

The Influence of FX Risk on Credit Spreads

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Philippe Ehlers
ETH Zurich

`ehlers@math.ethz.ch`

`www.math.ethz.ch/~ehlers`

Agenda

- A. The Problem
- B. Mathematical Model
- C. Empirical Results
- D. Application in Cat Insurance
- E. Conclusion

A. The Problem

Q: *What is the price of Credit Event Protection in different currencies?*

Some Notation

- X_t is the US\$-JPY FX rate at time t .
- τ is the default time of an obligor.
- q_τ is the loss given default (LGD) rate of this obligor's bonds.
- $S_\$$ and S_Y are T -year US\$ and JPY Credit Default Swap (CDS) rates at time 0.
- $b_\$(t)$ is a US\$- and $b_Y(t)$ a JPY-bank account at time t .

CDS Cash Flows

(Viewpoint of protection buyer)

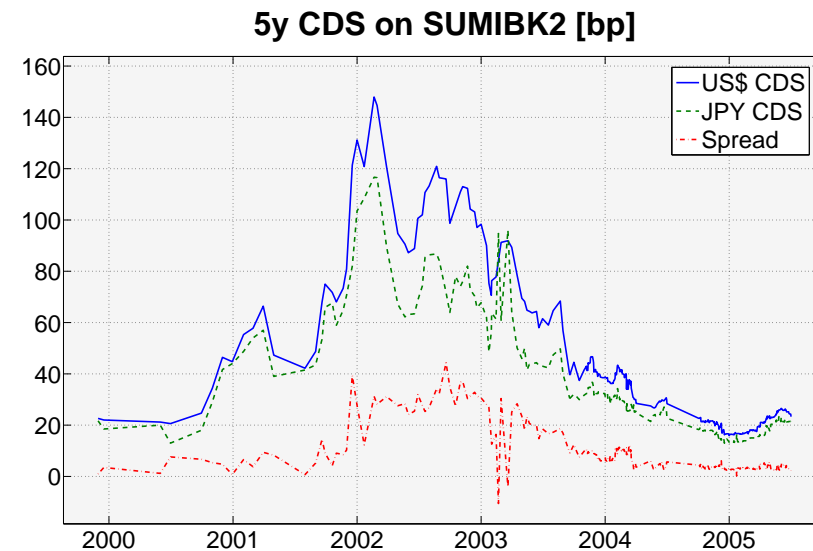
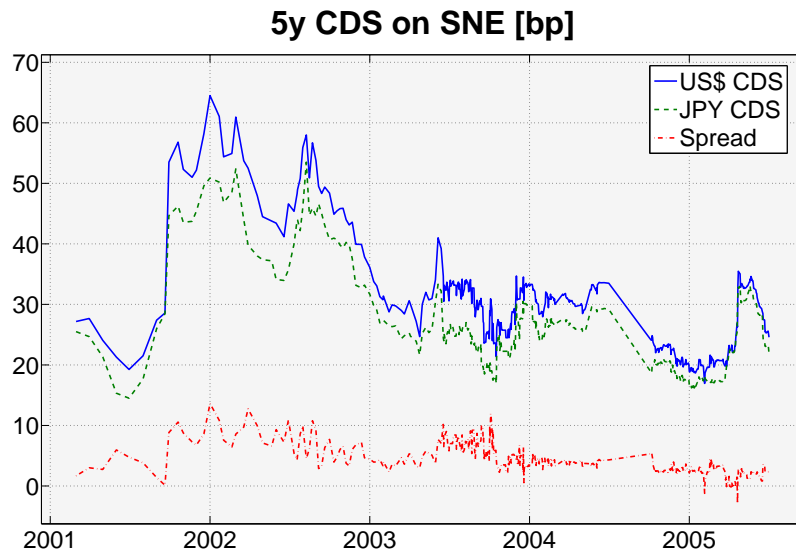
at	US\$ CDS	JPY CDS	Payment
$t \leq \min(\tau, T)$	$-S_{\$}dt$ US\$	$-S_Y dt$ JPY =	$-X_t S_Y dt$ US\$
τ iff $\tau \leq T$	q_{τ} US\$	q_{τ} JPY =	$X_{\tau} q_{\tau}$ US\$

Intuition. Dependence of X and (τ, q_{τ}) is essential for relationship of $S_{\$}$ and S_Y .

Benchmark case. X independent of (τ, q_{τ}) and $b_{\$}(t) = b_Y(t) = 1$, then

$$S_{\$} = S_Y.$$

Evidence from the CDS Markets



Economic Reasons

- US\$-JPY FX impacts on profits of Japanese Export/Import companies and on liability side of Japanese companies that issued debt in US\$ (*direct* dependence).
- Credit quality of major Japanese obligors (in particular banks) may influence investors' confidence in JPY (*indirect* dependence).

B. Mathematical Model

Preliminaries. Probability space $(\Omega, \mathcal{F}_T, \mathbf{Q})$, finite time horizon T , filtration $(\mathcal{F}_t)_{t \leq T}$, \mathbf{Q} the US\$ spot martingale measure (\$SMM) chosen by the market.

