

# **Martingale optimality and cross hedging of insurance derivatives\***

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# 1 Insurance derivatives

## Weather derivatives

- underlyings: e.g. temperature, rainfall or snowfall indices
- **Aim:** transfer exogenous risk caused by fluctuations in weather patterns to capital markets

## Catastrophe futures

- underlyings: e.g. loss, windspeed index (hurricane bond)
- **Aim:** transfer exogenous (insurer's) risk of abnormal losses to other parts of economy
- alternative to traditional catastrophe reinsurance

## 2 Hedging of portfolios of insurance derivatives

Problem: underlying is **not tradable**, but **correlated with tradable assets**

### Examples:

**temperature**  $\longleftrightarrow$  **heating oil futures, electricity futures**

**loss index**  $\longleftrightarrow$  **stock price of insurance companies**

Mutual hedging of weather derivatives

**rainfall in Spain**  $\longleftrightarrow$  **rainfall in Scandinavia**

### Aims:

determine **utility indifference price**

determine **explicit derivative hedge**, i.e. optimal cross hedging strategy

describe reduction of risk by **cross hedging**

compare **dynamic** with **static risk**

interpret **pricing** by **marginal utility**

### 3 The financial market model

Index process, e.g. temperature

$$dR_t = \rho(t, R_t)dW_t + b(t, R_t)dt,$$

$b : [0, T] \times \mathbf{R}^m \rightarrow \mathbf{R}^m$ ,  $\rho : [0, T] \times \mathbf{R}^m \rightarrow \mathbf{R}^{m \times d}$  deterministic functions, globally Lipschitz and of sublinear growth.  $R$  Markov process,  $R_s^{t,r}$ : start at  $t$  in  $r$

Insurance derivative  $F(R_T)$ ,  $F : \mathbf{R}^m \rightarrow \mathbf{R}$  bounded

Correlated financial market,  $k$  risky assets with price process:

$$\frac{dS_t^i}{S_t^i} = \beta_i(t, R_t)dW_t + \alpha_i(t, R_t)dt = \beta_i(t, R_t)[dW_t + \theta_t dt], \quad i = 1, \dots, k,$$

$\alpha : [0, T] \times \mathbf{R}^m \rightarrow \mathbf{R}^k$ ,  $\beta : [0, T] \times \mathbf{R}^m \rightarrow \mathbf{R}^{k \times d}$ ,  $\theta = \beta^*[\beta\beta^*]^{-1}\alpha$ .

$W$   $d$ -dimensional Brownian motion,  $\beta\beta^*$  uniformly elliptic

## 4 The optimal investment problem

(N. El Karoui, R. Rouge '00; J. Sekine '02; J. Cvitanic, J. Karatzas '92, Kramkov, Schachermayer '99,...)

investment strategy  $\lambda$  : value of portfolio fraction invested in risky assets

wealth gain on  $[t, s]$

$$G_s^{\lambda,t} = \sum_{i=1}^k \int_t^s \lambda_u^i \frac{dS_u^i}{S_u^i} = \int_t^s \lambda_u \beta_u [dW_u + \theta_u du],$$

utility function:  $U(x) = -e^{-\eta x}$  ( $0 < \eta$  risk aversion); maximal expected utility from terminal wealth **without** and **with** derivative:

$$V^0(t, v, r) = \sup_{\lambda \in \mathcal{A}^{t,r}} EU(v + G_T^{\lambda,t,r}), \quad V^F(t, v, r) = \sup_{\lambda \in \mathcal{A}^{t,r}} EU(v + G_T^{\lambda,t,r} - F(R_T^{t,r}))$$

$\lambda^0$  resp.  $\lambda^F$  optimal strategies

$\Delta = \lambda^F - \lambda^0$  **derivative hedge**

## 5 Optimization under non-convex constraints

interpretation as maximization problem **with constraints**

$$\pi(t, r) = \lambda(t, r)\beta(t, r) \in C(t, r) = \{x\beta(t, r) : x \in \mathbf{R}^k\}$$

here:  $C(t, r)$  **convex**

**Aim:** construct solution **using BSDE**, even for **non-convex constraints**

(N. El Karoui, R. Rouge '00 for convex constraints)

$$\tilde{C} \subset \mathbf{R}^k \text{ closed}$$

$\tilde{\mathcal{A}}$  set of strategies  $\lambda$  such that

- $\lambda \in \tilde{C}$   $P \otimes l$ -a.s. ( $l$  Lebesgue measure)
- $\{\exp(-\eta \int_0^\tau \lambda_s \frac{dS_s}{S_s}) : \tau \text{ stopping time in } [0, T]\}$  uniformly integrable

