

# Mathematical analysis of a class of path-dependent options

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# Obsah

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  - Asian options
- 2 Asian options**
  - PDE for Asian options
  - Algorithm
- 3 Numerical results**
  - Arithmetic averaged options
  - Exponentially weighted averaged options

## Financial derivatives

- Financial derivatives are financial instruments that are linked to a specific financial instrument or indicator or commodity, and which provide for market financial risk in a form that can be traded or otherwise offset in the market.
- The value of the financial derivative derives from the price of the underlying items.

## Usage

Financial derivatives are used for a number of purposes including risk management, hedging and **speculation**. Between financial derivatives belongs e.g. futures, forwards, **options** and swaps.

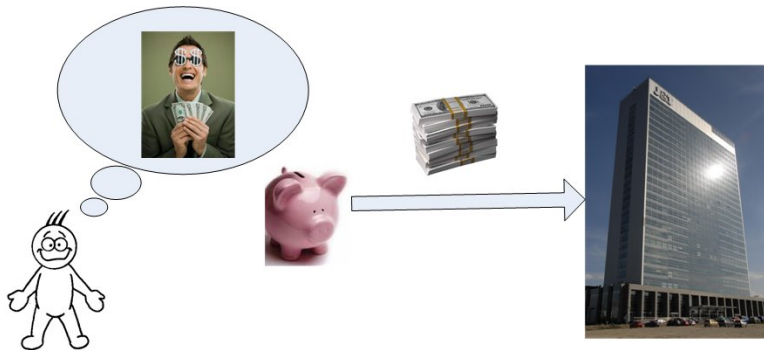
## Options

- Option is a right (**but not obligation**) to buy or sell an asset by a certain date for a predetermined price  $X$ .
- can be exercised before expiration date?
  - yes - American-style option
  - no - European-style option

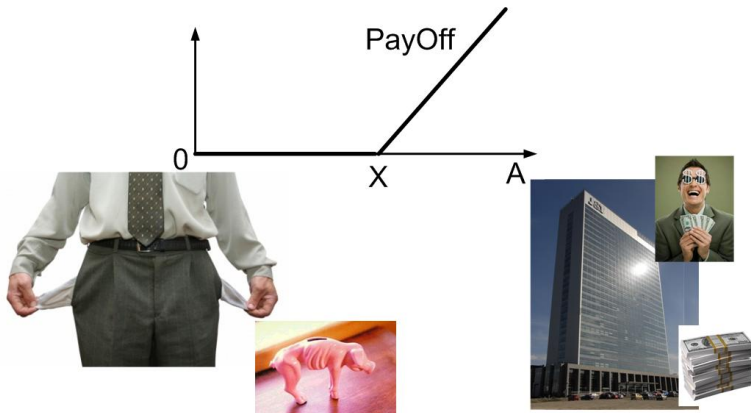
## Options' types

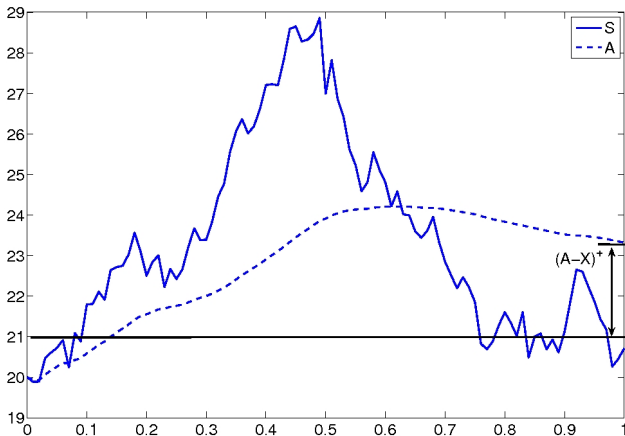
- plain-vanilla call  $V = (S - X)^+$
- floating strike Asian call  $V = (S - A)^+$
- floating rate Asian call  $V = (A - X)^+$
- lookback options  $V = (S - m)^+$

# How does it work?



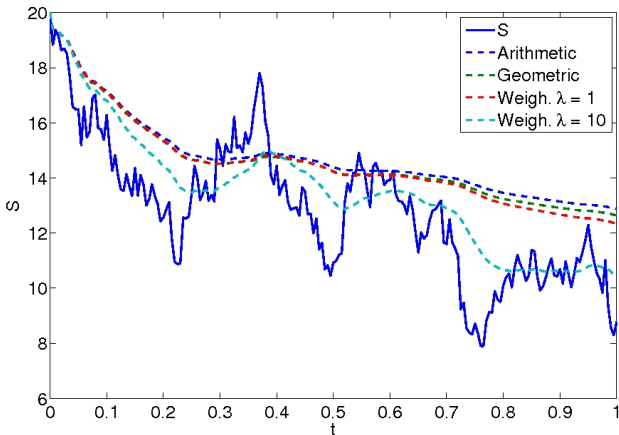
# How does it work?





