
Greeks of Bermuda Swaptions:
Financial Interpretation and Fast Computations

Quantitative Finance 2002

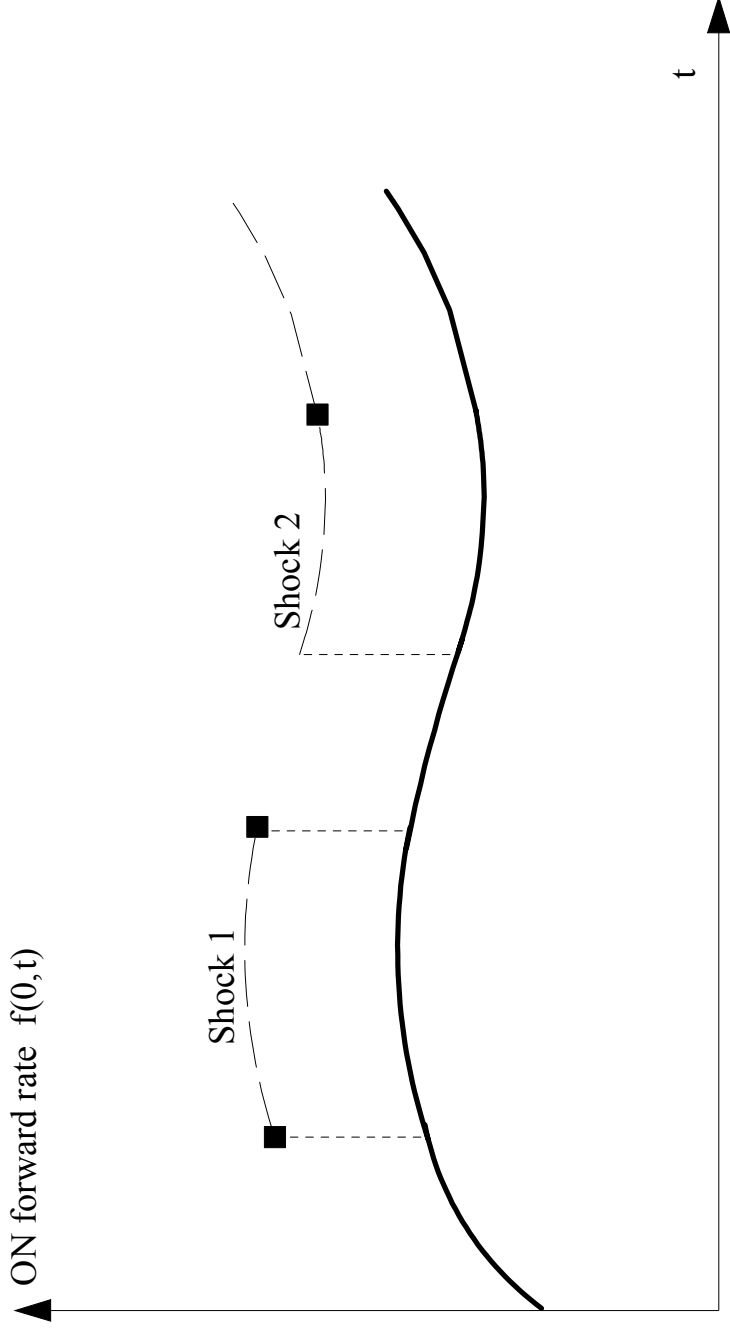
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1 Goal

- Bermuda swaptions are American-style options on interest rate fixed-rate for floating-rate swaps.
- The most liquid of all interest rate exotics.
- Typically valued by numerically solving a PDE.
- Risk sensitivities – deltas, gammas, vegas – typically computed by numerical differentiation: “shock inputs and revalue the Bermudan”.
- Problems:
 - Slow: One PDE solution per Greek.
 - Inaccurate: Numerical noise is magnified by numerical differentiation.
- Is anything better possible?

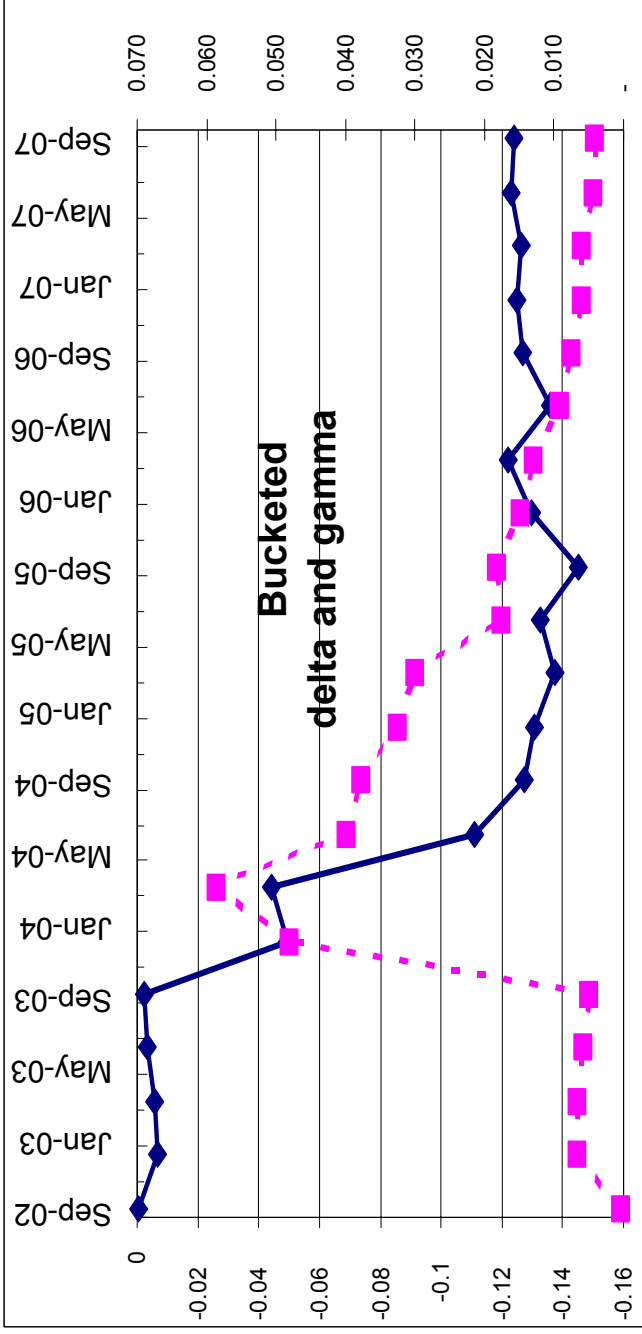
2 Deltas to shocks of initial interest rate curve

- Many equivalent parametrizations of the interest rate curve are possible.
- For the discussion on deltas, we parametrize it by instantaneous forward rates.
- Possible shocks:

