

In some space-time probability measures beneath Black-Scholes type formulae.

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1. The Black-Scholes paradigm.

(1.1) The paper originates from the following result:

consider the put parity:

$$C^-(K, t) = E[(K - E_t)^+], \quad (1)$$

where $K \geq 0$, and $E_t = \exp(Bt - \frac{1}{2}t^2)$, $t \geq 0$, with $(Bt) t \geq 0$

Standard Breiman method. Then exist a probability, which we denote by \mathbb{Q} .

Theorem 1:

on $[0, 1] \times [0, \infty)$ such that:

$$C^-(K, t) = \mathbb{Q}([0, K] \times [0, t]) \quad (2)$$

We give three descriptions of this probability:

Definition #1: (3) $\mathbb{Q}(dK, dt) = \frac{1}{2} E[dK, dt](K)$

where $(K) \in \mathbb{R}$ is the family of local times of $(E_t, s \geq 0)$; a little more formally:

$$\int_0^\infty \int_0^\infty q(K, dt) dK, dt(K)$$

extends as a probability, for any function g to ≥ 0 Borel one

(Then, we should refer to [3], and [4])

May 26th, 2008

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Description # 2: The probability γ in $[0, 1] \times [0, \infty)$ is the

law of the pair: (4) $\left(\text{map} \left\{ \exp(-2E), \exp(-\sqrt{2E})(4B_1^2) \right\}, 4B_1^2 \right)$

where B_1, E, E' are independent, with $E, E' \sim 2$ standard exponential variables -

Description # 3: The probability γ in $[0, 1] \times [0, \infty)$ is the law

of the pair: (5) $\left(U, \gamma \left(\frac{1}{2}, E(U) \right) \right)$, $f(x) = \log(1/x)$

where U is uniform on $[0, 1]$, independent of the "brown" $(R_{a,b}, a, b > 0)$

which is now defined. Lemma: \mathcal{N} the distribution function of the standard Gaussian variable, is:

(6) $\mathcal{N}(x) = \frac{1}{\sqrt{2\pi}} \int_x^{-\infty} dy \exp(-y^2/2)$

Then, for any $a, b > 0$, one can associate a p.v. $\gamma_{a,b}$ taking values

(7) $\mathcal{P}(\gamma_{a,b} \leq t) = \mathcal{N}\left(a\sqrt{t} - \frac{\sqrt{t}}{b}\right), t \geq 0$

(8) $\frac{d}{dt} \mathcal{P}(\gamma_{a,b} \leq t) = \frac{\sqrt{2\pi}}{1} e^{ab} \exp\left(-\frac{1}{2}\left(a\sqrt{t} + \frac{\sqrt{t}}{b}\right)^2\right) \left\{ \frac{2\sqrt{t}}{a} + \frac{b}{2\sqrt{t}} \right\}$

(9) $\frac{d}{dt} \mathcal{P}(\gamma_{a,b} \leq t) = \frac{1}{2} \left[\mathcal{P}(T_{(a)} \leq t) + \mathcal{P}(G_{(a)} \leq t) \right]$

Remark: It may be worth mentioning that this lemma admits a wide extension, since

to any distribution function F on \mathbb{R} , and any $a, b > 0$, we can associate a new distribution function $F_{a,b}$ on \mathbb{R}_+

$$F_{a,b}(t) = F\left(a\sqrt{t} - \frac{\sqrt{t}}{b}\right), \quad t \geq 0$$

However, the particular case $F \equiv W$ is extremely well fitted within our discussion.

Upon comparing the descriptions #2 and #3, it is quite natural to look for some understanding of the implied identities in law between the first, resp. second, components of (4) and (5); precisely, we wish to check directly that:

$$(14) \quad \max \left\{ \text{sup}(-2\epsilon), \text{sup}(-\sqrt{(2\epsilon)^2 4B_2^2}) \right\} \stackrel{(\text{law})}{=} U$$

as well as:

$$(15) \quad \text{sup} \left(\frac{1}{2}, \epsilon(t) \right) \stackrel{(\text{law})}{=} (4B_2^2)^{-1/2} \int_0^t \text{sup} \left\{ t: B_t - \frac{t}{2} = 0 \right\}$$

[Indeed, I should have written following notation in the remark:]

$$G^{(-1/2)}$$

- Checking (14) is easy, since: $\Phi \stackrel{(\text{law})}{=} \sqrt{2\epsilon^2 B_2^2}$, this is a probability transition, this is a probability transition, see, e.g., Chacón-Kifer of the duplication formula for the game function, see, e.g., Chacón-Kifer [1]; Exercise (??).