A holder of the Asian option can be secured for the risk arising from a sudden underlying price jumps

## Asian options



Geometric average

Arithmetic average

Weighted arithmetic average

$$\ln A_T = \frac{1}{T} \int_0^T \ln S_\xi \, d\xi$$

$$A_T = \frac{1}{T} \int_0^T S_\xi \, d\xi$$

$$A_T = \frac{1}{\int_0^T a(\xi) \, d\xi} \int_0^T a(T - \xi) S_\xi \, d\xi$$

$$a(\xi) = \exp(-\lambda \xi)$$

## PDE for Asian options

$$\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} S^2 \frac{\partial^2 V}{\partial S^2} + (r - q) S \frac{\partial V}{\partial S} + Af\left(\frac{S}{A}, t\right) \frac{\partial V}{\partial A} - rV = 0,$$

$$V(S, A, t) = \max(S - A, 0),$$

where  $S, A > 0$ ,  $t \in (0, T)$  and

$$f(x, t) = \begin{cases} \frac{x - 1}{t}, & \text{for arithmetic avg.}, \\ \frac{\lambda(x - 1)}{1 - \exp(-\lambda t)}, & \text{for exp. weig. avg.}, \\ \frac{\ln(x)}{t}, & \text{for geometric avg.} \end{cases} \quad (1)$$

## Dimension reduction

- new state variable  $x = \frac{S}{A}$
- new function  $W(x, \tau) = \frac{1}{A} V(S, A, t)$ , where  $\tau = T - t$ .

$$0 = \frac{\partial W}{\partial \tau} + [f(x, T - \tau) - r + q] x \frac{\partial W}{\partial x} - \frac{\sigma^2}{2} x^2 \frac{\partial^2 W}{\partial x^2} + (r - f(x, T - \tau)) W,$$

$$W(x, 0) = \max(x - 1, 0),$$

where  $\tau \in (0, T)$ ,  $x > 0$ .

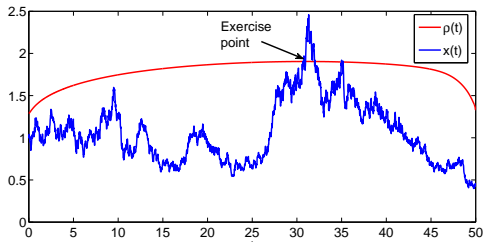
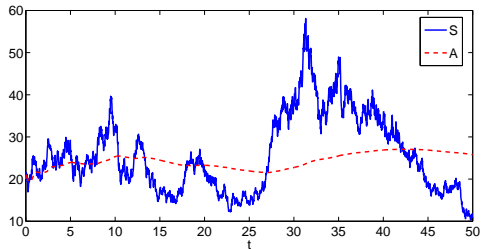
## American style - Asian options with early exercise possibility

### Free boundary problem

American style

- $0 < x < \rho(\tau), \tau \in (0, T),$
- $W(0, \tau) = 0,$
- $W(x, \tau) = x - 1,$  (terminal condition)
- $\frac{\partial W}{\partial x}(x, \tau) = 1,$  at  $x = \rho(\tau),$  (smooth pasting principle)
- **spatial domain depends on boundary function  $\rho.$**