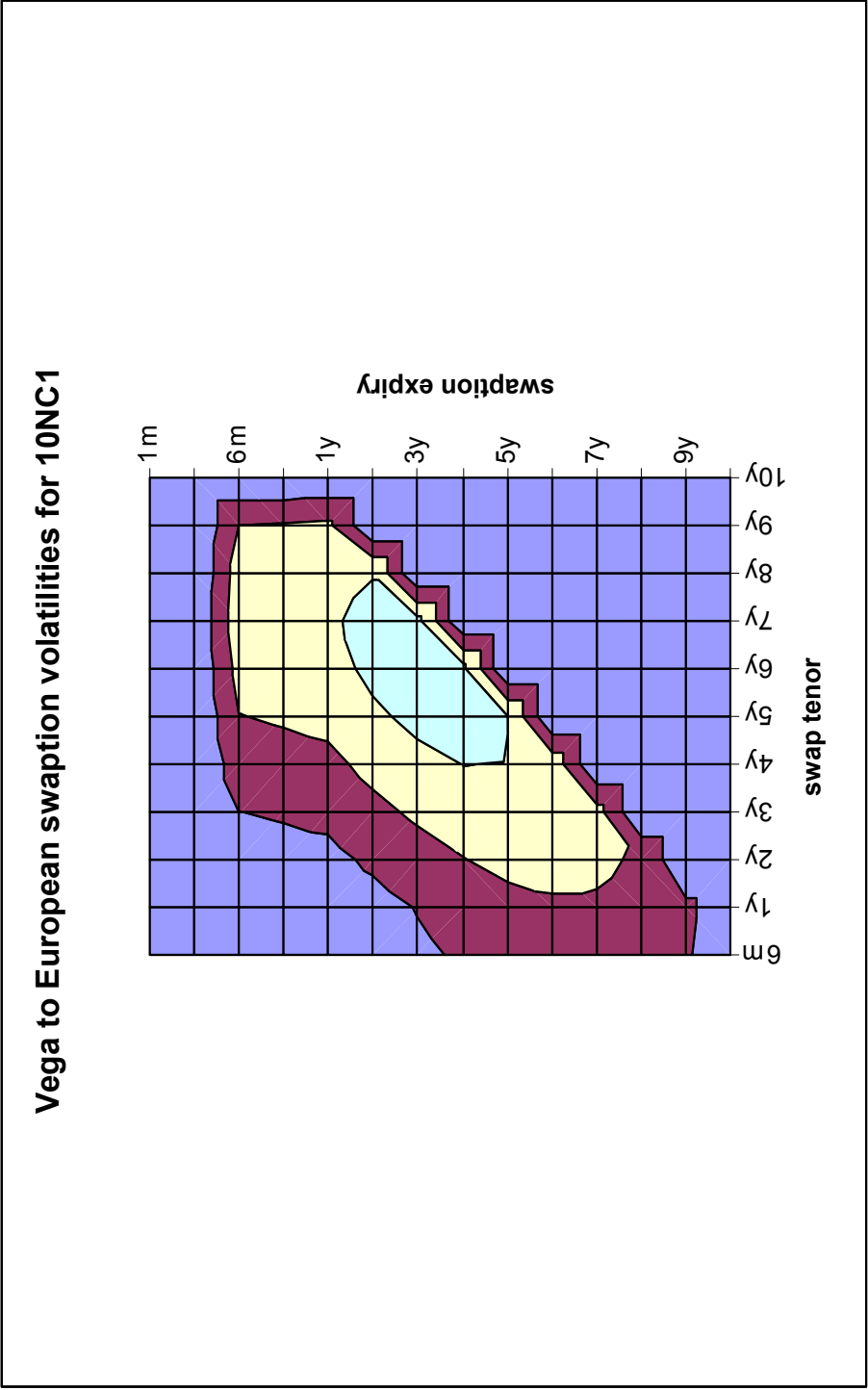


3 Curve sensitivities of Bermuda swaptions

- Bucketed deltas (solid line) – sensitivities of a Bermuda swaption to shocks $f(0, t) + \varepsilon 1_{\{T_i < t \leq T_{i+1}\}}$ for a collection of dates T_i .
- Bucketed gammas (dotted line) – sensitivities of bucketed deltas to a parallel shift of the whole curve $f(0, \cdot) + \varepsilon$.



4 Volatility sensitivities of Bermuda swaptions



5 Are we stuck with bump-and-revalue method for deltas?

- Motivational example 1: Black's formula for options on forwards

$$FN(d_1) - KN(d_2).$$

Delta is $N(d_1)$, computed along with the value.

- Motivational example 2: Black's formula on a PDE grid (x_j, t_i) . If instrument value is $V(x=0, t=0)$ then its delta $(2 \cdot \Delta x)^{-1} [V(\Delta x, 0) - V(-\Delta x, 0)]$, computed at no extra computational cost.
- Motivational example 3: Monte-Carlo, integration by parts:

$$\begin{aligned} \frac{\partial}{\partial \theta} \mathbf{E}^\theta(V(\xi)) &= \frac{\partial}{\partial \theta} \int V(x) p_\xi(x; \theta) dx \\ &= \int V(x) \left(\frac{\partial}{\partial \theta} \log p_\xi(x; \theta) \right) p_\xi(x; \theta) dx \\ &= \mathbf{E}(V(\xi) w(\xi)). \end{aligned}$$

- Clearly in many cases, information needed for computing deltas is available during valuation.

6 Notations

- Instantaneous forward rates at time t for forward period $[T, T + 0]$ denoted by $f(t, T) = f(\omega, t, T)$.
- Zero coupon discount bonds:

$$P(t, T) = \exp\left(-\int_t^T f(t, s) ds\right).$$

- Discount (cc) rates

$$F(t, T) = -(T - t)^{-1} \log P(t, T) = (T - t)^{-1} \int_t^T f(t, s) ds.$$

- Money market account

$$B_T = \exp\left(\int_0^T f(t, t) ds\right) = \exp\left(\int_0^T r(t) ds\right).$$

