Pricing exotic path-dependent options The Singular Points method [GZA10; GZ11]

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Motivation

Asian options

Cliquet options

Conclusion

(Financial) Market models: introduction

- Financial assets
 - basic assets
 riskless (deterministic) treasury bonds: S_t^O = e^{rt}
 risky (stochastic) stocks: (S_t)_t, parameter σ characterises risk
 - derivatives contracts on other assets (underlying), till maturity T futures symmetric risk options asymmetric risk – historically used for insurance
- Problem: pricing derivatives
- Assumptions
 - ▶ viable / no arbitrage / no free lunch
 - frictionless
 - infinitely divisible assets

Option types

 Simple options - classification of the basis of: exercise time European or American right of owner call or put

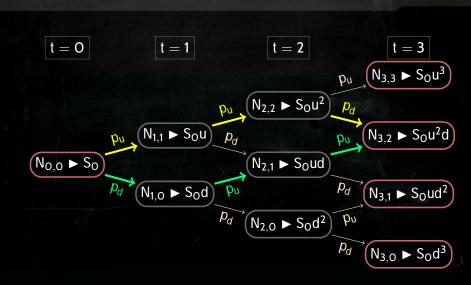
Example (European call)

payoff: $h(S_T) = (S_T - K)_+ = max\{S_T - K, 0\}$, exercise at maturity

Exotic options – usually path-dependent

Asian payoff: function of average of the underlying. lookback payoff: function of extrema of the underlying. cliquet a series of globally or locally bounded at-the-money options.

digital existence depends on pre-set barriers.



2015-201 [BS73] model (continuous)

riskless
$$S_t^0 = e^{rt}$$

$$\text{risky } S_t = s_0 e^{(r - \frac{\sigma^2}{2})t + \sigma W_t}$$

Theorem (Convergence of prices from CRR to BS) Prices of basic assets under CRR $\stackrel{d}{\rightarrow}$ prices of basic assets under BS.

Corollary (Convergence of evaluation formulae)
The previous theorem implies that evaluation formulae under CRR converge in distribution to evaluation formulae for BS.

Approximate BS price by using CRR model.

Quest: Find algorithms with reduced computational complexity.

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Market models: discrete vs. continuous

Parameter	Discrete	Continuous
Example	[CRR79]	[BS73]
Theoretical complexity	Easy	Hard
Ease of implementation	Hard	Easy
Closed-form formula	No ¹	Yes
Computational complexity	Hard: $O(2^n)^1$	Easy: O(1)
Versatile	Yes	No

¹CRR: backward recursive



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Asian options: introduction

Payoff: function of some form of average price.

Average	Discrete	Continuous
AM	$A_n = \frac{1}{n+1} \sum_{i=0}^n S_n$	$A_T = \frac{1}{T} \int_0^T S_t dt$
GM	$G_n = \left(\prod_{i=0}^n S_n\right)^{\frac{1}{n+1}}$	$G_T = exp\left(\frac{1}{T}\int_0^T log(S_t) dt\right)$

Example (fixed-strike Asian call of European type) Given strike-price K, payoff $= (A_T - K)_+$, exercised only at maturity.

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Asian option: pre-existing methods

<u>Arithmetic</u>	mean	STEEL STATE OF	
Method	Type	Complex	Remarks
[CRR79]	Tree	O(2 ⁿ)	simple, accurate, convergence
[HW93]	Tree	$O(n^3)$	accuracy & convergence problems
[BP96]	Tree	$O(n^3)$	accuracy & convergence problems
[Cha+99]	Tree	$O(n^4)$	thin bounds, very large memory
[VecO1]	PDE	$O(n^2)$	not universally applicable
[dFLO5]	PDE		more general than [VecO1]

Geometric mean Closed-form formula exist under BS.

2015-2027 SPM for Asian options: idea

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N<sub>i,j</sub> Node of the binomial tree
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 $A_{i,j}^{l} \:\: \text{Average upto} \: N_{i,j}, I \in \binom{i}{j}$

P_{i,i} Corresponding option price

$$\left\{\left(A_{i,j}^{I},P_{i,j}^{I}\right)\right\}_{I}$$
 singular points (SPs) – completely characterise price price function continuous, convex, piecewise-linear; found by joining SPs.