# Early exercise boundary



## Fixed domain transformation

- $\xi = \ln \left( \frac{\rho(\tau)}{x} \right)$
- $x \in (0, \rho(\tau))$  iff  $\xi \in (0, \infty)$  for  $\tau \in (0, T)$
- $\Pi(\xi, \tau) = W(x, \tau) - x \frac{\partial W}{\partial x}(x, \tau)$  (financial meaning – synthetised portfolio)

$$0 = \frac{\partial \Pi}{\partial \tau} + \left[ \frac{\dot{\rho}}{\rho} - f(\rho e^{-\xi}, T - \tau) + r - q - \frac{\sigma^2}{2} \right] \frac{\partial \Pi}{\partial \xi} - \frac{\sigma^2}{2} \frac{\partial^2 \Pi}{\partial \xi^2} + \left[ r + x \frac{\partial f}{\partial x} - f(x, T - \tau) \right]_{x=\rho e^{-\xi}} \Pi.$$

$$\Pi(\xi, 0) = \begin{cases} -1, & \xi < \ln \rho(0), \\ 0, & \xi > \ln \rho(0). \end{cases}, \quad \Pi(0, \tau) = -1, \quad \Pi(\infty, \tau) = 0.$$

## Algebraic constraint between $\rho(\tau)$ and $\Pi(\xi, \tau)$

### Derivative form

$$q\rho(\tau) - r - \frac{\sigma^2}{2} \frac{\partial \Pi}{\partial \xi}(0, \tau) + f(\rho(\tau), T - \tau) = 0.$$

### Integral form

$$0 = \frac{d}{d\tau} \left( \ln \rho(\tau) + \int_0^\infty \Pi(\xi, \tau) d\xi \right) + q\rho(\tau) - q - \frac{\sigma^2}{2} \\ + \int_0^\infty (r - f(\rho(\tau)e^{-\xi}, T - \tau)) \Pi(\xi, \tau) d\xi.$$

## Algebraic constraint between $\rho(\tau)$ and $\Pi(\xi, \tau)$

### Derivative form

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### Integral form

$$\begin{aligned} 0 = & \frac{d}{d\tau} \left( \ln \rho(\tau) + \int_0^\infty \Pi(\xi, \tau) d\xi \right) + q\rho(\tau) - q - \frac{\sigma^2}{2} \\ & + \int_0^\infty (r - f(\rho(\tau) e^{-\xi}, T - \tau)) \Pi(\xi, \tau) d\xi. \end{aligned}$$

## Characterization

- discretization
- restrict spatial domain to a finite interval  $\xi \in (0, L)$
- $k = \frac{T}{m}$  - discretization time step
- $h = \frac{L}{n} > 0$  - spatial step
- $\tau_j = jk$
- $\Pi_i^j$  - full space-time approximation of  $\Pi(\xi_i, \tau_j)$
- $\rho^j = \rho(\tau_j)$

## Euler backward in time finite difference approximation

$$0 = \frac{\Pi^j - \Pi^{j-1}}{k} + c^j \frac{\partial \Pi^j}{\partial \xi} - \left( \frac{\sigma^2}{2} + f(\rho^j e^{-\xi}, T - \tau) \right) \frac{\partial \Pi^j}{\partial \xi} - \frac{\sigma^2}{2} \frac{\partial^2 \Pi}{\partial \xi^2} + \left[ r + x \frac{\partial f}{\partial x} - f(x, T - \tau) \Big|_{x=\rho^j e^{-\xi}} \right] \Pi^j,$$

where  $c(\tau) = \frac{\rho(\tau)}{\rho(\tau)} + r - q$ .

## Operator splitting method

- auxiliary intermediate step  $\Pi^{j-\frac{1}{2}}$
- Convection part

$$\frac{\Pi^{j-\frac{1}{2}} - \Pi^{j-1}}{k} + c^j \frac{\partial \Pi^{j-\frac{1}{2}}}{\partial \xi} = 0$$

- Diffusive part

$$0 = \frac{\Pi^j - \Pi^{j-\frac{1}{2}}}{k} - \left( \frac{\sigma^2}{2} + f(\rho^j e^{-\xi}, T - \tau) \right) \frac{\partial \Pi^j}{\partial \xi} - \frac{\sigma^2}{2} \frac{\partial^2 \Pi^j}{\partial \xi^2} + \left[ r + x \frac{\partial f}{\partial x} - f(x, T - \tau) \Big|_{x=\rho^j e^{-\xi}} \right] \Pi^j$$

## Transport equation

### Approximation of convective part

$$\frac{\partial \tilde{\Pi}}{\partial \tau} + c(\tau) \frac{\partial \tilde{\Pi}}{\partial \xi} = 0,$$

for  $\xi > 0$  and  $\tau \in (\tau_{j-1}, \tau_j)$

- initial condition  $\tilde{\Pi}(\xi, \tau_{j-1}) = \Pi^{j-1}(\xi)$
- boundary condition  $\tilde{\Pi}(0, \tau) = -1$ .