Proposition.

$$\left. \frac{d\tilde{\mathbf{Q}}}{d\mathbf{Q}} \right|_{\mathcal{F}_t} := L_t := \frac{X_t b_Y(t)}{X_0 b_{\$}(t)}$$

defines a probability $\tilde{\mathbf{Q}}$ on (Ω, \mathcal{F}_T) . $\tilde{\mathbf{Q}}$ is a JPY spot martingale measure (YSMM).

Change of Measure Method. We have

$$S_{\$} = \frac{\mathbf{E}^{\mathbf{Q}} \left[\frac{1}{b_{\$}(\tau)} \mathbf{1}_{\{\tau \leq T\}} q_{\tau} \right]}{\int_0^T \mathbf{E}^{\mathbf{Q}} \left[\frac{1}{b_{\$}(t)} \mathbf{1}_{\{t \leq \tau\}} \right] dt} \quad \text{and} \quad S_Y = \frac{\mathbf{E}^{\tilde{\mathbf{Q}}} \left[\frac{1}{b_Y(\tau)} \mathbf{1}_{\{\tau \leq T\}} q_{\tau} \right]}{\int_0^T \mathbf{E}^{\tilde{\mathbf{Q}}} \left[\frac{1}{b_Y(t)} \mathbf{1}_{\{t \leq \tau\}} \right] dt}.$$

Default Risk under \mathbf{Q}

- $(\mathcal{F}_t)_{t \leq T}$ carries a Brownian Motion (BM) W and a jump measure $\mu(dz, dt)$ on $\mathcal{Z} \times [0, T]$ with predictable \mathbf{Q} -compensator of the form

$$\nu^{\mathbf{Q}}(dz, dt) = dF_t(z)\lambda(t)dt,$$

where $F_t(\cdot)$ is a distribution function on \mathcal{Z} and $d\lambda(t) = \gamma_\lambda(t)dt + \sigma'_\lambda(t)dW_t$.

- $\tau = \inf\{t > 0; N_t > 0\}$, where $N_t := \int_0^t \int_{\mathcal{Z}} \mu(dz, ds)$.
- $q_\tau = \int_0^\tau \int_{\mathcal{Z}} q(z, t)\mu(dz, dt)$ for some predictable function $q(z, t)$.

Limiting Property.

$$\lim_{T \downarrow 0} S_{\S}(T) = \bar{q}(0)\lambda(0) \quad \mathbf{Q}\text{-a.s.},$$

where $\bar{q}(t) := \int_{\mathcal{Z}} q(z, t)dF_t(z)$.

FX Rate Modelling

Key Assumption. L satisfies the stochastic differential equation (SDE)

$$\frac{dL_t}{L_{t-}} = dM_t := \sigma'_X(t) dW_t - \int_{\mathcal{Z}} \delta(z, t) [\mu(dz, dt) - \nu^{\mathbf{Q}}(dz, dt)]$$

where $\delta(z, t) < 1$ is a predictable function.

That means there can be

- Correlation *before default*: $d[X, \lambda]_t = \sigma'_X(t) \sigma_\lambda(t) X_{t-} dt$,
- Devaluation *at default*: $X_\tau = (1 - \delta_\tau) X_{\tau-}$, where $\delta_\tau := \int_0^\tau \int_{\mathcal{Z}} \delta(z, t) \mu(dz, dt)$.

Default Risk under $\tilde{\mathbf{Q}}$ (Girsanov)

- Correlation \implies Drift changes in underlying diffusions (e.g. λ)
- Devaluation \implies
 - ★ Different default intensity $\tilde{\lambda}(t) := [1 - \bar{\delta}(t)] \lambda(t)$,
 - ★ Different LGD distribution $d\tilde{F}_t(z)$

where $\bar{\delta}(t) := \int_{\mathcal{Z}} \delta(z, t) dF_t(z)$.

Remember:

$$\lim_{T \downarrow 0} S_Y(T) = \tilde{q}(0) \tilde{\lambda}(0) = \tilde{q}(0) [1 - \bar{\delta}(0)] \lambda(0) \quad \mathbf{Q}\text{-a.s.},$$

where $\tilde{q}(t) := \int_{\mathcal{Z}} q(z, t) d\tilde{F}_t(z)$.

C. Empirical Results

Word on the Data. ValuSpread CDS data base by *Lombard Risk Systems Ltd.*

Devaluation Fraction. We use the estimator $\hat{\delta}_T := \text{mean} \left\{ 1 - \frac{S_Y(T)}{S_{\$}(T)} \right\}$.