Thus, (14) with: $\max(U_2, (U_1)^2) \stackrel{(\text{law})}{=} U$, which is obvious, by looking at the distribution function: $(\cdot)^2 \stackrel{(\text{law})}{=} U_1$ and $(\cdot) \stackrel{(\text{law})}{=} U_2$.

$$(\cdot)^2 \stackrel{(\text{law})}{=} U_1 \quad (\cdot) \stackrel{(\text{law})}{=} U_2 \quad (\cdot) \stackrel{(\text{law})}{=} U$$

- Checking the validity of (15) is a little less immediate!
 From the description (9) of the law of $Y_{a,b}$, we see that (15)

amounts to:

$$P(G_{(-1/2)}^0 \in dt) = \frac{1}{2} \int_1^0 du \left\{ P(T_{(1/2)}^{(u)}) e^{dt} + P(G_{(1/2)}^{(u)} \in dt) \right\}$$

$$(16) \quad = \frac{1}{2} \int_0^\infty dx e^{-x} \left\{ P(T_{(1/2)}^x) e^{dt} + P(G_{(1/2)}^x \in dt) \right\}$$

We shall use the well-known formulae:

$$(a > 0) \quad P(T^a \in dt) = \frac{d}{dt} \frac{\sqrt{2\pi t^3}}{a} e^{-\frac{at}{2}}$$

$$(17) \quad P(G^a \in dt) = |v| \frac{d}{dt} \frac{\sqrt{2\pi t}}{(a-vt)^2} \exp\left(-\frac{at}{2}\right)$$

Due to (17), (16) is equivalent to:

$$\left\{ \frac{1}{2} \int_0^\infty dx e^{-x} \exp\left(-\frac{x}{2}\right) \left[\frac{d}{dt} \frac{\sqrt{2\pi t^3}}{2} \right] - \frac{1}{2} \int_0^\infty dx e^{-x} \left[\frac{d}{dt} \frac{\sqrt{2\pi t}}{2} \right] \right\} e^{-\frac{at}{2}}$$

(18)?

$$\left\{ \frac{(a^2 - tx)}{(2t)} - \frac{(x^2 - tx)}{(2t)} \right\} e^{-\frac{at}{2}}$$

Amplify into:

$$\left\{ \frac{t}{2} e^{-\frac{at}{2}} - \frac{(x^2 - tx)}{(2t)} e^{-\frac{at}{2}} + \frac{1}{2} e^{-\frac{at}{2}} \right\}$$

The RHS equals:

$$\int_0^\infty dx e^{-x} \left[\frac{t}{2} e^{-\frac{at}{2}} - \frac{(x^2 - tx)}{(2t)} e^{-\frac{at}{2}} + \frac{1}{2} e^{-\frac{at}{2}} \right]$$

Now, integrate by parts yields:

$$= 1 - \int_0^\infty dx \left(\frac{1}{2} \right) e^{-x/2} e^{-x^2/2t}$$

Consequently, we have indeed checked (16)

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(1.1) An extension of Theorem 1. In the next statement we shall replace the standard Brownian martingale

$(\mathcal{E}_t, \mathcal{F}_t)$ by the semimartingale:

$$\mathcal{E}_{\sigma, \nu}^t = \exp(\sigma B_t - \nu t). \text{ Then, we}$$

can show:

Theorem 2:

There exists a probability $\mathbb{P} \sim \mathbb{P}$ on $[0, 1] \times [0, \infty)$, which shall

denote by $\mathcal{P}_{\sigma, \nu}$ and that:

$$(20) \quad \mathbb{E}_{\mathcal{P}_{\sigma, \nu}}[K, t] = \mathbb{E}[(K - \mathcal{E}_{\sigma, \nu}^t)^+] = \mathcal{P}_{\sigma, \nu}([0, K] \times [0, t])$$

Moreover, $\mathcal{P}_{\sigma, \nu}$ is the law of

$$(21) \quad (U, Y_{\frac{\sigma}{\nu}}(\frac{\sigma}{\nu}, \frac{\sigma}{\nu})) \quad (\mathcal{L}(U) = \mathcal{L}_n(\frac{U}{\nu})).$$

where U is uniform on $[0, 1]$, independent of the process $(Y_a, a, b > 0)$ introduced

in the preceding lemma.

