



# It takes two: the break-even strike of a call option from joint physical and pricing density estimation

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# Outline

1. Concepts
2. Calculating Physical and Pricing densities
3. Data example

# Introduction

1

What normal people see:



What Financial Engineers see:





## $\mathcal{P}$ -world

- ▶ Physical world
- ▶ Actual world in which payoffs are realized
- ▶ Physical density  $p$  estimates real probabilities

$$\begin{aligned} &\text{Expected Payoff (t=0)} \\ &= e^{-rT} \cdot \mathbb{E}_{\mathcal{P}}[\text{payoff}] \end{aligned}$$

## $\mathcal{Q}$ -world

- ▶ Pricing world
- ▶ Artificial setting under which one determines the price
- ▶ Pricing density  $q$  reflects price a representative agent is willing to pay

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Under no-arbitrage condition



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## The European Call Option - Expected Payoff

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- ▶ European call option on asset  $S$  with maturity  $T$  and strike  $K$
- ▶ Payoff =  $(S_T - K)^+$



Expected Payoff European Call (at time 0)

$$= e^{-rT} \cdot \mathbb{E}_{\mathcal{P}}[(S - K)^+]$$

$$= e^{-rT} \cdot \int_{-\infty}^{\infty} (x - K)^+ p(x) dx$$

risk-free rate  $r$

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Do both worlds agree at some point?

### European call option

For a fixed maturity  $T$ , determine the **break-even strike**  $K_T$  such that

$$e^{-rT} \cdot \mathbb{E}_{\mathcal{P}}[(S - K_T)^+] = e^{-rT} \cdot \mathbb{E}_{\mathcal{Q}}[(S - K_T)^+]$$

→ Expectations in both worlds are equal

→ Expected return of European call =  $\frac{\text{Price} - \text{Expected Payoff}}{\text{Price}} := 0$

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→ Efficient estimation of the physical density  $p$  and pricing density  $q$  is needed

# Estimating Physical and Pricing densities

# Estimating the Physical and Pricing density

7

## Traditional approach

### Physical density

- ▶ Estimated based on historical data
  - backward looking
  - only one new observation each day

### Pricing density

- ▶ Estimated based on option data
  - forward looking
  - a number of new observations each day
- ▶ Depending on an asset pricing model

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## Traditional approach

Physical density  room for improvement!

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## Pricing density

- ▶ Estimated based on option data  rich source of information
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# Alternative Physical density estimation - Step 1

8

(Madan, Schoutens & Wang, 2020)

**Step 1:** Pricing density as U-shaped perturbation of physical density

## Assumptions:

- ▶ Investors are risk-averse
- ▶ Investors have heterogeneous beliefs
  - long position is allowed
  - short position is allowed

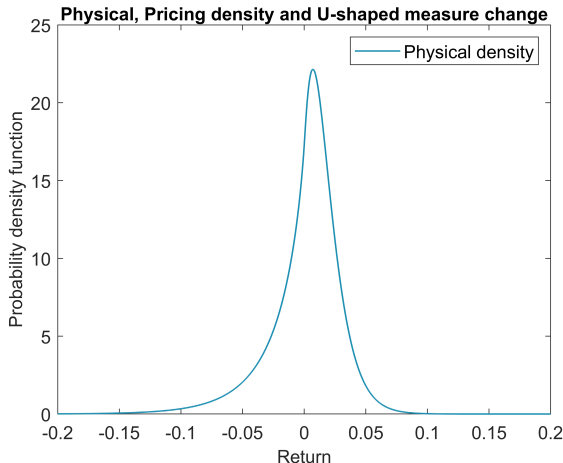
Empirical evidence in e.g. (Bakshi et al., 2010)

# Alternative Physical density estimation - Step 1

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(Madan, Schoutens & Wang, 2020)

- ▶ **Long investor:** wealth loss in negative return state  
→ loss protection leads to heavier left tail
- ▶ **Short investor:** wealth loss in positive return state  
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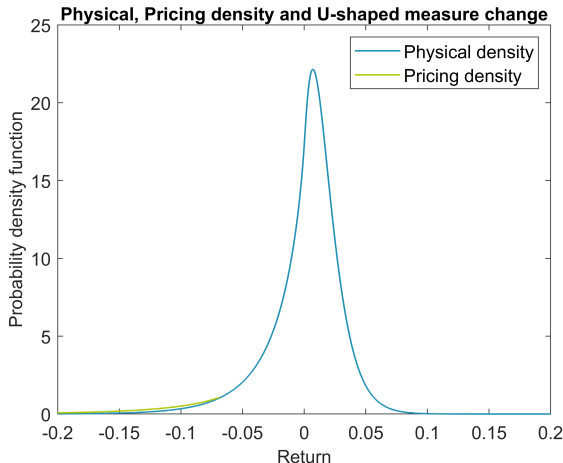