7 Instruments

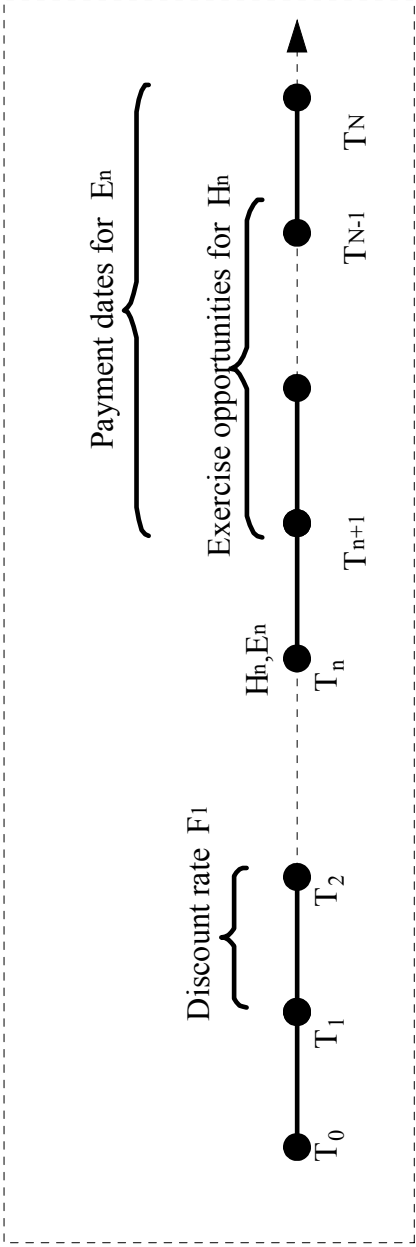
- Tenor structure: $0 = T_0 < T_1 < \dots < T_N$, $\tau_n = T_n - T_{n-1}$.
- The n -th swap starts at T_n . Fixed rate c , value at time t ,

$$E_n(t) = P(t, T_n) - P(t, T_N) - c \sum_{k=n+1}^N P(t, T_k) \tau_k.$$

- European swaption with exercise date T_k : European option with payoff $\max\{E_k(T_k), 0\}$.
- Bermuda swaption H_0 : Can exercise on any dates T_1, \dots, T_{N-1} ; if exercised on date T_k receive $E_k(T_k)$.
- Bermuda swaption H_n : Can exercise on dates T_{n+1}, \dots, T_{N-1} (“sub-Bermudan”).

8 Backward recursion for valuation

- Timeline



- Main recursion: $n = N - 2, \dots, 0$,
- $$H_n(T_n) = e^{-\tau_{n+1} F_n(x)} \mathbf{E}_{T_n}^{T_{n+1}} \max \{ H_{n+1}(T_{n+1}), E_{n+1}(T_{n+1}) \},$$
- $$H_{N-1} \equiv 0.$$
- $\mathbf{E}_{T_n}(\cdot)$'s are usually computed on a PDE grid.

9 What inputs to shock?

- Deltas = sensitivity to interest rate curve shocks.
- What shocks to use? A shock to the interest rate curve affects many “moving parts”.
- It is common to use a different “basis” for computing deltas.
- Look at slide 2. It is sometimes better to use shocks of “Shock 2” type:

$$f(0, t) + \varepsilon 1_{\{t > T_i\}}$$

and compute deltas to those shocks and convert back to deltas for shocks of type “Shock 1”:

$$f(0, t) + \varepsilon 1_{\{T_i < t \leq T_{i+1}\}},$$

than compute the deltas to “Shock 2” type shocks directly.

10 Bright idea

- Let us shock $E_n(T_n)$'s directly and individually! They enter the formula for Bermudans explicitly, deltas must be the easiest to compute!
- What does it mean to shock $E_n(T_n)$ directly? Think Hull-White
- Short rate state $x(t)$, everything in terms of $x(t)$:

$$\begin{aligned}dx(t) &= (\theta(t) - ax(t)) dt + \sigma(t) dW(t), \\r(t) &= f(0, t) + x(t), \\P(x(t), t, T) &= P(0, t, T) \exp(-b(t, T)x(t) + A(t, T)).\end{aligned}$$

- All exercise values are deterministic functions of $x(\cdot)$,

$$E_n(T_n) = E_n(x(T_n)).$$

- We will describe our method for HW, but it is applicable in a much broader setting.

11 Bermuda swaption valuation as a functional 1

- Define a set $C_0(\mathbb{R})$ of continuous functions that “do not grow too fast”. Think of them as functions of the HW state variable $x(\cdot)$ and require

$$|f(x)| < Ke^{a|x|^{2-\varepsilon}}, \varepsilon > 0.$$

- Think of the recursion $n = N - 1, \dots, 1$,

$$H_{n-1}(x) = e^{-\tau_n F_{n-1}(x)} \mathbf{E}^{T_n}(\max\{H_n(x(T_n)), E_n(x(T_n))\} | x(T_{n-1}) = x),$$

$$H_{N-1} \equiv 0.$$

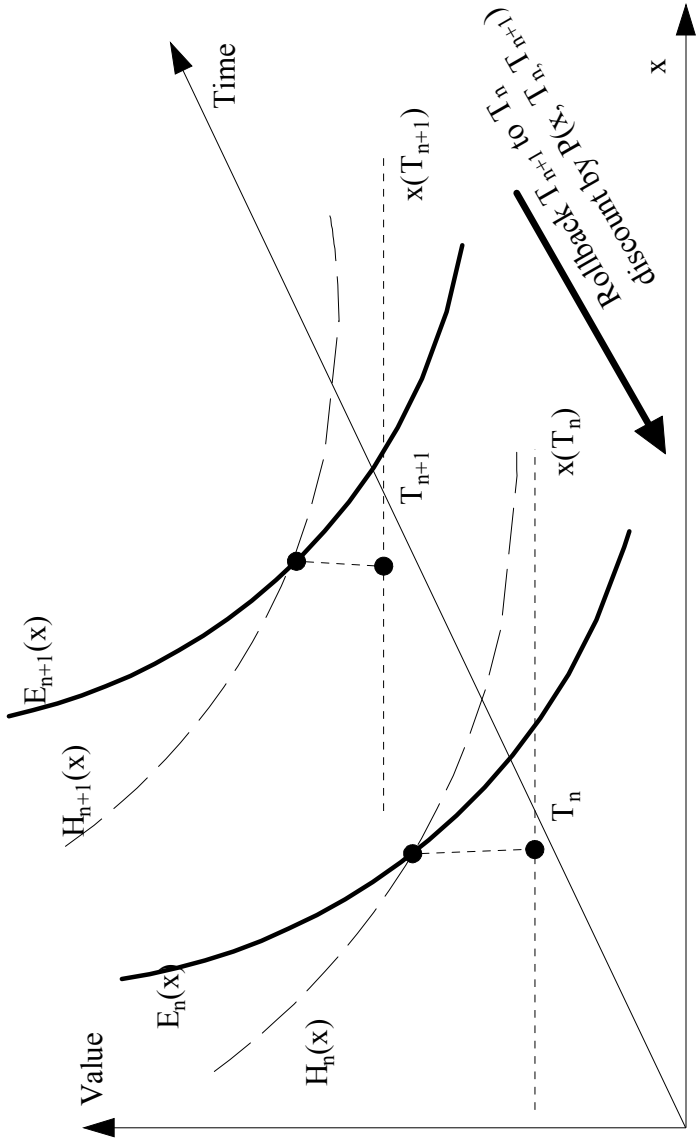
as a definition of functions $\{H_n(x)\}_{n=0}^{N-1}$ from arbitrary functions $E_n(x(T_n))$, $F_{n-1}(x) \in C_0(\mathbb{R})$. Treat \mathbf{E}^{T_n} as an operator on $C_0(\mathbb{R})$:

$$f(x) \mapsto \mathbf{E}^{T_n}(f(x(T_n)) | x(T_{n-1}) = x).$$

- For any k we regard the Bermuda value H_k as a functional on $(C_0(\mathbb{R}))^{2(N-1)}$,

$$H_k : (C_0(\mathbb{R}))^{2(N-1)} \longrightarrow C_0(\mathbb{R}).$$

12 Bermuda swaption valuation as a functional 2



- Apply shocks to functions $E_n(x)$ and $F_{n-1}(x)$ directly!
- Compute deltas as sensitivities (as $\varepsilon \rightarrow 0$) to individual shocks

$$E_n(x) + \varepsilon D_n^e(x), \quad F_{n-1}(x) + \varepsilon D_{n-1}^f(x).$$

13 Model deltas

- Define two sets of model deltas, “underlying” deltas and “discount” deltas. Regard H_k as a functional $(C_0(\mathbb{R}))^{2(N-1)} \rightarrow C_0(\mathbb{R})$, differentiate individual inputs in “directions” $D_n^e(x)$, $D_{n-1}^f(x)$,

$$\begin{aligned}\Delta_n^e H_k &= \left. \frac{\partial}{\partial \varepsilon} H_k(\dots, E_n(x) + \varepsilon D_n^e(x), \dots) \right|_{\varepsilon=0}, \\ \Delta_{n-1}^f H_k &= \left. \frac{\partial}{\partial \varepsilon} H_k(\dots, F_{n-1}(x) + \varepsilon D_{n-1}^f(x), \dots) \right|_{\varepsilon=0}.\end{aligned}$$

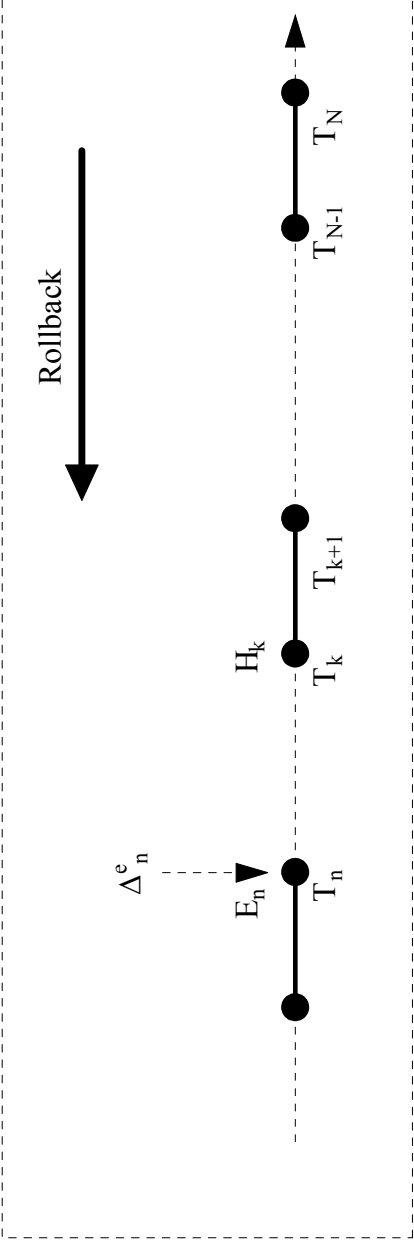
14 Recursion for model deltas 1

- Fix n . Let us try to compute $\Delta_n^e H_k$.
- Recall

$$H_k(x) = \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (\max\{H_{k+1}, E_{k+1}\} | x(T_k) = x).$$

- If $n < k + 1$, then the computation of H_k will not be affected by a shock to $E_n(x)$ at all. So

$$\Delta_n^e H_k = 0.$$



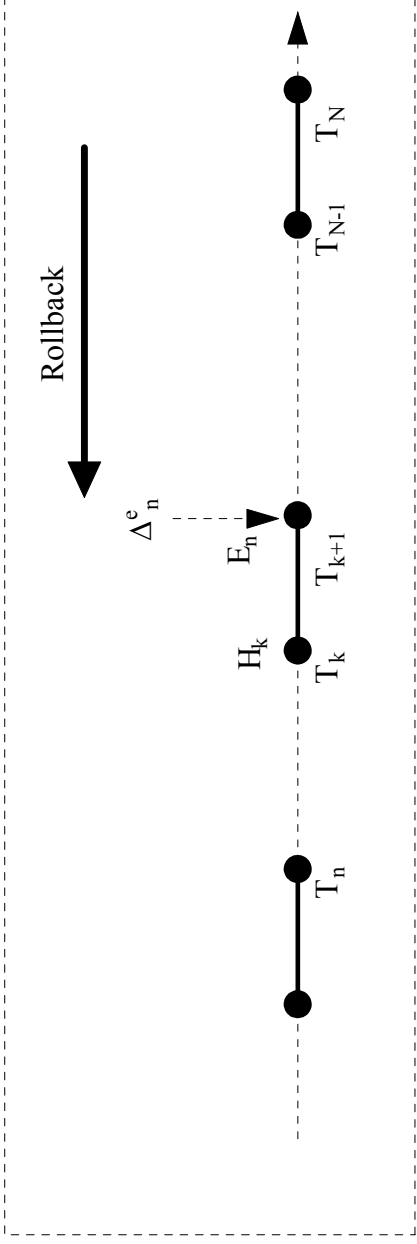
15 Recursion for model deltas 2

- Recall

$$H_k(x) = \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (\max\{H_{k+1}, E_{k+1}\} | x(T_k) = x).$$

- If $n = k + 1$, then H_{k+1} is unaffected by a shock to $E_n(x)$ but E_{k+1} obviously is affected. Formally differentiating the recursion above we get $(\max(x, k))' = 1_{\{x \geq k\}}$,

$$\begin{aligned} \Delta_n^e H_k &= \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (1_{\{E_{k+1} \geq H_{k+1}\}} \Delta_n^e E_{k+1} | x(T_k) = x) \\ &= \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (1_{\{E_{k+1} \geq H_{k+1}\}} D_{k+1}^e | x(T_k) = x). \end{aligned}$$



