\right)=-u_{1} e^{r T}+f_{1}(\theta) \tag{A3}
\end{equation*}
$$

$$
\begin{equation*}
d_{2}=\left(+1 ; f_{2}(.) ; u_{2}\right), \quad L\left(\theta, d_{2}\right)=-u_{2} e^{r T}+f_{2}(\theta) \tag{A4}
\end{equation*}
$$

$$
\begin{align*}
& d_{3}=\left(+1 ; \frac{f_{1}(.)+f_{2}(.)}{2} ; \frac{u_{1}+u_{2}}{2}\right)  \tag{A5}\\
& L\left(\theta, d_{3}\right)=-\frac{u_{1}+u_{2}}{2} e^{r T}+\frac{f_{1}(\theta)+f_{2}(\theta)}{2}
\end{align*}
$$

Then it is obvious that

$$
\begin{equation*}
L\left(\theta, d_{1}\right)+L\left(\theta, d_{2}\right)=2 L\left(\theta, d_{3}\right), \forall \theta \in \Theta . \tag{A6}
\end{equation*}
$$

Provided

$$
\begin{equation*}
L_{Z}^{*}(d)=\min _{p \in P} \int L(\theta, d) p(d \theta) \tag{A7}
\end{equation*}
$$

show that

$$
\begin{equation*}
L_{Z}^{*}\left(d_{1}\right)+L_{Z}^{*}\left(d_{2}\right) \leq 2 L_{Z}^{*}\left(d_{3}\right) \tag{A8}
\end{equation*}
$$

Indeed,

[^3]\[

$$
\begin{align*}
& L_{Z}^{*}\left(d_{1}\right)=\min _{p \in P} \int L\left(\theta, d_{1}\right) p(d \theta)=-u_{1} e^{r T}+\min _{p \in \boldsymbol{P}} \int f_{1}(\theta) p(d \theta)  \tag{A9}\\
& L_{Z}^{*}\left(d_{2}\right)=\min \int L\left(\theta, d_{2}\right) p(d \theta)=-u_{2} e^{r T}+\min _{p \in \boldsymbol{P}} \int f_{2}(\theta) p(d \theta)  \tag{A10}\\
& L_{Z}^{*}\left(d_{3}\right)=\min _{p \in P} \int L\left(\theta, d_{3}\right) p(d \theta)= \\
& =-\frac{u_{1}+u_{2}}{2} e^{r T}+\min _{p \in \boldsymbol{P}} \int \frac{f_{1}(\theta)+f_{2}(\theta)}{2} p(d \theta)
\end{align*}
$$
\]

Since

$$
\begin{equation*}
\min _{p \in P} \int f_{1}(\theta) p(d \theta)+\min _{p \in P} \int f_{2}(\theta) p(d \theta) \leq \min _{p \in P} \int\left(f_{1}(\theta)+f_{2}(\theta)\right) p(d \theta) \tag{A12}
\end{equation*}
$$

we obtain (A8).
This confirms that to use criterion (Al), or (Al'), in order to estimate decisions ex ante and to prefer in situations of uncertainty neutral or fully diversified strategies are equivalent.

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[^0]:    ${ }^{1}$ Disclaimer: the ideas expressed and the results obtained in this article reflect the views of the author and are not necessarily shared by the institution to which the author is affiliated

[^1]:    ${ }^{2}$ Unlike stochastically random phenomena, nonstochastically random phenomena are those random phenomena that do not possess the property of stability of frequencies. A model of realization of such a phenomenon is constructed without usual probabilistic assumptions and is called sampling directedness, or directedness in the space of samples. The theorem that states that any sampling directedness has a statistical regularity, and any statistical regularity corresponds to a sampling directedness is proved in Ivanenko (1990, 2010). Note that the words directedness, net and generalized sequence are synonyms.
    ${ }^{3}$ This is achieved by means of the change of sign of the loss function. In this connection see as well De Groot (1970), Ivanenko (1986), Gilboa (1989) and the Appendix.
    ${ }^{4}$ When the buyer and the seller belong to different classes of decision makers and do not share the views on randomness, the equality of this type would not make sense. What is, probably, more important, the buyer and the seller may have very different asset-liability and financing constraints, resulting in diversity of the profit and loss functions $L$.

[^2]:    ${ }^{5}$ We consider that in this case transaction does not happen, which is in line with the efficiency market hypothesis.

[^3]:    ${ }^{6}$ Generally speaking, the elements of $F$ are the elements of the Banach space of all real bounded functions on $\Theta$.

