Stochastic Modelling of the Collateral Choice Option and its Practical Implications

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The collateral choice option



- Assets are perfectly collateralized: no frictions, entire asset is covered at all times.
- No default risk, discounting with time value of money.
- Single curve framework based on *collateral rate r*.
- Asset valuation becomes

$$V_0 = \mathbb{E}^{\mathbb{Q}} \big[\mathrm{e}^{-\int_0^T r(t) \mathrm{d}t} V_T \big].$$

The collateral choice option



$$V_0 = \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_0^T \cdot ? \, \mathrm{d}t} V_T \right]$$

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The collateral choice option



$$V_0 = \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_0^T \max(r_0(t), \dots, r_N(t)) dt} V_T \right]$$

Goal: Express the option without path and asset dependence.

$$\mathbb{E}^{\mathbb{Q}}\left[\exp\left(-\int_{0}^{T}\max(r_{0}(t),\ldots,r_{N}(t))\mathrm{d}t\right)V_{T}\right]$$

$$=\mathbb{E}^{\mathbb{Q}}\left[\exp\left(-\int_{0}^{T}r_{0}(t)+\max(0,(r_{1}-r_{0})(t),\ldots,(r_{N}-r_{0})(t))\mathrm{d}t\right)V_{T}\right]$$

$$= \mathbb{E}^{\mathbb{Q}}\left[\exp\left(-\int_{0}^{T} \max(0, q_{1}(t), \dots, q_{N}(t))dt\right) \exp\left(-\int_{0}^{T} r_{0}(t)dt\right) V_{T}\right]$$

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$$V_0 = \mathbb{E}^{\mathbb{Q}} \left[e^{-\int_0^T \max(0, q_1(t), \dots, q_N(t)) dt} e^{-\int_0^T r_0(t) dt} V_T \right]$$

Strong independence (all q_i from r_0 and V):

$$= \mathbb{E}^{\mathbb{Q}} \left[\mathrm{e}^{-\int_0^T \max(0, q_1(t), \dots, q_N(t)) \mathrm{d}t} \right] \mathbb{E}^{\mathbb{Q}} \left[\mathrm{e}^{-\int_0^T r_0(t) \mathrm{d}t} V_T \right].$$

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Strong independence (all q_i from r_0 and V):

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Weak independence (all q_i from V):

$$=\mathbb{E}^{\mathbb{Q}^{T}}\left[\mathrm{e}^{-\int_{0}^{T}\max(0,q_{1}(t),\ldots,q_{N}(t))\mathrm{d}t}\right]\mathbb{E}^{\mathbb{Q}}\left[\mathrm{e}^{-\int_{0}^{T}r_{0}(t)\mathrm{d}t}V_{T}\right].$$

For either measure in $\{\mathbb{Q}, \mathbb{Q}^T\}$:

$$\operatorname{CTD}(0, T) = \mathbb{E}\left[e^{-\int_0^T \max(0, q_1(t), \dots, q_N(t)) \mathrm{d}t}\right].$$

Hard to evaluate (in general)!

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Pragmatic solution: deterministic model

$$\mathrm{CTD}_{\mathrm{det}}(0, T) = \mathrm{e}^{-\int_0^T \max(0, \mathbb{E}[q_1(t)], \dots, \mathbb{E}[q_N(t)]) \mathrm{d}t}.$$

The deterministic price is easily obtained with a composite curve.



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With stochastic dynamics...



With stochastic dynamics the maximum gets bigger!



Jensen's inequality: $\max(0, \mathbb{E}[q_{\mathsf{EUR}}], \mathbb{E}[q_{\mathsf{GBP}}]) \leq \mathbb{E}[\max(0, q_{\mathsf{EUR}}, q_{\mathsf{GBP}})]$



Spreads q_i behave similar to interest rates.

Modelling them with stochastic interest rate models (Gaussian processes) gives rise to the maximum process $M(t) = \max(0, q_1(t), \dots, q_N(t))$.

- No closed form marginal distributions $M(t) = \max(0, q_1(t), \dots, q_N(t)).$
- No closed form process distribution $(M(t))_{t \in [0,T]}$.

• CTD(0,
$$T$$
) = $\mathbb{E}\left[\exp\left(-\int_0^T M(t) dt\right)\right] = \mathbb{E}\left[\sum_{k=0}^\infty \frac{\left(-\int_0^T M(t) dt\right)^k}{k!}\right].$

Stochastic model: Taylor approximation

$$\operatorname{CTD}(0, T) = \mathbb{E}\Big[\sum_{k=0}^{\infty} \frac{\left(-\int_{0}^{T} M(t) \mathrm{d}t\right)^{k}}{k!}\Big].$$