## 5 Optimization under non-convex constraints

$$F = F(R_T) \text{ insurance derivative}$$

For simplicity  $t = 0$ ,  $G^{\lambda,0} = G^\lambda$ ,  $V(v) = V^F(0, v, r)$ , etc.

### First formulation:

Find

$$V(v) = \sup_{\lambda \in \tilde{\mathcal{A}}} E(U(G_T^\lambda - F)) = \sup_{\lambda \in \tilde{\mathcal{A}}} E(U(v + \int_0^T \lambda_s \beta_s [dW_s + \theta_s ds] - F)).$$

For simplicity:

$$\begin{aligned} \pi &= \lambda \beta, \\ C &= \tilde{C} \beta, \\ \mathcal{A} &= \tilde{\mathcal{A}} \beta. \end{aligned}$$

$$G_t^\pi = v + \int_0^t \pi_s [dW_s + \theta_s ds], \quad t \in [0, T]$$

### Second formulation:

Find

$$V(v) = \sup_{\pi \in \mathcal{A}} E(U(G_T^\pi - F)) = \sup_{\pi \in \mathcal{A}} E(-\exp(-\eta(x + \int_0^T \pi_s [dW_s + \theta_s ds] - F))).$$

## 6 A solution method based on BSDE

**Idea:** Construct family of processes  $Q^{(\pi)}$  such that

**form 1**

$$\begin{aligned}
 Q_0^{(\pi)} &= \text{constant}, \\
 Q_T^{(\pi)} &= -\exp(-\eta(G_T^\pi - F)), \\
 Q^{(\pi)} &\text{ supermartingale, } \pi \in \mathcal{A}, \\
 Q^{(\pi^*)} &\text{ martingale, for (exactly) one } \pi^* \in \mathcal{A}.
 \end{aligned}$$

Then

$$\begin{aligned}
 E(-\exp(-\eta[G_T^\pi - F])) &= E(Q_T^{(\pi)}) \\
 &\leq E(Q_0^\pi) \\
 &= V(v) \\
 &= E(Q_0^{(\pi^*)}) \\
 &= E(-\exp(-\eta[G_T^{(\pi^*)} - F])).
 \end{aligned}$$

Hence  $\pi^*$  optimal strategy.



## 6 A solution method based on BSDE

### Introduction of BSDE into problem

Find generator  $f$  of BSDE

$$Y_t = F - \int_t^T Z_s dW_s - \int_t^T f(s, Z_s) ds, \quad Y_T = F,$$

such that with

$$Q_t^{(\pi)} = -\exp(-\eta[G_t^\pi - Y_t]), \quad t \in [0, T],$$

we have

$$\begin{aligned} Q_0^{(\pi)} &= -\exp(-\eta(v - Y_0)) \\ &= \text{constant}, \end{aligned} \quad (\text{fulfilled})$$

**form 2** 
$$Q_T^{(\pi)} = -\exp(-\eta(G_T^\pi - F)) \quad (\text{fulfilled})$$

$$\begin{aligned} Q^{(\pi)} & \text{ supermartingale, } \pi \in \mathcal{A}, \\ Q^{(\pi^*)} & \text{ martingale, for (exactly) one } \pi^* \in \mathcal{A}. \end{aligned}$$

This gives solution of valuation problem.