SPM for Asian options: start at maturity (i = n)

For $N_{n,j}$: calculate $A_{i,j}^{min}$ and $A_{i,j}^{max}$.

Singular points:

1.
$$j \in \{0, n\}: (A, (A - K)_+)$$

$$2. \ j \notin \{0,n\} \ \text{and} \ K \in \left(A_{n,j}^{min},A_{n,j}^{max}\right):$$

2.1
$$(A_{n,i}^{min}, O)$$

$$2.3~(A_{n,j}^{max},A_{n,j}^{max}-K)$$

3.
$$j \notin \{0, n\}$$
 and $K \notin (A_{n,j}^{min}, A_{n,j}^{max})$:

3.1
$$\left(A_{n,j}^{\min}, (A_{n,j}^{\min} - K)_{+}\right)$$

3.2
$$\left(A_{n,j}^{max}, (A_{n,j}^{max} - K)_{+}\right)$$

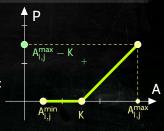


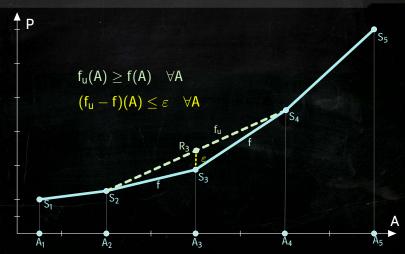
Figure: Case 2

SPM for Asian options: exact price

- Up movement (for N_{i,i}): t = i1. $\forall A_{i+1,j}^l : B^l = \frac{(i+2)A_{i+1,j}^l - S_{i+1,j}}{i+1}$. Step 2 $\textbf{2.} \ \ \textbf{If} \ \textbf{B}^{\textbf{I}} \in \left[\textbf{A}^{min}_{\textbf{i},\textbf{j}}, \textbf{A}^{max}_{\textbf{i},\textbf{j}}\right] \implies \textbf{B}^{\textbf{I}} \ \textbf{is} \ \textbf{SA}.$ 3. For each SA, $B_{ij}^{I} = \frac{(i+1)B^{I} + S_{i+1;j+1}}{i+2}$. 4. $v_{i,i}(B^{l}) =$ $\tfrac{1}{R} \left\lceil p_u v_{i+1,j+1} \left(B_u^l \right) + p_d v_{i+1,j} \left(A_{i+1,j}^l \right) \right\rceil.$ Step 1 5. $(B^I, V_{i,j}(B^I))$ is a SP. down Down movement (for Ni i). Aggregate and sort by SAs.
- Repeat for all j. Iterate backward till i = 0. $P_{0,0}^{1}$ is
- the exact binomial price.

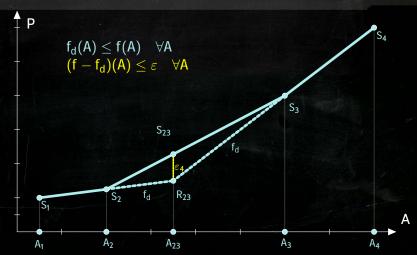
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SPM for Asian options: approximation Upper estimates



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SPM for Asian options: approximation Lower estimates



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SPM for Asian options: numerical results

Data: $s_0 = 100, T = 1, r = 0.1, q = 0.03$.

	K		90	K = 110	
	n	$\sigma = 0.2$	$\sigma = 0.4$	$\sigma = 0.2$	$\sigma = 0.4$
D.	10	14.5912	17.8033	2.5100	6.6523
Bin	25	15.1535	18.6786	2.6270	7.3451
SP	10	14.5925	17.8068	2.5090	6.6511
	25	15.1535	18.6785	2.6270	7.3449
	50	15.3524	19.0420	2.6673	7.4563
	100	15.4732	19.2696	2.6886	7.5174
	200	15.5453	19.4065	2.6996	7.5502
	400	15.5861	19.4845	2.7053	7.5674

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SPM for Asian options: summary

- □ Introduced by Gaudenzi et al [GZA10] in 2010.
- Convergent to exact CRR and thus BS.
- ☑ Easily generalised to American case and lookback options.
- Approx: A priori error bounds.
- ☑ Difficult to compute theoretical complexity.
- Depends on the recombinant nature of the underlying's tree.
- Not extensible to GM, since the price function is non-linear.

$$G_u = \left(G^{i+1}S_{i+1,j}\right)^{\frac{1}{i+2}} \propto G^{\frac{i+1}{i+2}}$$

oxdots Constant volatility assumption \implies local volatility models fail.