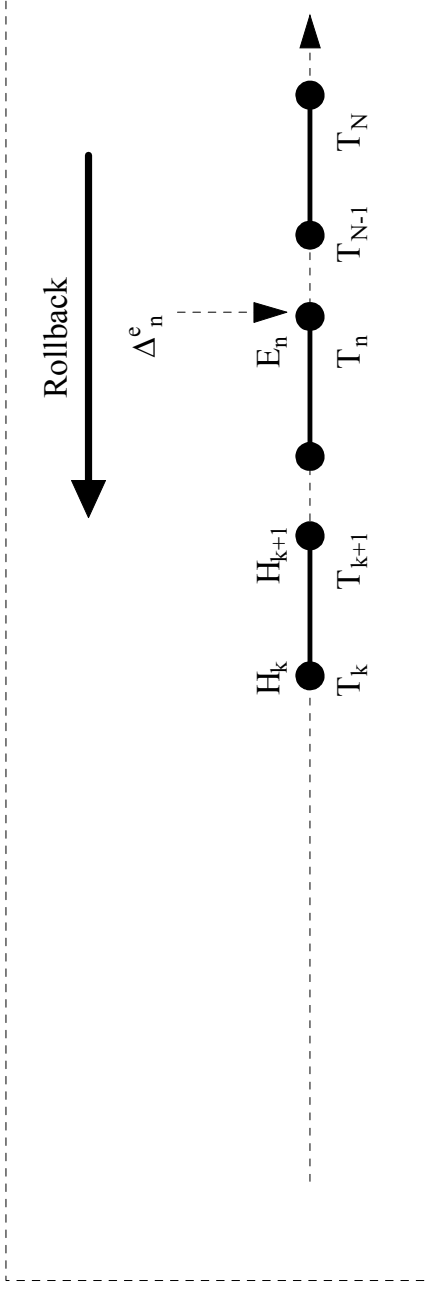
16 Recursion for model deltas 3

- Recall

$$H_k(x) = \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (\max\{H_{k+1}, E_{k+1}\} | x(T_k) = x).$$

- If $n > k + 1$, then H_{k+1} is affected by a shock to $E_n(x)$ via recursive formulas but E_{k+1} is not (independent bumps to E_i). Thus

$$\begin{aligned} \Delta_n^e H_k &= \exp(-\tau_{k+1} F_k(x)) \mathbf{E}_{T_k}^{T_{k+1}} (1_{\{E_{k+1} < H_{k+1}\}} \Delta_n^e H_{k+1} | x(T_k) = x), \\ \Delta_n^e H_k &= B_{T_k} \mathbf{E}_{T_k} (B_{T_{k+1}}^{-1} 1_{\{E_{k+1} < H_{k+1}\}} \Delta_n^e H_{k+1} | x(T_k) = x) \end{aligned}$$



17 Lemma on differentiation

- Exchanging differentiation and expectation is justified by the following lemma

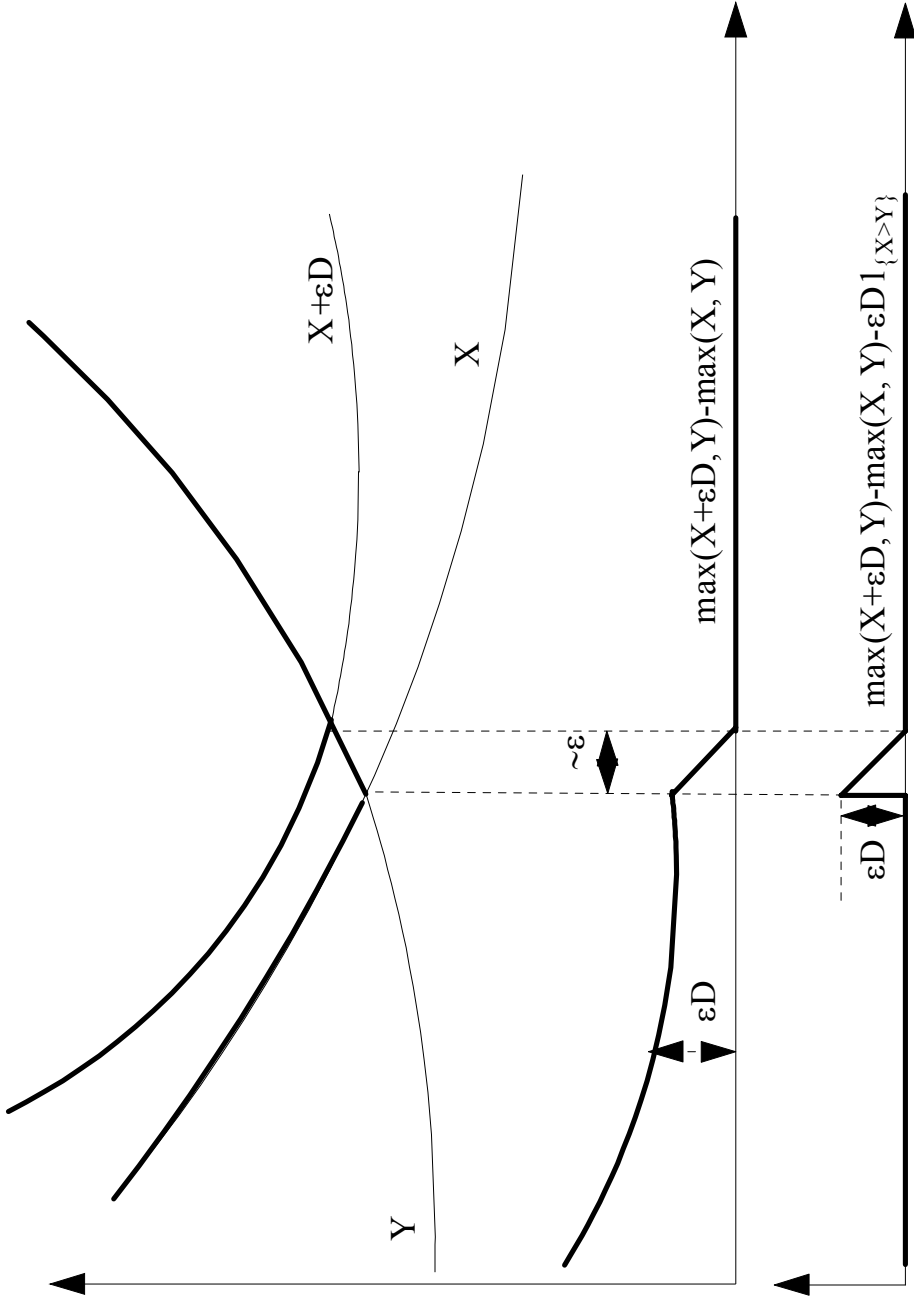
Lemma 1 : *Let X , Y and D be random variables such that*

$$\mathbf{E}|X| < \infty, \quad \mathbf{E}|Y| < \infty, \quad \mathbf{E}|D| < \infty.$$

If $\mathbf{P}(X = Y) = 0$ then

$$\lim_{\varepsilon \rightarrow 0} \varepsilon^{-1} \mathbf{E}(\max(X + \varepsilon D, Y) - \max(X, Y)) = \mathbf{E}(1_{\{X > Y\}} D) .$$

18 Proof of the lemma



19 Unwrapping the recursion

- Our goal – deltas of H_0 .
- Use recursive relations to “push deltas through”.