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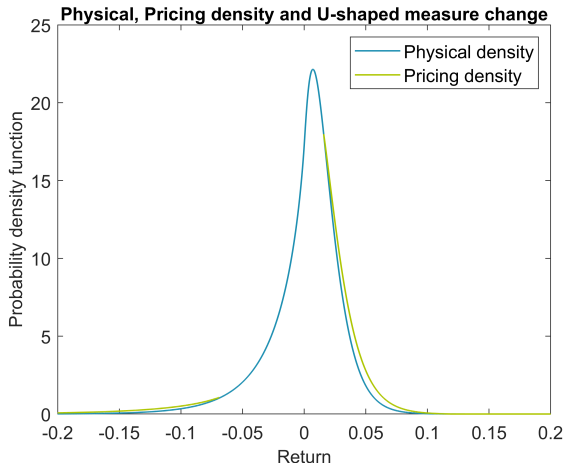


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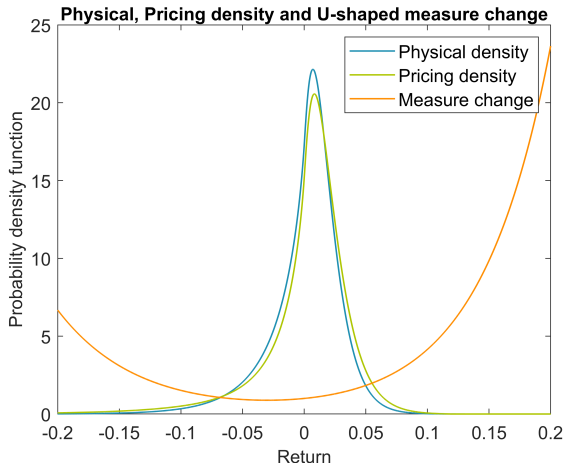


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## Alternative Physical density estimation - Step 1

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(Madan, Schoutens & Wang, 2020)

**Step 1:** Pricing density as U-shaped perturbation of physical density

$$q(x) = C \cdot \left( (1 - \alpha) \cdot e^{-\eta x} + \alpha \cdot e^{\zeta x} \right) \cdot p(x)$$

- ▶  $C$  := normalization constant
- ▶  $\eta$  := risk-aversion coefficient for being in a long position
- ▶  $\zeta$  := risk-aversion coefficient for being in a short position

### Step 2: Physical density follows a Bilateral Gamma model

- ▶ Bilateral Gamma (Küchler & Tappe, 2008) models the asset as

$$\log(S_t) = \log(S_0) + b_p \cdot \gamma_p(c_p t) - b_n \cdot \gamma_n(c_n t),$$

where  $\gamma_p$  and  $\gamma_n$  are two independent standard Gamma processes

- ▶ Substantiated by different speed and scale for upward and downward movements of a stock (Madan & Wang, 2017)
  - Escalator up
  - Elevator down

**Step 2:** Physical density follows a Bilateral Gamma model



Physical density  $p$

- ▶ Bilateral Gamma
- ▶ Characterized by  $[b_p, c_p, b_n, c_n]$

$$\cdot C \cdot \left( (1 - \alpha) \cdot e^{-\eta x} + \alpha \cdot e^{\zeta x} \right)$$



Pricing density  $q$

- ▶ Tilted Bilateral Gamma
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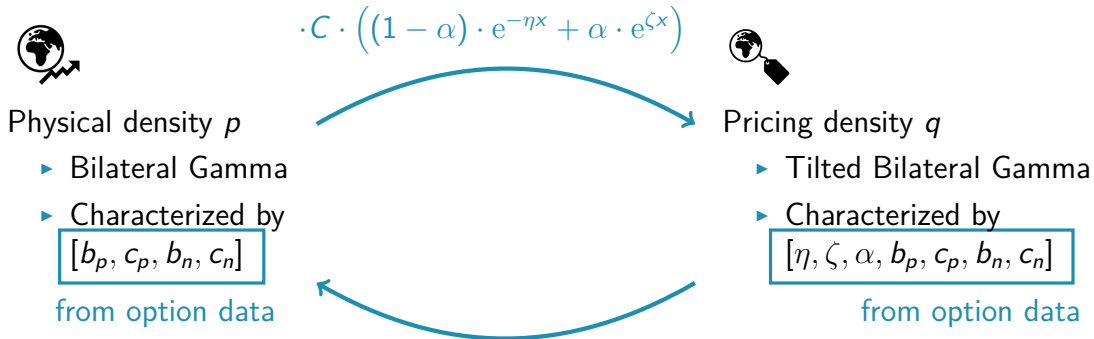
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# Expected return and break-even strike of a call option