### Operator $\mathcal{T}$

$$\Pi_i^{j-\frac{1}{2}} = \begin{cases} \Pi^{j-1}(\nu_i), & \text{if } \nu_i = \xi_i + \ln \frac{\rho^{j-1}}{\rho^j} - (r - q)k > 0, \\ -1, & \text{otherwise.} \end{cases}$$

## A comparison of different interpolation methods in operator $\mathcal{T}$

m	interp. method	$\varepsilon_\rho$	$\varepsilon_{\Pi_1}$	$\varepsilon_{\Pi_2}$
200	linear	0.355520	0.008903	0.379569
	spline	0.355506	0.008234	0.378894
	cubic	0.167454	0.008234	0.378894
400	linear	0.151797	0.007341	0.190087
	spline	0.151780	0.007878	0.188898
	cubic	0.137178	0.007897	0.188898
800	linear	0.052020	0.006184	0.087705
	spline	0.052014	0.007696	0.083035
	cubic	0.049852	0.007702	0.083044
1600	linear	0.027937	0.004279	0.034823
	spline	0.027936	0.005383	0.027244
	cubic	0.010368	0.005394	0.027352

**linear** - Linear interpolation; **spline** - Cubic spline interpolation; **cubic** - Piecewise cubic Hermite interpolation.

$$\varepsilon_\rho = \|\rho - \rho_{\text{benchmark}}\|_\infty, \quad \varepsilon_{\Pi_1} = \|\Pi(\cdot, 25) - \Pi_{\text{benchmark}}(\cdot, 25)\|_\infty,$$

$$\varepsilon_{\Pi_2} = \|\Pi(\cdot, 49.5) - \Pi_{\text{benchmark}}(\cdot, 49.5)\|_\infty.$$

## Finite differences approximation of diffusive part

$$0 = \frac{\Pi_i^j - \Pi_i^{j-\frac{1}{2}}}{k} + \left[ r + x \frac{\partial f}{\partial x} - f(x, T - \tau) \right]_{x=\rho^j e^{-\xi_i}} \Pi_i^j - \left( \frac{\sigma^2}{2} + f(\rho^j e^{-\xi_i}, T - \tau) \right) \frac{\Pi_{i+1}^j - \Pi_{i-1}^j}{2h} - \frac{\sigma^2}{2} \frac{\Pi_{i+1}^j - 2\Pi_i^j + \Pi_{i-1}^j}{h^2}.$$

## Operator A

### Tridiagonal system of linear equations

$$\alpha_i^j \Pi_{i-1}^j + \beta_i^j \Pi_i^j + \gamma_i^j \Pi_{i+1}^j = \Pi_i^{j-\frac{1}{2}},$$

$$\alpha_i^j(\rho^j) = -\frac{k}{2h^2} \sigma^2 + \frac{k}{2h} \left( \frac{\sigma^2}{2} + f(\rho^j e^{-\xi_i}, T - \tau_j) \right),$$

$$\beta_i^j(\rho^j) = 1 + b(\xi_i, T - \tau_j)k - (\alpha_i^j + \gamma_i^j),$$

$$\gamma_i^j(\rho^j) = -\frac{k}{2h^2} \sigma^2 - \frac{k}{2h} \left( \frac{\sigma^2}{2} + f(\rho^j e^{-\xi_i}, T - \tau_j) \right),$$

## Algebraic constraint

$$\begin{aligned} \ln \rho^j &= \ln \rho^{j-1} + \int_0^\infty \Pi^{j-1}(\xi) d\xi - \int_0^\infty \Pi^j(\xi) d\xi \\ &+ k \left( q + \frac{\sigma^2}{2} - q\rho^{j-1} - \int_0^\infty (r - f(\rho^{j-1} e^{-\xi}, T - \tau_j)) \Pi^j d\xi \right). \end{aligned} \quad (2)$$

## Operator $\mathcal{F}$

In  $\mathcal{F}(\Pi^j)$  is right side of equation (2).