- We have

$$\begin{aligned}
\Delta_n^e H_0 &= \mathbf{E}_0 \left(B_{T_1}^{-1} \cdot 1_{\{E_1 < H_1\}} \cdot \Delta_n^e H_1 \right) \\
&= \mathbf{E}_0 \left(B_{T_1}^{-1} \cdot 1_{\{E_1 < H_1\}} \cdot B_{T_1} \mathbf{E}_{T_1} B_{T_2}^{-1} \left(1_{\{E_2 < H_2\}} \cdot \Delta_n^e H_2 \right) \right) \\
&= \dots \\
&= \mathbf{E}_0 \left(B_{T_{n-1}}^{-1} \cdot \prod_{i=1}^{n-1} 1_{\{E_i < H_i\}} \cdot \Delta_n^e H_{n-1} \right) \\
&= \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot \prod_{i=1}^{n-1} 1_{\{E_i < H_i\}} \cdot 1_{\{E_n \geq H_n\}} \cdot D_n^e \right).
\end{aligned}$$

- This is our first main result.

20 Discount deltas

- Similar recursions and formulas hold for “discount” deltas $\Delta_n^f H_k$, model deltas of H_k with respect to shocks to discount rates $F_n(x)$.
- For the Bermudan H_0 we have

$$\Delta_n^f H_0 = -\tau_{n+1} \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot \prod_{i=1}^{n-1} 1_{\{H_i > E_i\}} \cdot 1_{\{H_n > E_n\}} \times D_n^f \times H_n \right).$$

21 Optimal exercise time

- Our formula for deltas allows for a number of interesting interpretations of deltas.
- Define “**exercise**” regions for each exercise opportunity T_n ,

$$R_n = \{x \in \mathbb{R} : H_n(x) \leq E_n(x)\}, \quad 0 \leq n \leq N-1$$

and “**hold**” regions

$$R_n^c = \{x \in \mathbb{R} : H_n(x) > E_n(x)\}, \quad 0 \leq n \leq N-1.$$

- Define optimal exercise time (index)

$$\eta = \min \{n \geq 1 : x(T_n) \in R_n\}.$$

- Bermuda value via optimal exercise time:

$$H_0 = \sum_{m=1}^{N-1} \mathbf{E}_0 \left(B_{T_m}^{-1} \cdot 1_{\{\eta=m\}} \cdot E_m(T_m) \right) = \mathbf{E}_0 \left(B_{T_{\eta \wedge N}}^{-1} \cdot E_{\eta \wedge N}(T_{\eta \wedge N}) \right).$$

22 Model deltas via optimal exercise time

- Recall

$$\Delta_n^e H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot \prod_{i=1}^{n-1} 1_{\{E_i < H_i\}} \cdot 1_{\{E_n \geq H_n\}} \cdot D_n^e \right).$$

- Note that

$$1_{\{\eta=n\}} = \prod_{i=1}^{n-1} 1_{\{E_i < H_i\}} \cdot 1_{\{E_n \geq H_n\}}.$$

- Thus, model deltas can also be expressed via optimal exercise time:

$$\Delta_n^e H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta=n\}} \cdot D_n^e \right).$$

23 Model deltas as values of knock-outs

- Define T_n -knock-out instrument as a contingent claim that disappears (“knocks out”) when $x(T_i)$ enters R_i for any $i = 1, \dots, n - 1$.
- A value of a T_n -knock-out with payoff ξ at time T_n is equal to

$$\mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta > n-1\}} \cdot \xi \right)$$

- Recall from previous slide

$$\Delta_n^e H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta = n\}} \cdot D_n^e \right) = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta > n-1\}} \cdot 1_{\{E_n \geq H_n\}} \cdot D_n^e \right).$$

- Thus $\Delta_n^e H_0$ is the price of a T_n -knock-out instrument with time- T_n payoff

$$1_{\{E_n \geq H_n\}} \cdot D_n^e.$$

- Carr in [Car01] established the interpretation of European option deltas as prices of contingent claims; we establish a similar interpretation for Bermudan swaptions, except the instruments are knock-outs.

24 The survival measure

- Value of a European option with payoff ξ at time T = integral of ξ with respect to state density at time T .
- Value of a knock-out option with payoff ξ at time T = integral of ξ with respect to the state “survival” density at time T .
- Define $\Psi(Y, t)$ for $Y \subset \mathbb{R}$ by

$$\Psi(Y, t) = \mathbf{E}_0 \left(B_t^{-1} \cdot 1_{\{t \leq T_\eta\}} \cdot 1_{\{x(t) \in Y\}} \right).$$

“Discounted probability of $x(t) \in Y$ given that the state process $x(\cdot)$ did not knockout before t ”. Density $\psi(y, t)$ defined by

$$\Psi(Y, t) = \int_Y \psi(y, t) \, dy.$$

25 Model deltas via the survival density

- The formula for the n -th “underlying” delta reads

$$\Delta_n^e H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{T_n \leq T_\eta\}} \cdot 1_{\{E_n \geq H_n\}} \cdot D_n^e \right).$$

- The formula for the survival density reads

$$\psi(y; T_n) = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{T_n \leq T_\eta\}} \cdot \delta_y \{x(T_n)\} \right).$$

- Using the formula of full probability we obtain from the two formulas that

$$\Delta_n^e H_0 = \int 1_{\{R_n\}}(y) \cdot D_n^e(y) \cdot \psi(y; T_n) \, dy.$$

26 Equation for the occupational density

- Consider

$$\Phi_{s,x}(Y, t) \triangleq B_s^{-1} \mathbf{E}_{s,x} (B_t^{-1} 1_{\{x(t) \in Y\}})$$

(the value of the contingent claim that pays $1_{\{x(t) \in Y\}}$ at time t , evaluated at time s conditioned on $x(s) = x$). We denote by $\phi_{s,x}(y; t)$ its density

$$\Phi_{s,x}(Y, t) = \int_Y \phi_{s,x}(y, t) dy.$$

- The density ϕ satisfies the forward Kolmogorov equation (Λ is the infinitesimal generator for $x(\cdot)$, Λ_y means it is applied to y -argument functions),

$$\frac{\partial}{\partial t} \phi_{s,x}(y, t) = (\Lambda_y^* \phi_{s,x})(y, t) - r(t) \phi_{s,x}(y, t).$$

27 Equation for the survival density

- For

$$T_n < t \leq T_{n+1},$$

by conditioning on $x(T_n)$, we can obtain

$$\psi(y, t) = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta > n\}} \cdot \phi_{T_n, x(T_n)}(y, t) \right).$$

- Therefore, since both the expectation operator and Λ_y^* are linear,

$$\frac{\partial}{\partial t} \psi(y, t) = (\Lambda_y^* \psi)(y, t) - r(t) \psi(y, t). \quad (1)$$

- What happens when t “crosses over” T_{n+1} ? Have special conditions (continuity)

$$\psi(y, T_{n+1} + 0) = \psi(y, T_{n+1}) \times 1_{\{y \in R_{n+1}^c\}}(y). \quad (2)$$

- Algorithm: Start with $\psi(y, 0) = \delta_0(y)$. Roll forward using the PDE (1). For $t = T_1$, apply the condition (2). Then roll forward using (1), and so on.