Ticker (<i>Company</i>)	1 year CDS		5 year CDS	
	$\hat{\delta}_1$ [%]	p-Value [%]	$\hat{\delta}_5$ [%]	p-Value [%]
EJRAIL (<i>East Japan Railway Company</i>)	20.2	0.00	18.4	0.00
HONDA (<i>Honda Motor Co., Ltd.</i>)	12.5	0.00	19.2	0.00
MATSEL (<i>Matsushita Electric Industrial Co., Ltd.</i>)	24.6	0.00	17.0	0.00
MITCO1 (<i>Mitsubishi Corp</i>)	18.4	0.00	10.7	0.00
MITSCO (<i>Mitsui & Co., Ltd.</i>)	14.7	0.00	11.8	0.00
NIPSTL (<i>Nippon Steel Corporation</i>)	16.2	0.00	11.8	0.00
NTT (<i>Nippon Telegraph & Telephone Corporation</i>)	27.0	0.00	19.0	0.00
SHARP (<i>Sharp Corporation</i>)	31.4	0.00	21.1	0.00
SNE (<i>Sony Corporation</i>)	27.0	0.00	16.5	0.00
SUMIBK2 (<i>Sumitomo Mitsui Banking Corporation</i>)	21.7	0.00	20.6	0.00
SUMT (<i>Sumitomo Corporation</i>)	13.5	0.00	9.6	0.00
TOKELP (<i>Tokyo Electric Power Co., Inc.</i>)	31.0	0.00	18.9	0.00
TOKIO (<i>Tokio Marine and Fire Insurance Company Limited</i>)	29.0	0.00	22.8	0.00
TOYOTA1 (<i>Toyota Motor Corporation</i>)	30.6	0.00	25.6	0.00
YAMAHA (<i>Yamaha Motor Co., Ltd.</i>)	16.9	0.00	16.6	0.00

Affine Diffusion Correlation Model

$$d\lambda(t) = \kappa[\theta - \lambda(t)]dt + \sigma\sqrt{\lambda(t)}dW_t^1$$

$$dM(t) = \xi_1\sqrt{\lambda(t)}dW_t^1 + \xi_2dW_t^2 - \delta[dN_t - \lambda(t)dt]$$

Ticker (<i>Company</i>)	$\widehat{\left(\frac{\xi_1}{q\sigma}\right)}$	95%-CI	Implied $\widehat{\delta}_5$ [%]
EJRAIL (<i>East Japan Railway Co.</i>)	-2.97	[-10.22, 4.29]	1.0
HONDA (<i>Honda Motor Co. Ltd.</i>)	-2.38	[-7.28, 2.52]	0.8
MATSEL (<i>Matsushita Electric Industrial Co. Ltd.</i>)	-1.05	[-4.03, 1.93]	0.9
MITCO1 (<i>Mitsubishi Corp.</i>)	-3.97	[-8.18, 0.23]	1.7
MITSCO (<i>Mitsui & Co. Ltd.</i>)	-4.20	[-8.29, -0.12]*	1.2
NIPSTL (<i>Nippon Steel Corp.</i>)	-5.61	[-9.19, -2.03]*	-0.8
NTT (<i>Nippon Telegraph & Telephone Corp.</i>)	-2.73	[-10.58, 5.12]	2.3
SHARP (<i>Sharp Corp.</i>)	-3.96	[-7.28, -0.65]*	0.6
SNE (<i>Sony Corp.</i>)	-1.97	[-6.12, 2.17]	-3.0
SUMIBK2 (<i>Sumitomo Mitsui Banking Corp.</i>)	-1.94	[-5.69, 1.81]	-4.6
SUMT (<i>Sumitomo Corp.</i>)	-3.80	[-7.39, -0.21]*	-1.2
TOKELP (<i>Tokyo Electric Power Co. Inc.</i>)	-3.10	[-7.892, 1.70]	-2.0
TOKIO (<i>Tokyo Marine & Fire Insurance Co. Ltd.</i>)	-1.51	[-4.39, 1.37]	0.9
TOYOTA1 (<i>Toyota Motor Corp.</i>)	-1.68	[-11.02, 7.66]	1.5
YAMAHA (<i>Yamaha Motor Co., Ltd.</i>)	0.26	[-1.76, 2.28]	3.4

D. Application in Cat Insurance

- τ is time of next earthquake in Tokio of magnitude > 8.5
- X_t^* is replacement value [US\$] of a skyscraper in Tokio ($X_0^* = 1\text{bn US\$}$)

Example: 3 Property Insurance Contracts

Contract	Premium [US\$]	Sum Insured
A	$S_{\$}$	1 bn US\$
B	S_Y	111 bn JPY
C	S_{repl}	X_{τ}^* US\$

Analogy. If FX rate or replacement value may be devaluated at τ , then possibly

$$S_{\$} \neq S_Y \quad \text{or} \quad S_{repl} \neq S_{\$}.$$

E. Conclusion

- US\$ and JPY CDSs on the same Japanese reference entity are not equal.
- FX risk and default risk of these obligors are not independent.
- Our model explains this spread by
 - ★ Correlation of US\$-JPY FX and CDS rates before default,
 - ★ Devaluation of JPY at default.
- Correlation does not sufficiently explain our CDS data.
- The market expects JPY devaluation at default of major Japanese obligors.
- Applications in Cat Insurance when τ is “Catastrophe Time”.