Proof: From the definition (20), we may write:

$$\mathbb{E}[(K - \mathcal{E}_{\sigma, \nu}^t)^+] = \mathbb{E}[\int_0^t dx \mathbb{1}_{(\sigma B_t - \nu t < (\ln(x)))]]$$

$$= \int_0^t dx \mathbb{P}(B_1 < \sqrt{t} - \frac{\sigma}{\nu} \sqrt{t})$$

$$= \int_0^t dx \mathbb{P}(Y_{\frac{\sigma}{\nu}}(\frac{\sigma}{\nu}, \frac{\sigma}{\nu}) \leq t),$$

which implies the description (21).

Note that (21) corresponds to Description (#3) in the particular case $\sigma=1, \nu=1/2$.

We would like to see whether there is a description #2; in particular

what is the law of $(\frac{\sigma}{\nu}, \mathcal{L}(U))$?

$$= - \left\{ u(x) v'(x) \right\}_{x=0}^{x=\infty} - \int_0^{\infty} dx v''(x) u(x) \cdot \left\{ u'(x) \right\}$$

$$= - \int_0^{\infty} dx e^{-\mu x} \left[-\frac{t \sigma^2}{x} e^{-\frac{x^2}{2\sigma^2}} \right]$$

$$= \int_0^{\infty} dx e^{-\mu x} \left(\frac{t \sigma^2}{x} \right) e^{-\frac{x^2}{2\sigma^2}}$$

where: $I_1 = \int_0^{\infty} dx e^{-\mu x} \left(\frac{t \sigma^2}{x} \right) e^{-\frac{x^2}{2\sigma^2}}$

We now introduce the parameter: $\mu = \left(1 - \frac{\nu}{\sigma^2}\right)$, and we compute the integral:

$$= \left(\frac{2}{\sigma} \right) \frac{\sqrt{2\pi} t}{\sigma} \exp\left(-\frac{\nu^2 t}{2\sigma^2}\right) \int_0^{\infty} dx e^{-x^2} \left\{ \frac{t \sigma^2}{x} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu^2}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu^2}{\sigma^2} \right\}$$

$$= \left(\frac{2}{\sigma} \right) \frac{1}{\sqrt{2\pi} t} \exp\left(-\frac{\nu^2 t}{2\sigma^2}\right) \int_0^{\infty} dx e^{-x^2} \left\{ \frac{t}{x} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu^2}{\sigma^2} e^{-\frac{x^2}{2\sigma^2}} + \frac{\nu^2}{\sigma^2} \right\}$$

$$= \frac{2}{\sigma} \int_0^{\infty} dx e^{-x^2} \left\{ \frac{\sqrt{2\pi} t^3}{(\sigma)} e^{-\frac{(x-\nu t)^2}{2\sigma^2}} + \frac{\sigma}{2} \frac{1}{\sqrt{2\pi} t} e^{-\frac{(x-\nu t)^2}{2\sigma^2}} \right\}$$

$$\left(= \frac{2}{\sigma} \int_0^{\infty} dy e^{-y^2} \left\{ P\left(\frac{\nu}{\sigma}\right) + P\left(\frac{\nu}{\sigma}\right) e^{-y^2} \right\} \right)$$

$$= \frac{1}{\sigma} \int_0^{\infty} dx e^{-x^2} \left\{ P\left(\frac{\nu}{\sigma}\right) + P\left(\frac{\nu}{\sigma}\right) e^{-x^2} \right\}$$

$$\Phi_{\nu, \sigma}(t) = \frac{1}{2} \int_{-\nu}^{\nu} du \left\{ P\left(\frac{\nu}{\sigma}\right) + P\left(\frac{\nu}{\sigma}\right) e^{-\frac{u^2}{\sigma^2}} \right\}$$

this for us:

Thus: $I_1 = 1 + \int_{-\infty}^{\infty} dx v'(x)u(x)$

$$= 1 + \int_{-\infty}^{\infty} -\mu dx e^{-\mu x} - \alpha^2/2\sigma^2 e^{-\alpha^2/2\sigma^2}$$

Finally:

$$I_1 + \frac{\sigma^2}{2} I_2 = 1 - \mu I_2 + \frac{\sigma^2}{2} I_2 \quad (\text{note: } \mu = 1 - \frac{\sigma^2}{2})$$

(We note that the simplification is valid when: $\frac{\sigma^2}{2} = 1$, in agreement with our previous computation: $\nu = 1/2, \sigma = 1$.)

In any case, we have obtained that the second moment of $\mathcal{L}_{\sigma, \nu}$ is

$$(2.2) \quad \theta_{\sigma, \nu}(\frac{1}{2} dt) = \left(\frac{dt}{2}\right)^{\frac{\nu}{\sigma}} \frac{\sqrt{2\nu t}}{\sigma} e^{-\left(\frac{\nu t}{2\sigma^2}\right)} \left\{ 1 + \left(\frac{\sigma^2}{2\nu} - 1\right) \int_{-\infty}^{\infty} dx e^{-\mu x} - \frac{\alpha^2}{2\sigma^2} \right\}$$

(1.3) A signed measure on $\mathbb{R}^+ \times \mathbb{R}^+$.
 Interpretation??

In this paragraph, we show that the formula:

$$(2) \quad G_-(k, t) \equiv E[(k - \varepsilon_t)^+] = \mathcal{L}([0, k] \times [0, t])$$

is true for all values $k, t \geq 0$, with \mathcal{L} a signed measure on $\mathbb{R}^+ \times \mathbb{R}^+$, such that: \mathcal{L} is a probability.

In fact, since $E[(k - \varepsilon_t)^+] = \int_{\mathbb{R}^+} dx P(\varepsilon_t \leq x)$

$$(2.3) \quad = \int_0^k dx \mathcal{N}\left(\frac{x}{\sqrt{t}} + \frac{\ln(x)}{\sqrt{t}}\right)$$

We get that: $\mathcal{L}(dx, dt) = g(k, t) dx dt$, with:

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$$g(k, t) = \frac{\partial^2}{\partial k^2} (E[(k - \epsilon_t)^+])$$

which, from (23), yields:

$$g(k, t) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{k}{\sqrt{t}} + \frac{\sqrt{t}}{2}\right)^2\right) \cdot \left\{ -\frac{(k/t)}{2t^{3/2}} + \frac{1}{4t^{1/2}} \right\}$$

$$= \frac{1}{2} \left\{ \frac{1}{\sqrt{2\pi t}} \exp\left(-\frac{1}{2t} \left(\ln \frac{k}{t} - \frac{t}{2}\right)^2\right) \right\}$$

$$\left\{ \frac{1}{2} + \frac{\ln(k/t)}{t} \right\}$$

$$g(k, dt) = \begin{cases} [k < 1] \left\{ \frac{1}{2} P\left(T(1/2) \leq \ln \frac{k}{t}\right) + P\left(G(1/2) \leq \ln \frac{k}{t}\right) \right\} e^{-dt} \\ [k > 1] \left\{ \frac{1}{2} P\left(G(1/2) \leq \ln \frac{k}{t}\right) - P\left(T(1/2) \leq \ln \frac{k}{t}\right) \right\} e^{-dt} \end{cases}$$

This is a remarkable formula which is quite close to our previous discussion in "Winfyung", in fact we should really compare this with theorem 0 in our [2].

References:

- [1] D Madan, B Roysette, H Jais: Option prices as probabilities - Finance Research Letters, 5, 79-87 (2008).
- [2] D Madan, B Roysette, H Jais: Winfying Black-Scholes type formulas which involve last passage times up to a strike (Asian-Pacific Financial Market) - Journal of Applied Probability, 45, 1000-1010 (2008).

1 Springer (2007)

[3] N Embrechts: Local time-space calculus for variable

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