28 Fast computation of model deltas

- Remember formula for deltas

$$\Delta_n^e H_0 = \int 1_{\{R_n\}}(y) \cdot D_n^e(y) \cdot \psi(y; T_n) dy, \quad n = 1, \dots, N-1.$$

- Remember formulas for the survival density

$$\begin{aligned} \frac{\partial}{\partial t} \psi(y, t) &= (\Lambda_y^* \psi)(y, t) - r(t) \psi(y, t), \quad T_n < t \leq T_{n+1}, \\ \psi(y, T_{n+1} + 0) &= \psi(y, T_{n+1}) \times 1_{\{y \in R_{n+1}^c\}}(y). \end{aligned}$$

- Conclusion: At the expense of 1 PDE solution and $N-1$ integrations, we can get all “underlying” deltas $\Delta_n^e H_0$! (another $N-1$ integrations needed for “discount” deltas).
- Contrast that to the usual “shock-and-revalue” method that requires $N-1$ PDE solutions.
- See [And96] for forward-equation based algorithm for computing European option deltas.

29 Market deltas 1

- Really not that interested in the “model” deltas, i.e. deltas to shocks in $E_n(T_n)$, $F_n(T_n)$. Want deltas to shocks to the initial term curve.

- Let

$$f(0, t) \longmapsto f(0, t) + \varepsilon \theta(t),$$

be a shock to the initial curve. Denote by ∂_θ a derivative, of anything, with respect to that shock.

- This shock affects three things: exercise values at future times $E_n(T_n)$; discount rates at future times $F_n(T_n)$; and expectation operator \mathbf{E}_{T_n} (NOT for Hull-White, but in general)
- Apply chain rule! Assume \mathbf{E}_{T_n} is independent of the shock (can deal with general case, but it is a bit messier).

30 Market deltas 2

- Use

$$\begin{aligned} D_n^e(x) &= \partial_\theta E_n(x), \\ D_n^f(x) &= \partial_\theta F_n(x), \end{aligned}$$

as shocks to $E_n(x)$, $F_n(x)$ (“model” shocks). Can usually get them in closed form (definitely the case for HW model)

- Then

$$\begin{aligned} \partial_\theta H_0 &= \sum_{n=1}^{N-1} \int 1_{\{R_n\}}(x) \cdot \partial_\theta E_n(x) \cdot \psi(x; T_n) dx \\ &\quad - \sum_{n=0}^{N-2} \tau_{n+1} \int 1_{\{R_n^c\}}(x) \cdot \partial_\theta F_n(x) \cdot \psi(x; T_n) dx. \end{aligned}$$

- Market delta for any shock of initial term curve is a sum of integrals of known payoffs against the survival density.
- Multiple shocks θ – can use the same survival density ψ .

31 Do we need to compute all integrals?

- Can speed things up more by not computing all integrals. How? For example choose shocks θ_i such that $(\partial_i = \partial_{\theta_i})$

$$\partial_i E_n(t=0) = \delta_{\{n=i\}}$$

(can always “rotate” to any other basis later).

- With such shocks it is likely that for $i \neq n$,

$$\partial_i E_n(x, T_n) \ll \partial_n E_n(x, T_n).$$

Can reduce the number of integrals by ignoring smaller contributions.
Can implement an adaptive scheme that computes integrals to a certain accuracy.

- Bottom line: once we have the representation for deltas as above, we can do a lot of model-specific optimization.

32 Vegas 1

- Analysis for vegas (sensitivity to volatilities) is more model specific, but can be performed along similar lines.
- Same general idea: identify volatility parameters internal to the model, bump them to compute model vegas, derive recursions for them, convert back to “market” vegas.
- Usually we can identify $N - 1$ volatility parameters v_n , $n = 0, \dots, N - 2$, such that a shock to v_n does not affect $\mathbf{E}_{T_k} f(x(T_{k+1}))$ for $k \neq n$.
- For HW:

$$\begin{aligned}\sigma(t) &= \sum 1_{t \in [T_n, T_{n+1}]} \times \sigma_n, \\ v_n &= \sigma_n.\end{aligned}$$

33 Vegas 2

- If ∇_n is the derivative with respect to a bump to v_n then (assume $n \neq 0$) as before, can differentiate through the recursive relation,

$$\begin{aligned} H_0 &= \mathbf{E}_0 B_{T_1}^{-1} \max(H_1, E_1), \\ \nabla_n H_0 &= \mathbf{E}_0 (B_{T_1}^{-1} \cdot 1_{\{H_1 > E_1\}} \cdot \nabla_n H_1) + \mathbf{E}_0 (B_{T_1}^{-1} \cdot 1_{\{H_1 \leq E_1\}} \cdot \nabla_n E_1) \\ &\approx \mathbf{E}_0 (B_{T_1}^{-1} \cdot 1_{\{H_1 > E_1\}} \cdot \nabla_n H_1). \end{aligned}$$

- Can iterate until get to H_n ,

$$\nabla_n H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot \prod_{i=1}^n 1_{\{H_i > E_i\}} \cdot \nabla_n H_n \right).$$

- The same as for deltas, $\nabla_n H_0$ can be obtained by integrating $\nabla_n H_n(x)$ against the survival density $\psi(T_n, x)$ over the set $\{H_n(x) > E_n(x)\}$.

- The quantity $\nabla_n H_n$ has to be computed numerically: bump v_n and roll-back the payoff

$$\max (H_{n+1}(x), E_{n+1}(x))$$

from T_{n+1} to T_n using the bumped volatility.

- Still saving a lot because the rollback only over one period $[T_n, T_{n+1}]$, not over $[0, T_{n+1}]$ as needed in the standard method.