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Definitions

forward start option $\;$ price option today with payoff $=(S_T-S_u)_+,$ $0 \leq u < T.$

cliquet option a series of consecutive at-the-money forward start options, with bounded returns.

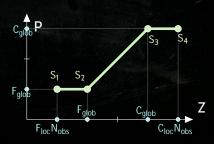
Pre-existing methods for pricing

- No prominent tree-based method.
- ► [WilO2]: PDE based, FD approach.
- [WFVO6]: PDE based, FD approach; generalisations.

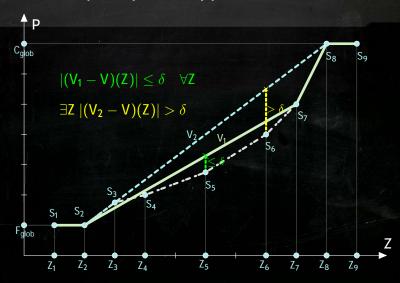
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Cliquet options: terminology

- N: observation times (equidistant).
- ► Return: $R_i = \frac{S_i S_{i-1}}{S_{i-1}} = \frac{S_i}{S_{i-1}} 1$.
- ► Running sum: $Z_i = \sum_{k=1}^{i} max\{F_{loc}, min\{C_{loc}, R_k\}\}.$
- ▶ Payoff = $\max\{F_{glob}, \min\{C_{glob}, Z_N\}\}$.
- m: computational time steps.
- ▶ 2^m possible paths; \sim Bin(p).
- Z depends on paths, probs.
- $ightharpoonup (Z,P) \Longrightarrow SP (P \text{ is price}).$
- Price function at maturity.
- Proceed as in the Asian case.
- Iterate backwards. P¹_O is the exact binomial price.



^{¬o's"}SPM for cliquet options: approximation



SPM for cliquet options: numerical results

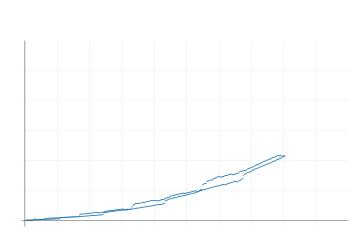
Data

$$ightharpoonup$$
 $F_{loc} = O, C_{loc} = O.08, F_{glob} = O.16, C_{glob} = \infty$

$$T = 5, N = 5, r = 0.03$$

σ m	m	Price		Time (s) 2	
		Bin	SP	Bin	SP
	200	0.173716366	0.173716366	0.0165	0.00828
0.2	500	0.173922597	0.173922671	0.0875	0.0437
	1000	0.174051949	0.174051983	2.38	0.183
	200	0.150465004	0.150466828	600	6.09
0.0	2 500	0.150508871	0.150510526	∞	24.2
	1000	0.150522368	0.150524027	∞	55

 $^{^{2}\}infty$ means time taken is more than an hour.



SPM for cliquet options: summary

- Approximation: A priori error bounds.
- Approximation: converges to binomial price, no bounds.
- Significant speed improvement in low volatility cases against binomial model.
- ☑ Can be used for local volatility models and varying interest rates in each period.
- Difficult to compute theoretical complexity.



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Recapitulation

- Efficient polynomial-time technique.
- Theory and flexibility varies with option type.

Further research

- ► Theoretical complexity: dependence of singular point redundancy on initial data.
- Verify cliquet complexity for large m.
- Customising the method for other path-dependent options.

Questions?

Thank you!

²⁰¹⁵⁻¹⁰⁻¹²Bibliography I

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