## 7 Construction of generator of BSDE

How to determine  $f$ :

Suppose  $f$  generator of BSDE. Then

$$\begin{aligned}
 Q_t^{(\pi)} &= -\exp(-\eta[G_t^\pi - Y_t]) \\
 &= -\exp(-\eta[v - Y_0]) \cdot \exp(-\eta[\int_0^t (\pi_s - Z_s)dW_s - \int_0^t [f(s, Z_s) - \pi_s\theta_s]ds]) \\
 &= \exp(-\eta[v - Y_0]) \cdot \exp(-\eta\int_0^t (\pi_s - Z_s)dW_s - \frac{\eta^2}{2}\int_0^t (\pi_s - Z_s)^2 ds) \\
 &\quad \cdot \exp(\int_0^t [\eta f(s, Z_s) - \eta\pi_s\theta_s + \frac{\eta^2}{2}(\pi_s - Z_s)^2]ds) \\
 &= M_t^{(\pi)} \cdot A_t^{(\pi)},
 \end{aligned}$$

with  $M^{(\pi)}$  nonnegative martingale.  $Q^{(\pi)}$  satisfies **(form 2)** iff for

$$q(\cdot, \pi, z) = f(\cdot, z) - \pi\theta + \frac{\eta}{2}(\pi - z)^2, \quad \pi \in \mathcal{A}, z \in \mathbb{R},$$

we have

## 7 Construction of generator of BSDE

**form 3**  $q(\cdot, \pi, z) \geq 0, \pi \in \mathcal{A}$  (supermartingale cond.)  
 $q(\cdot, \pi^*, z) = 0,$  for (exactly) one  $\pi^* \in \mathcal{A}$  (martingale cond.).

Now

$$\begin{aligned} q(\cdot, \pi, z) &= f(\cdot, z) - \pi\theta + \frac{\eta}{2}(\pi - z)^2 \\ &= f(\cdot, z) + \frac{\eta}{2}(\pi - z)^2 - (\pi - z) \cdot \theta + \frac{1}{2\eta}\theta^2 - z\theta - \frac{1}{2\eta}\theta^2 \\ &= f(\cdot, z) + \frac{\eta}{2}\left[\pi - \left(z + \frac{1}{\eta}\theta\right)\right]^2 - z\theta - \frac{1}{2\eta}\theta^2. \end{aligned}$$

Under **non-convex constraint**  $\pi \in C$ :

$$\left[\pi - \left(z + \frac{1}{\eta}\theta\right)\right]^2 \geq d^2\left(C, z + \frac{1}{\eta}\theta\right).$$

with **equality** for at least one possible choice of  $p^*$  due to **closedness** of  $C$ .  
Hence **(form 3)** is solved by the choice

## 7 Construction of generator of BSDE

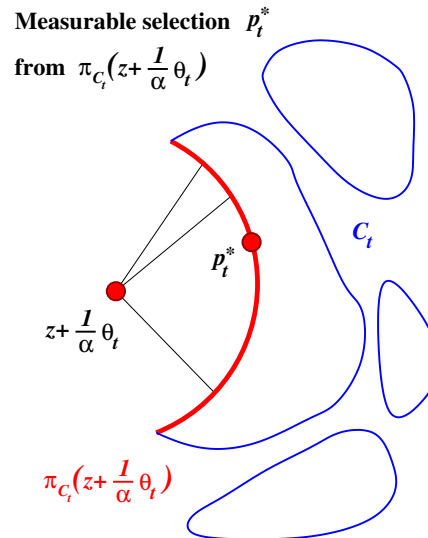
**form 4**

$$f(\cdot, z) = -\frac{\eta}{2}d^2(C, z + \frac{1}{\eta}\theta) + z \cdot \theta + \frac{1}{2\eta}\theta^2 \quad (\text{supermartingale})$$

such that  $d(C, z + \frac{1}{\eta}\theta) = d(\pi^*, z + \frac{1}{\eta}\theta)$  (martingale).

**Problem:** Let

$\Pi_C(v) = \{\pi \in \mathbb{R}^d : d(C, v) = d(\pi, v)\}$ . Find measurable selection  $\pi_t^*$  from  $\Pi_{C_t}(Z_t + \frac{1}{\eta}\theta_t)$ . Solved by classical **measurable selection method**.