First-order approximation is not precise enough, second-order approximation appears suitable:

$$\operatorname{CTD}(0, T) \approx \exp\left(-\mathbb{E}\left[\int_{0}^{T} M(t) \mathrm{d}t\right]\right) \left(1 + \frac{1}{2} \mathbb{V}\operatorname{ar}\left[\int_{0}^{T} M(t) \mathrm{d}t\right]\right),$$

but is still dependent on the marginal distributions M(t) and covariances $\mathbb{C}ov[M(t), M(s)]!$

Stochastic model: process approximation



- Obtain marginal distribution from conditional independence assumption (common factor model)
- Obtain process dynamics by approximation with similar processes for which it is known (Itô or mean-reverting diffusion)

Finally, we have three quantities:

• True value:
$$\operatorname{CTD}(0, T) = \mathbb{E}\left[\exp(-\int_0^T M(t) dt\right].$$

- Deterministic model: $CTD_{det}(0, T)$.
- Stochastic model: $CTD_{CF}(0, T)$.

Stochastic and deterministic CTD models in practise

Consider an interest rate swap with the collateral choice option:

$$\begin{aligned} \mathcal{V}^{c}(t) &= \sum_{k=1}^{m} \operatorname{CTD}(t, T_{k}) U_{k}(t) \\ &= \bar{N} \sum_{k=1}^{m} \operatorname{CTD}(t, T_{k}) \Big(\tau_{k} P(t, T_{k}) \big(\ell(t, T_{k-1}, T_{k}) - \mathcal{K} \big) \Big). \end{aligned}$$

"True values" $CTD(t, T_k)$ are unknown but we can approximate the discount factors with our models CTD_{det} and CTD_{CF} .

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Consider portfolios Π_j , $j \in \{\det, CF\}$:

$$\Pi_j(t) = V^c(t) - \bar{N} \sum_{T_k \ge t} \operatorname{CTD}_j(t, T_k) \tau_k(\ell(t, T_{k-1}, T_k) - K) P(t, T_k).$$



Accumulated valuation error over the lifetime of a swap.

$$\Pi_j(t) = V^c(t) - \bar{N} \sum_{T_k \ge t} \operatorname{CTD}_j(t, T_k) \tau_k(\ell(t, T_{k-1}, T_k) - K) P(t, T_k).$$

These portfolios are comparable to a Delta hedge of the swap V^c .

$$V^{c}(t) = \sum_{k=1}^{m} \operatorname{CTD}(t, T_{k}) U_{k}(t).$$

The linear product $\sum_{k} U_k$ only requires a Delta hedge, which risk factors are introduced by the CTD factor?

Hedging perspective



The stochastic maximum depends on all components at all times.

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Hedging perspective



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Risk factors are the "ingredients" of the asset price that change stochastically.

Changes in a risk factor X move the price of the asset:

$$X \uparrow \Longrightarrow V = f(X) \uparrow$$

Hedges are assets that move in opposite direction under these changes:

$$X\uparrow \Longrightarrow W = g(X)\downarrow$$

Together, asset and hedge are *neutral* to the risk factor:

$$X \uparrow \implies (V + W) \uparrow \downarrow \text{(no change)}.$$

Call Option example (Black–Scholes market)

Consider Geometric Brownian motion stock

 $\mathrm{d}S_t = \mu(t)S_t\mathrm{d}t + \sigma(t)S_t\mathrm{d}W_t.$

A European call option C pays $max(S_T - K, 0)$, hence

 $S\uparrow \Longrightarrow C\uparrow$.

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Delta-Hedge: Risk factor S has hedge -S: $S \uparrow \implies -S \downarrow$ $S \uparrow \implies (C + \Delta S) \uparrow \downarrow$ is neutral (for some factor Δ).

Call Option example (Black-Scholes market)

Delta-hedge portfolio $\Pi(t) = C_t + \Delta(t)S_t$



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Call Option example (Heston market)

If stock S moves with Heston dynamics

$$dS_t = \mu S_t dt + \sqrt{v_t} S_t dW_t^S,$$

$$dv_t = \kappa(\theta(t) - v_t) dt + \xi \sqrt{v_t} dW_t^v,$$

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Volatility v is another risk factor!

 $v \uparrow \Longrightarrow C \uparrow$



Delta-Vega-hedge portfolio $\Pi(t) = C_t + \Delta^*(t)S_t + \nu(t)C_t$

Hedging the collateral choice option

What are the risk factors in an asset with the collateral choice option:

$$V^{c}(t) = \operatorname{CTD}(t, T)V(t)$$
?

- V is linear \implies only risk factor is its underlying $\implies \Delta_V$.
- CTD is convex $\implies \Delta_{\text{CTD}}, \nu_{\text{CTD}}$?

Collateral spreads cannot be traded, what is a hedging asset for them?

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Collateral spreads cannot be traded, what is a hedging asset for them?

- Create artificial products "bond on collateral spreads" $Q_i(t, T) = \mathbb{E}[\exp(-\int_t^T q_i(s) ds)]$ and options on it.
- Are there liquidly traded instruments that can serve as a proxy for the same risk factors?