## Successive iterations procedure

- for  $j \geq 1$ :  $\Pi^{j,0} = \Pi^{j-1}$ ,  $\rho^{j,0} = \rho^{j-1}$ .
- $(\rho + 1)$ -th approximation of  $\Pi^j$  and  $\rho^j$ :

$$\rho^{j,\rho+1} = \mathcal{F}(\Pi^{j,\rho}), \quad (3)$$

$$\Pi^{j-\frac{1}{2},\rho+1} = \mathcal{T}(\rho^{j,\rho+1}), \quad (4)$$

$$\mathcal{A}(\rho_{\rho+1}^j) \Pi^{j,\rho+1} = \Pi^{j-\frac{1}{2},\rho+1}. \quad (5)$$

## Algorithm

## Algorithm

**Input variables:**  $q \geq 0, L, r, \sigma, n, m, T > 0, \lambda > 0, \rho_{max} = 500, toll = 10^{-8}$

**Initialization:**

$$k = T/m$$

$$h = L/n$$

$$\rho^0 = \begin{cases} \max\{\frac{1+rT}{1+qT}, 1\}, & \text{for arithmetic average,} \\ \max\{\bar{x}, 1\}, & \text{for geometric average,} \\ \max\{\frac{\lambda+r(1-e^{-\lambda T})}{\lambda+q(1-e^{-\lambda T})}, 1\}, & \text{for exponentially weighted average,} \end{cases}$$

$$\Pi^0 = \begin{cases} -1, & \xi < \ln \rho^0, \\ 0, & \xi > \ln \rho^0, \end{cases}$$