## 8 Main result

### Thm 1

$(Y, Z)$  unique solution of BSDE

$$Y_t = F - \int_t^T Z_s dW_s - \int_t^T f(s, Z_s) ds, \quad t \in [0, T],$$

with

$$f(t, Z_t) = -\frac{\eta}{2} d^2(C_t, Z_t + \frac{1}{\eta} \theta_t) + Z_t \cdot \theta_t + \frac{1}{2\eta} \theta_t^2.$$

Then **value function** of utility optimization problem under **constraint**  $\pi \in \mathcal{A}$  given by

$$V(v) = -\exp(-\eta[v - Y_0]).$$

There exists an (non-unique) **optimal trading strategy**  $\pi^* \in \mathcal{A}$  such that

$$\pi_t^* \in \Pi_{C_t}(Z_t + \frac{1}{\eta} \theta_t), \quad t \in [0, T].$$

### Proof:

- **existence, uniqueness for BSDE** with quadratic non-linearity in  $z$  (M. Kobylanski '00)
- **measurable selection theorem** for  $\Pi_{C_t}(Z_t + \frac{1}{\eta} \theta_t)$
- **BMO properties** of the martingales  $\int Z_s dW_s, \int \pi_s^* dW_s$  for **uniform integrability of exponentials** (regularity of coefficients) •

## 9 Calculation of derivative hedge

generalization to  $[t, T]$  instead of  $[0, T]$ , cond. on  $R_t = r$ :

$(Y^{t,r}, Z^{t,r}), \pi^{t,r}$  (without  $F$ ) resp.  $(\hat{Y}^{t,r}, \hat{Z}^{t,r}), \hat{\pi}^{t,r}$  (with  $F$ ) instead of  $(Y, Z), \pi$   
yields

$$V^0(t, v, r) = -\exp(-\eta(v - Y_t^{t,r})), \quad V^F(t, v, r) = -\exp(-\eta(v - \hat{Y}_t^{t,r})),$$

instead of  $V(v) = -\exp(v - Y_0)$ .

due to **linearity of  $C(t, r)$**  projections unique and linear, hence

$$\pi_s^{t,r} = \Pi_{C(t,r)}[Z_s^{t,r} + \frac{1}{\eta}\theta(s, R_s^{t,r})], \quad \hat{\pi}_s^{t,r} = \Pi_{C(t,r)}[\hat{Z}_s^{t,r} + \frac{1}{\eta}\theta(s, R_s^{t,r})],$$

and so

$$\Delta\beta(s, R_s^{t,r}) = \Pi_{C(t,r)}[\hat{Z}_s^{t,r} - Z_s^{t,r}].$$

## 10 Markov property and its consequences

Markov property of  $R$  implies (Kobylanski '00, El Karoui, Peng, Quenez '97):

### Thm 2

There are measurable (deterministic) functions  $u$  and  $\hat{u}$  such that

$$Y_s^{t,r} = u(s, R_s^{t,r}), \quad \hat{Y}_s^{t,r} = \hat{u}(s, R_s^{t,r}).$$

There are measurable (deterministic) functions  $v$  and  $\hat{v}$  such that

$$Z_s^{t,r} = v\rho(s, R_s^{t,r}), \quad \hat{Z}_s^{t,r} = \hat{v}\rho(s, R_s^{t,r}).$$

### Corollary 1

$$p(t, r) := Y_t^{t,r} - \hat{Y}_t^{t,r} = u(t, r) - \hat{u}(t, r)$$

is the **indifference price**, i.e.  $V^F(t, v - p(t, r), r) = V^0(t, v, r)$ .

$p$  depends only on  $R$ , not on  $S$

**Aim:** Explicit description of  $\Delta$

# 11 Differentiability

**Thm 3 (Parameter Differentiability)** smoothness conditions on  $F, f$   
There exists a version of  $(\widehat{Y}_s^{t,r}, \widehat{Z}_s^{t,r})$  such that a.s.

- $\widehat{Y}_s^{t,r}$  is continuous in  $s$  and **cont. differentiable in  $r$**  (classical sense)
- $\widehat{Z}_s^{t,r}$  is **differentiable in a weak sense** (norm topology)
- $(\nabla_r \widehat{Y}_s^{t,r}, \nabla_r \widehat{Z}_s^{t,r})$  solves the BSDE

$$\begin{aligned} \nabla_r \widehat{Y}_t^{t,r} &= \nabla_r F(R_s^{t,r}) \nabla_r R_s^{t,r} - \int_t^T \nabla_r \widehat{Z}_s^{t,r} dW_s \\ &\quad + \int_t^T \left[ \nabla_r f(s, R_s^{t,r}, \widehat{Z}_s^{t,r}) \nabla_r R_s^{t,r} \right. \\ &\quad \left. + \nabla_z f(s, R_s^{t,r}, \widehat{Z}_s^{t,r}) \nabla_r \widehat{Z}_s^{t,r} \right] ds. \end{aligned}$$