35 Volatility correction for deltas

- In the section on deltas we assumed that the model's volatility does not depend on interest rates. Can relax it.
- If volatility depends on rates, then full delta = delta assuming the volatility is not affected by a rates shock + vega from the change in volatility.
- Need the same type of condition as for vegas: "locality".
- For example can assume that the model's volatility is a deterministic function of $v_n(P(\cdot, \cdot))$, $n = 0, \dots, N - 2$, where each v_n is a function of rates, but shocks to v_n do not affect $\mathbf{E}_{T_k} f(x(T_{k+1}))$ for $k \neq n$. Then the same approach as for vegas can be used.

36 Applications to Monte-Carlo 1

- Valuation (lower bound) of Bermuda swaptions by Monte-Carlo is based on estimating the exercise and hold regions R_η and R_η^c in some optimization procedure and then valuing Bermuda swaption as a barrier swap using the formula

$$H_0 = \mathbf{E}_0 \left(B_{T_{\eta \wedge N}}^{-1} \cdot E_{\eta \wedge N} (T_{\eta \wedge N}) \right),$$

where $\eta = \eta(\omega)$ is now the index of the first time when a Monte-Carlo path ω enters the exercise region R_η .

- Recall that model deltas can be expressed in terms of η as well!

$$\Delta_n^e H_0 = \mathbf{E}_0 \left(B_{T_n}^{-1} \cdot 1_{\{\eta=n\}} \cdot D_n^e \right).$$

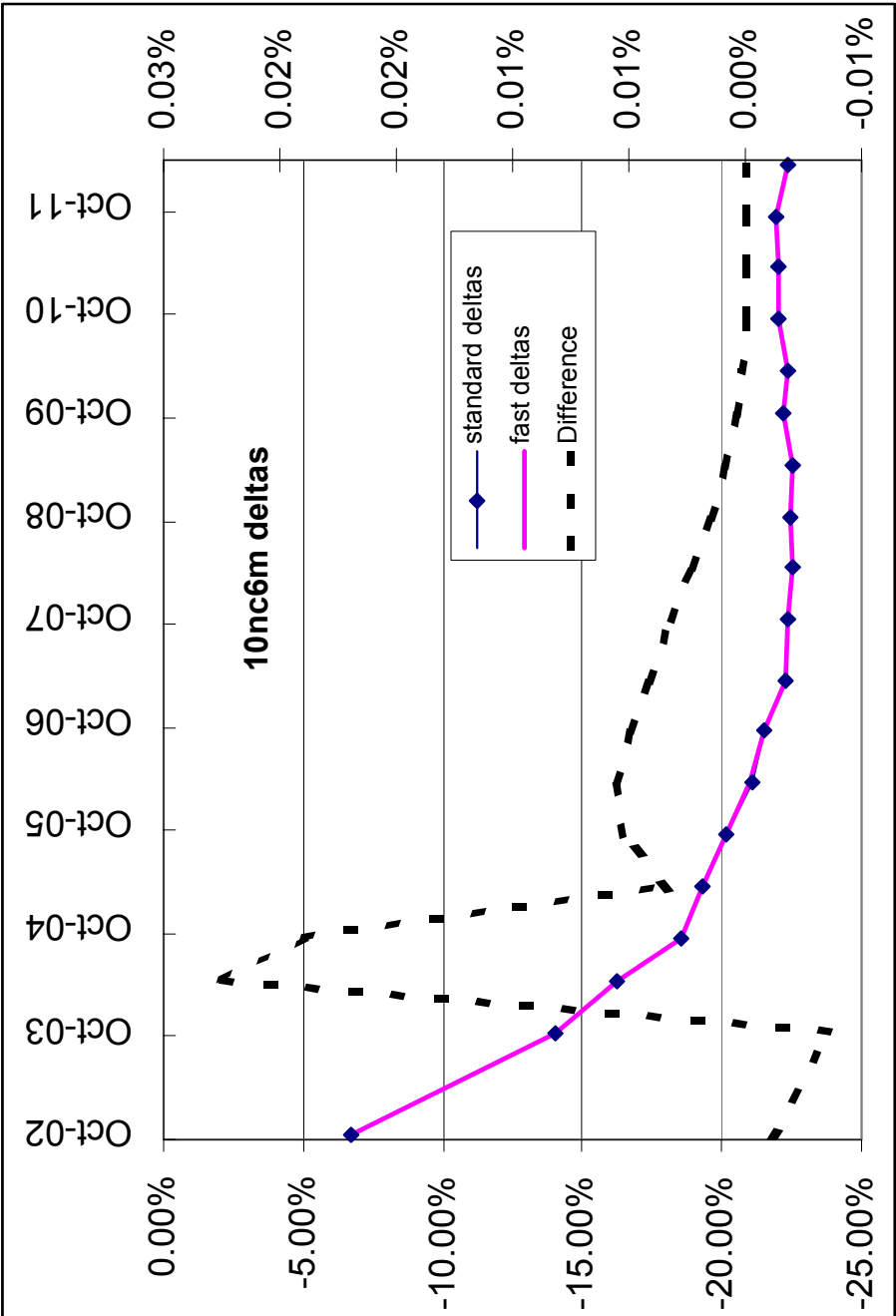
- Thus the deltas can be computed in the same simulation if D_n^e (and D_n^f) are known.

37 Applications to Monte-Carlo 2

- The future exercise values $E_n(T_n)$ and future discount rates $F_n(T_n)$ are “smooth” functions of the initial interest rate curve.
- For any shock θ to the initial curve, $\partial_\theta E_n(T_n)$, $\partial_\theta F_n(T_n)$ can be computed by e.g. “pathwise differentiation” for each path in the same simulation.
- See [GZ99] for details on that. Our results for Bermuda deltas can be combined with theirs on computing $\partial_\theta E_n$, $\partial_\theta F_n$ in simulation to yield a viable delta computation scheme for Monte-Carlo.

38 Sample results

Deltas for a 10y Bermudan 4.75% receiver with 1y no-call period as of October 1, 2002. Deltas on the left scale, difference on the right.



39 Conclusions 1

We developed an algorithm for fast computation of bucketed Greeks of Bermuda swaptions. The algorithm works like this:

- Shock inputs that are “natural” to each Bermuda swaption; define “model” deltas as sensitivities to those shocks;
- Derive recursive relations for “model” deltas;
- Use recursive relations to derive representations of “model” deltas as integrals with respect to the survival density;
- Derive a forward PDE for the survival density;
- Express “market” deltas in terms of “model” deltas using the chain rule;
- Compute “market” deltas as sums of integrals;
- Compute vegas and volatility adjustment to deltas the same way.

Much faster algorithm + many possibilities for further model-specific optimization + more accurate!

In addition to deriving a fast algorithm for Greeks computation, we developed

- Financial interpretations of deltas as prices of knockout contingent claims;
- Expressed deltas in terms of the optimal exercise time;
- Demonstrated how to use this representation to compute Bermuda swap-tion deltas in Monte-Carlo simulation.

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