**for**  $j = 1$  **to**  $m$ :

$$\Pi_0^j = \Pi^{j-1}$$

$$\rho_0^j = \rho^{j-1}$$

**for**  $p = 0$  **to**  $\rho_{max}$ :

$$\rho_{p+1}^j = \mathcal{F}(\Pi_p^j)$$

$$\Pi_{p+1}^{j-\frac{1}{2}} = \mathcal{T}(\rho_{p+1}^j)$$

$$\mathcal{A}(\rho_{p+1}^j)\Pi_{p+1}^j = \Pi_{p+1}^{j-\frac{1}{2}}$$

**if**  $(|\rho_{p+1}^j - \rho_p^j| < toll)$

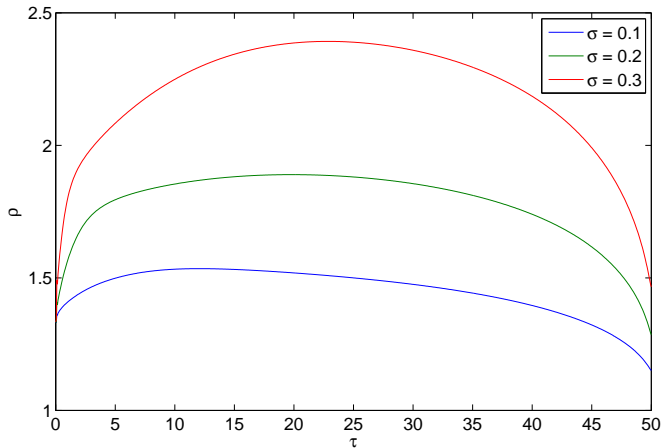
**break**

**endif**

**end**

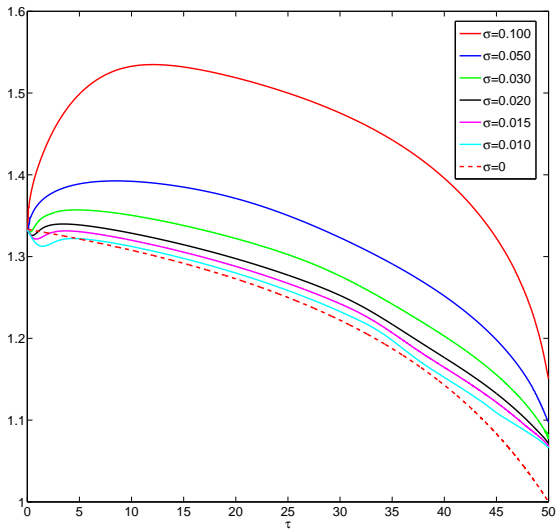
**end,**

# A comparison of the free boundary position for various $\sigma$

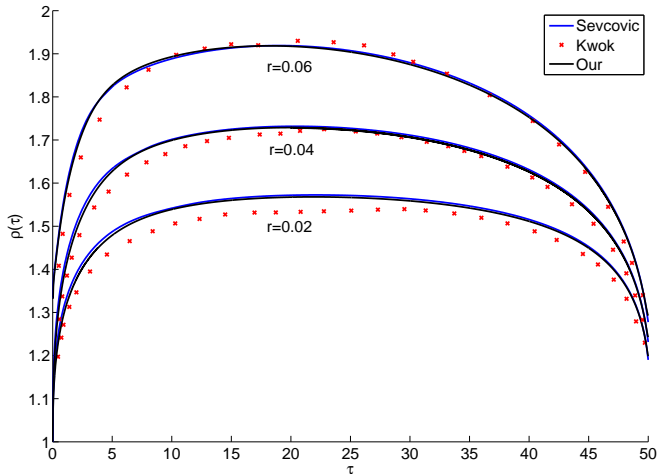


Arithmetic averaged options

# A comparison of the free boundary position for various small $\sigma$

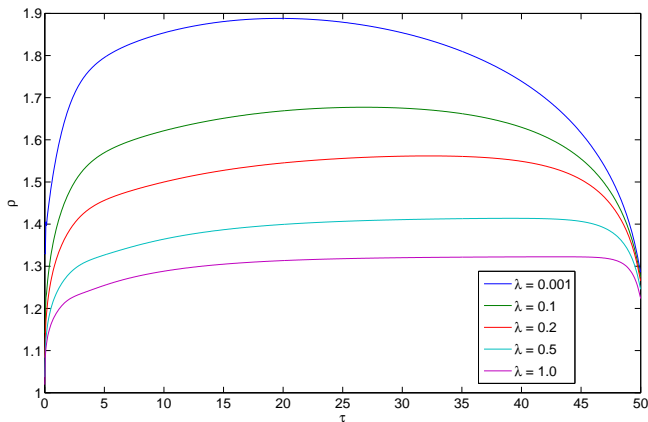


# A comparison position of the free boundary position for various $r$



Exponentially weighted averaged options

# A comparison of the free boundary position for exponentially weighted Asian options

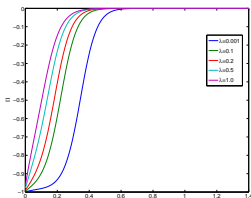
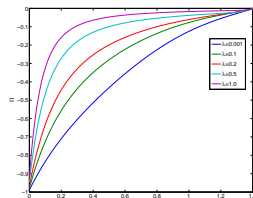
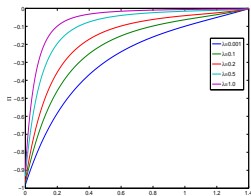
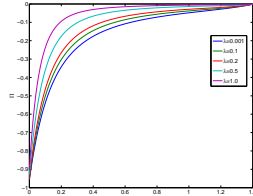


Exponentially weighted averaged options

**Experimental order of convergence for  $\rho_\lambda$** 

$\lambda$	$\ \rho_\lambda - \rho_\infty\ _\infty$	$\alpha_\infty$
0.001	0.888104	—
0.1	0.677320	0.058835
0.2	0.561828	0.269710
0.5	0.413783	0.333796
1.0	0.320136	0.370188
2.0	0.247010	0.374115
3.0	0.212705	0.368769
4.0	0.191862	0.358490
5.0	0.177658	0.344686
10.0	0.147227	0.271064
20.0	0.113350	0.377261
30.0	0.094150	0.457702

## Exponentially weighted averaged options

 $t = 0.25$  $t = 5$  $t = 25$  $t = 45$ 

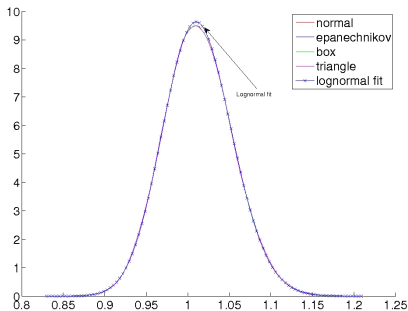
**Thank you**

Thanks for your attention!