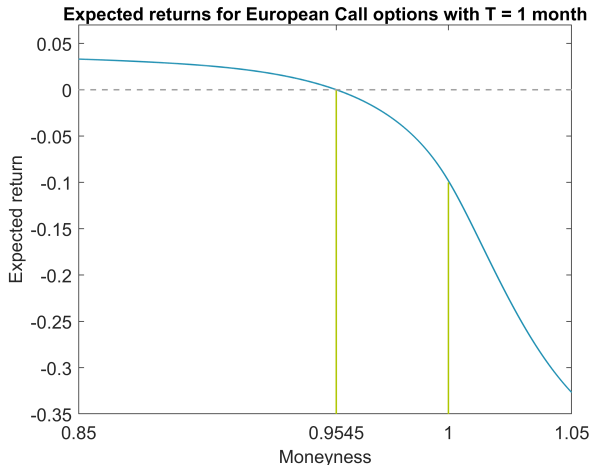
illustrated based on S&P 500 index option data

# Expected Return European Call on S&P 500 index

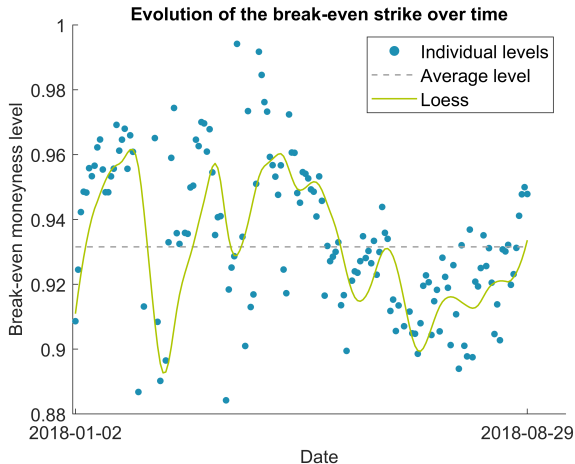
- March 15, 2018

13

- ▶ Fixed  $T = 1$  month,  
 $0.85 \leq \frac{K}{S_0} \leq 1.05$
- ▶ Expected return is **decreasing** with moneyness  
→ theoretical implication  
of U-shaped measure change  
(Bakshi et al., 2010)
- ▶ Break-even strike  
 $K_T = 0.95 \cdot S_0$



- ▶ Average break-even moneyness level of 93.15%
- ▶ Break-even strikes are located in-the-money
  - further away in-the-money call options are **cheap**
  - at-the-money and out-of-the-money call options are **expensive**
- ▶ Day-to-day fluctuations are small in absolute value



- ▶ The **Tilted Bilateral Gamma model** makes it possible to simultaneously estimate both physical and pricing density based on **option data** of the underlying asset
- ▶ This provides enough information to find the **break-even** strike of a call option
  - the data example shows a **rather stable pattern** over time
  - break-even strikes of S&P 500 index call options are **in-the-money**

Thank you!

- ▶ Bakshi G., Dilip B. M. & Panayotov G. (2010). Returns of claims on the upside and the viability of U-shaped pricing kernels. *Journal of Financial Economics*. 97(1). pp. 130–154.
- ▶ Carr P. & Madan D. B. (1998). Option valuation using the fast Fourier transform. *Journal of Computational Finance*. 2(4). pp. 61-73.
- ▶ Küchler U. & Tappe S. (2008). Bilateral gamma distributions and processes in financial mathematics. *Stochastic Processes and Their Applications*. 118(2). pp. 261–283.
- ▶ Madan D. B., Schoutens W. & Wang K. (2020). Bilateral Multiple Gamma Returns: Their Risks and Rewards. *International Journal of Financial Engineering*. 7(1).
- ▶ Madan D. B. & Wang K. (2017). Asymmetries in financial returns. *International Journal of Financial Engineering*. 04(04). 1750045.



- 1 Find **option data** with prices of call and put options
- 2 **Calibrate** the Tilted Bilateral Gamma model parameters
  - a. Calculate model prices of call options,  $EC(K, T)$ , with (Carr & Madan, 1998) formula

$$EC(K, T) = \frac{\exp(-\alpha \log(K))}{\pi} \int_0^\infty \exp(-i\nu \log(K)) \varrho(\nu) d\nu,$$

where

$$\varrho(\nu) = \frac{\exp(-rT) \mathbb{E}_{\mathcal{Q}}[\exp(i(\nu - (\alpha + 1)i) \log(S_T))]}{\alpha^2 + \alpha - \nu^2 + i(2\alpha + 1)\nu}$$

- b. Minimize distance between model prices and market prices
- 3 Use an **inverse Fourier transform** to find the pricing density  $q$  and physical density  $p$