Proof uses norm inequalities, and inverse Hölder inequalities, based on BMO properties of the stochastic integral processes of  $\widehat{Z}_s^{t,r}$

**Thm 4 (Malliavin Differentiability)**

$$D_\vartheta \widehat{Y}_s^{t,r} = \nabla_r \widehat{u}(s, R_s^{t,r}) D_\vartheta R_s^{t,r}$$

and

$$\widehat{Z}_s^{t,r} = D_s \widehat{Y}_s^{t,r} = \nabla_r \widehat{u}(s, R_s^{t,r}) \rho(s, R_s^{t,r})$$



## 12 Explicit description of derivative hedge

Properties of the BSDEs  $\implies$

### Thm 5

The **indifference price**  $p(t, r) = Y_t^{t,r} - \widehat{Y}_t^{t,r}$  is **differentiable** in  $r$ .

### Thm 6

The **derivative hedge**  $\Delta$  at time  $t$  depends only on  $R_t$ , and

$$\begin{aligned} \Delta(t, r)\beta(t, r) &= \Pi_{C(t,r)}[\widehat{Z}_t^{t,r} - Z_t^{t,r}] \\ &= \Pi_{C(t,r)}[\nabla_r(\widehat{Y}_t^{t,r} - Y_t^{t,r})\rho(t, r)] \\ &= -\Pi_{C(t,r)}[\nabla_r p(t, r)\rho(t, r)]. \end{aligned}$$

### Remarks:

- complete case:  $\Delta =$  'delta hedge'
- where is the **risk aversion**  $\eta$ ?

## 13 Example: Heating degree days

- common underlying of **weather derivatives**
- $T_i$  = average of the maximum and the minimum temperature on day  $i$  at a specific location
- $HDD_i = \max(0, 18 - T_i)$

Cumulative heating degree days

$$cHDD_t = \sum_{i=1}^{30} HDD_{t-i}$$

Derivatives:

- Option:  $(cHDD - K)^+$
- Swap:  $b(cHDD - K)$

## 13 Example: Heating degree days

*cHDD*:

- statistical analysis shows: cHDDs are log-normally distributed (M. Davis '01)
- *cHDD* can be modeled as a *geometric Brownian motion*

$$dX_t = \mu X_t dt + \nu X_t dW_t$$

(moving average)

Other indices: *cooling degree days*

$$CDD_i = \min(0, 18 - T_i)$$

## 13 Example: Heating degree days

- $R = \text{cHDDs}$  (geometric Brownian Motion)
- $d = 2$
- 1-dim market + index:  $k = m = 1$
- index volatility:  $\rho = \begin{pmatrix} \alpha & 0 \end{pmatrix}$
- price volatility:  $\beta = \begin{pmatrix} \beta_1 & \beta_2 \end{pmatrix}$  with  $\alpha, \beta_1, \beta_2 \in \mathbf{R} \setminus \{0\}$

Then

$$\Delta(t, r) = -\alpha \frac{\partial p(t, r)}{\partial r} \frac{\beta_1}{\beta_1^2 + \beta_2^2}.$$

# 14 Example: Heating degree days; diversification pressure

derivative hedge:

$$\Delta(t, r) = -\alpha \frac{\partial p(t, r)}{\partial r} \frac{\beta_1}{\beta_1^2 + \beta_2^2}.$$

Call option:  $F(R_T) = (R_T - K)^+$

$$\implies \frac{\partial p(t, r)}{\partial r} > 0$$

Comparison of the optimal strategies:

- $\beta_1 \alpha < 0$  (negative correlation)
  - $\implies F(X_T)$  diversifies portfolio  $\implies \Delta > 0$
  - $\implies \hat{\pi} > \pi$
- $\beta_1 \alpha > 0$  (positive correlation)
  - $\implies F(X_T)$  amplifies portfolio  $\implies \Delta < 0$
  - $\implies \hat{\pi} < \pi$