# Arithmetic rate Asian options

## Payoff diagram

$$V(S, A, T) = (A - X)^+$$



## Moments of $A_T$

### First moment

$$\mathbb{E}[A_T] = S_0 \frac{\lambda}{\lambda + r} \frac{e^{rT} - e^{-\lambda T}}{1 - e^{-\lambda T}}$$

### Second moment

$$\mathbb{E}[A_T^2] = e^{-2\lambda T} \frac{S_0^2}{k^2} \frac{2}{\tilde{\alpha}} \left[ \frac{\exp(\tilde{\beta}) - \exp(\tilde{\alpha})}{\tilde{\beta} - \tilde{\alpha}} - \frac{\exp(\tilde{\beta}) - 1}{\tilde{\beta}} \right]$$

$$\tilde{\alpha} = (r + \lambda)T, \tilde{\beta} = 2(r + \frac{1}{2}\sigma^2 + \lambda)T, k = \frac{1}{T} \int_0^T \exp(-\lambda\xi) d\xi$$

## Approximate formula

$$\begin{aligned}
 V(S, 0) &= e^{-rT} \mathbb{E}_Q[(A_T - X)^+] \\
 &= e^{-rT} \int_X^\infty (x - X) \frac{1}{x\chi\sqrt{2\pi}} \exp\left[-\frac{(\ln(x) - \varphi)^2}{2\chi^2}\right] dx
 \end{aligned}$$

where

$$\begin{aligned}
 \varphi &= 2 \ln(E[A_T]) - \frac{1}{2} \ln E[A_T^2], \\
 \chi^2 &= \ln \frac{E[A_T^2]}{(E[A_T])^2}.
 \end{aligned}$$

$\sigma$	X	r	RS-PDE	T-LB	T-UB	AA	LN	MC
0.05	95	0.05	7.157	7.1777	7.1779	7.1794	7.17802	7.14323
0.05	100	0.05	2.621	2.7162	2.7162	2.7279	2.72574	2.70776
0.05	105	0.05	0.439	0.3372	0.3374	0.3257	0.34352	0.33706
0.05	95	0.09	8.823	8.8088	8.8089	8.8091	8.80888	8.87211
0.05	100	0.09	4.185	4.3082	4.3084	4.3173	4.31292	4.31100
0.05	105	0.09	1.011	0.9583	0.9585	0.9561	0.96888	0.94235
0.05	95	0.15	11.090	11.0941	11.0943	11.0941	11.09409	11.13954
0.05	100	0.15	6.777	6.7944	6.7946	6.7963	6.79500	6.78619
0.05	105	0.15	2.639	2.7444	2.7446	2.7559	2.75309	2.73807
0.10	90	0.05	11.942	11.9511	11.9523	11.9666	11.95337	11.82228
0.10	100	0.05	3.624	3.6413	3.6416	3.6725	3.64798	3.64981
0.10	110	0.05	0.359	0.3311	0.3322	0.2855	0.32426	0.34403
0.10	90	0.09	13.382	13.3852	13.3862	13.3935	13.38630	13.43389
0.10	100	0.09	4.887	4.9151	4.9154	4.9597	4.92349	4.93918
0.10	110	0.09	0.659	0.6301	0.6310	0.5840	0.62376	0.65125
0.10	90	0.15	15.398	15.3988	15.3995	15.4015	15.39906	15.48592
0.10	100	0.15	7.000	7.0277	7.0286	7.0707	7.03506	7.03506
0.10	110	0.15	1.430	1.4133	1.4143	1.3901	1.41130	1.35585
0.20	90	0.05	12.589	12.5956	12.6008	12.7837	12.62990	12.51913
0.20	100	0.05	5.760	5.7627	5.7645	5.8330	5.78310	5.78487
0.20	110	0.05	1.996	1.9892	1.9927	1.8322	1.97131	2.00315
0.20	90	0.09	13.825	13.8312	13.8373	14.0072	13.86178	13.83059
0.20	100	0.09	6.773	6.7770	6.7787	6.8915	6.80379	6.83882
0.20	110	0.09	2.551	2.5455	2.5486	2.4269	2.53478	2.57984
0.20	90	0.15	15.636	15.6416	15.6491	15.7898	15.66540	15.69377
0.20	100	0.15	8.402	8.4085	8.4105	8.5691	8.44099	8.40859
0.20	110	0.15	3.558	3.5547	3.5578	3.5098	3.55665	3.55116

A comparison of different methods  $S_0 = 100$ ,  $T = 1$ . RS-PDE are values obtained by solve PDE (Roger a Shi)

T-LB and T-UB are lower and upper bounds Thompson (2000), AA is analytical approximations Zhang.

