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# Pricing of path-dependent cancelables



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Consider a **path dependent cancelable contract** which generates (possibly negative) cash-flows

$$C_1, \dots, C_\tau$$

up to a cancellation date  $\tau$ . The cash-flows of this contract are equivalent to an aggregated cash-flow at cancellation date,

$$B_*(\mathcal{I}_\tau) \mathcal{Z}_\tau := B_*(\mathcal{I}_\tau) \sum_{j=1}^{\tau} Z_j,$$

with  $Z_i := C_i / B_*(\mathcal{I}_i)$  being discounted cash-flows with respect to the numeraire  $B_*$ . Product price at time zero:

$$V_0^{cancel} := \sup_{\tau \in \{1, \dots, k\}} E^{\mathcal{F}_0} \mathcal{Z}_\tau = \sup_{\tau \in \{1, \dots, k\}} E^{\mathcal{F}_0} \sum_{j=1}^{\tau} Z_j,$$

where the supremum is taken over all stopping indices with values in the set  $\{1, \dots, k\}$ .

### Path-dependent callables

A path dependent **callable** contract generates

$$C_{\tau+1}, \dots, C_k$$

when called at  $\tau$ . It is equivalent to the sum of a non-callable and a cancelable one (and vice versa):

$$\begin{aligned} V_0^{call} &:= \sup_{\tau \in \{1, \dots, k\}} E^0 \sum_{j=\tau+1}^k Z_j \\ &= E^0 \sum_{j=1}^k Z_j + \sup_{\tau \in \{1, \dots, k\}} E^0 \sum_{j=1}^{\tau} (-Z_j) \end{aligned}$$

### Example: The cancelable snowball swap

Pays semi-annually a constant rate  $I$  over the first year and in the forthcoming years  $(\text{Previous coupon} + A - \text{Libor})^+$ , semi-annually, where  $A$  is given in the contract. For this case we take

$$\begin{aligned}K_i &:= I, \quad i = 0, 1, \\K_i &:= (K_{i-1} + A_i - L_i(T_i))^+, \quad i = 2, \dots, n - 1,\end{aligned}$$

with  $A_2 := 0.03$ ,  $A_{i+1} = A_i$  for even  $i$ ,  $A_{i+1} = A_i + 0.005$  for odd  $i$ . Cancellation is allowed for  $2 \leq \tau < n$ ,  $n = 20$  (10 years)

Effective discounted cashflows at  $T_j$ :

$$Z_j := \frac{(L_{j-1}(T_{j-1}) - K_{j-1})^+ \delta_{j-1}}{B_*(T_j)},$$

hence aggregated up to cancellation  $Z_\tau = \sum_{j=1}^{\tau} Z_j$ .

### Iterating the snowball swap

Take an input policy satisfying

$$\begin{aligned} i &\leq \tau_i \leq k, \quad \tau_k = k, \\ \tau_i > i &\Rightarrow \tau_i = \tau_{i+1}, \quad 0 \leq i < k, \end{aligned}$$

construct the new policy

$$\begin{aligned} \hat{\tau}_i &:= \inf\{j \geq i : Z_j \geq \max_{p: j \leq p \leq k} E^{\mathcal{F}_j} Z_{\tau_p}\} \\ &= \inf\{j \geq i : 0 \geq \max_{p: j \leq p \leq k} E^{\mathcal{F}_j} \sum_{q=j+1}^{\tau_p} Z_q\} \end{aligned}$$

and compute the iterated price

$$\hat{Y}_0 := E^{\mathcal{F}_0} Z_{\hat{\tau}_0},$$

which is generally an improvement of  $Y_0$  due to policy  $\tau$ .

### Numerical results for typical market data

#### Improved Andersen

$d$	$Y(0; \tau_A)$ (SD)	$\hat{Y}(0; \tau_A)$ (SD)	$Y_{up}(0; \tau_A)$ (SD)
1	127.77(0.238)	129.77(0.318)	130.33(0.247)
2	114.93(0.231)	120.00(0.389)	121.92(0.293)
19	76.725(0.217)	91.600(0.460)	98.107(0.476)

150 000 outer and 500 inner paths for  $\hat{Y}$  and 20 000 outer (with 500 inner) paths for  $Y_{up}$ .

#### Improved least-squares regression method (Piterbarg)

$d$	$Y(0; \tau_{LS})$ (SD)	$\hat{Y}(0; \tau_{LS})$ (SD)	$Y_{up}(0; \tau_{LS})$ (SD)
1	117.73(0.243)	128.81(0.632)	132.28(0.313)
2	103.70(0.238)	120.73(0.466)	123.54(0.346)
19	74.913(0.224)	93.515(0.469)	97.479(0.379)

200 000 outer and 500 inner paths for  $\hat{Y}$  and 20 000 outer (with 500 inner) paths for  $Y_{up}$ .

### Improving an Andersen-like optimization of the LS exercise boundary

$d$	$Y(0; \tau_{LS-A})$ (SD)	$\widehat{Y}(0; \tau_{LS-A})$ (SD)	$Y_{up}(0; \tau_{LS-A})$ (SD)
1	129.58(0.237)	128.70(0.349)	130.24(0.244)
2	119.58(0.230)	118.95(0.345)	120.77(0.244)
10	92.201(0.219)	97.376(0.456)	100.20(0.418)
19	87.787(0.217)	94.487(0.445)	95.843(0.430)

150 000 outer and 100 inner paths for  $\widehat{Y}$  and 5000 outer (with 500 inner) paths for  $Y_{up}$ .

$$\tau_{LS-A, i} = \inf\{j \geq i : Z_j \geq H_j + Y_{LS, j}\}$$

with optimized constants  $H_j$ .

### Message:

- (i) Price the callable using Pitterbarg's version of Longstaff-Schwartz;
- (ii) Improve the obtained exercise boundary with an Andersen-like optimization;
- (iii) Compute the Dual upperbound due to the stopping time  $\tau_{LS-A, 0}$ ;
- (iv) If there is still a significant gap between lower and upper bound, then improve the policy  $\tau_{LS-A, i}$  by the iteration method.

Bender, Kolodko, Schoenmakers (2006), revised version of WIAS prepr. 1061