Capital allocation under Fundamental Review of Trading Book

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Fundamental Review of Trading Book (FRTB)

Basel Committee on Banking Supervision

STANDARDS



Minimum capital requirements for market risk

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January 2016

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Basel 2 and 2.5

▶ 10 days P&L of different risk positions are aggregrated

Basel 2 and 2.5

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 Value-at-Risk (VaR) Incentive to take skewed risk, not sub-additive

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 Value-at-Risk (VaR) Incentive to take skewed risk, not sub-additive

 FRTB sets out revised standards for minimum capital requirements for market risk

- Incorporate the risk of market illiquidity
- An Expected Shortfall (ES) measure
- Constrain the capital-reducing effects of hedging

Structure and implementation

- Standardized approach (SA), Internal models approach (IMA)
- QIS shows that capital charge increases 128% in SA and 54% in IMA (average over 44 banks)
- Model approval down to desk level

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Implementation timeline (Picture from EY)



Impact of FRTB

Consulting firm Oliver Wyman estimates that banks need to spend 5 billion to get ready for FRTB

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Capital charge in SA is very expensive.

IMA requires 90 or more times of calculations than the current rule.

Outline

► FRTB ES and its properties

Capital allocation

Two allocation methods under FTRB

Simulation analysis

Risk factor and liquidity horizon bucketing

P&L of a risk position is attributed to

 $\{\mathsf{RF}_i : 1 \le i \le 5\} = \{\mathsf{CM}, \mathsf{CR}, \mathsf{EQ}, \mathsf{FX}, \mathsf{IR}\} \\ \{\mathsf{LH}_j : 1 \le j \le 5\} = \{10, 20, 40, 60, 120\}$

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BCBS (2016) 181(k)

Risk factor category	n
Interest rate: specified currencies - EUR, USD, GBP, AUD, JPY, SEK, CAD and	
domestic currency of a bank	10
Interest rate: - unspecified currencies	20
Interest rate: volatility	60
Interest rate: other types	60
Credit spread: sovereign (IG)	20
Credit spread: sovereign (HY)	40
Credit spread: corporate (IG)	40
Credit spread: corporate (HY)	60
Credit spread: volatility	120

Risk factor category	n
Equity price (small cap): volatility	60
Equity: other types	60
FX rate: specified currency pairs ³⁷	10
FX rate: currency pairs	20
FX: volatility	40
FX: other types	40
Energy and carbon emissions trading price	20
Precious metals and non-ferrous metals price	20
Other commodities price	60

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Risk profile Loss: Negative of P&L

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Consider a portfolio of N risk positions. $1 \le n \le N$

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Liquidity horizon adjusted loss:

$$X_n(i,j) = \sqrt{\frac{\mathsf{LH}_j - \mathsf{LH}_{j-1}}{10}} \sum_{k=j}^5 \tilde{X}_n(i,k), \quad 1 \le i,j \le 5$$

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We record the liquidity horizon bucketing by a 5×5 matrix:

$$X_n = \{X_n(i,j)\}_{1 \le i,j \le 5}$$

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and call the matrix the risk profile of position n.

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The risk profile of a portfolio is

$$X = \sum_{n} X_{n}$$



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FRTB ES

The FRTB expected shortfall for portfolio loss attributed to RF_i is

$$\mathsf{ES}(X(i)) = \sqrt{\sum_{j=1}^{5} \mathsf{ES}(X(i,j))^2},$$

where ES(X(i, j)) is the expected shortfall of X(i, j) calculated at the 97.5% quantile.

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Example: Consider a portfolio with only one risk position whose is loss is concentrated on RF_i with $LH_5 = 120$.

$$\tilde{X}(i,j)=0, \quad j=1,\ldots,4,$$

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Example: Consider a portfolio with only one risk position whose is loss is concentrated on RF_i with $LH_5 = 120$.

$$ilde{X}(i,j) = 0, \quad j = 1, \dots, 4, \quad ilde{X}(i,5) \sim N(0,\sigma^2)$$

Then the ES over 120 days is $\sqrt{120/10} \sigma \text{ES}(N(0,1))$.

On the other hand, $X(i,j) = \sqrt{rac{\mathsf{LH}_j - \mathsf{LH}_{j-1}}{10}} \widetilde{X}(i,5), \ 1 \leq j \leq 5.$ Then

$$\mathsf{ES}(X(i)) = \sqrt{\sum_{j=1}^{5} \frac{\mathsf{LH}_{j} - \mathsf{LH}_{j-1}}{10}} \mathsf{ES}(\tilde{X}(i,5))^{2} = \sqrt{\frac{120}{10}} \sigma \mathsf{ES}(N(0,1)).$$

Stress period scaling

 $ES^{F,C}(X(i))$: current 12-month, full set of risk factors $ES^{R,C}(X(i))$: current 12-month, reduced set of risk factors $ES^{R,S}(X(i))$: stress period, reduced set of risk factors Restriction: $ES^{R,C}(X(i)) \ge 75\% ES^{F,C}(X(i))$.

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FRTB ES capital charge BSBC (2016) 181 (d) :

$$\mathsf{IMCC}(X(i)) = \frac{\mathsf{ES}^{\mathsf{R},\mathsf{S}}(X(i))}{\mathsf{ES}^{\mathsf{R},\mathsf{C}}(X(i))} \mathsf{ES}^{\mathsf{F},\mathsf{C}}(X(i)), \quad 1 \le i \le 5.$$

Capital charge for modellable risk factors

Unconstrained portfolio:

$$X_n(6,j) = \sum_{i=1}^5 X_n(i,j), \quad X(6,j) = \sum_n X_n(6,j).$$

We add $X(6, \dot{)}$ as the 6-th row of 5 \times 5 matrix, and call it extended risk profile.

IMCC(X(6)) is calculated similarly as before.

IMCC: BCBS (2016) 189:

The aggregate capital charge for modellable risk factors is

$$\mathsf{IMCC}(X) = \rho \,\mathsf{IMCC}(X(6)) + (1 - \rho) \sum_{i=1}^{5} \mathsf{IMCC}(X(i)),$$

where $\rho = 0.5$.

Properties of IMCC

Proposition

- (i) (Positive homogeneity) $IMCC(aX) = a IMCC(X), a \ge 0$.
- (ii) (Sub-additivity for ES) If $ES((X + Y)(i, j)) \ge 0$, then

$$ES((X + Y)(i)) \leq ES(X(i)) + ES(Y(i)).$$

(iii) (Sub-additivity for IMCC) If $\frac{ES^{R,S}((X+Y)(i))}{ES^{R,C}((X+Y)(i))} \le \min\left\{\frac{ES^{R,S}(X(i))}{ES^{R,C}(X(i))}, \frac{ES^{R,S}(Y(i))}{ES^{R,C}(Y(i))}\right\},$ and $ES^{F,C}((X+Y)(i,j)) \ge 0$, then $IMCC((X+Y)(i)) \le IMCC(X(i)) + IMCC(Y(i)).$

Profit and sub-additivity

Example:

Consider X and Y concentrating on RF_i and LH_j .

$$\mathbb{P}(X(i,j) = -1) = \mathbb{P}(X(i,j) = 0) = 0.5, \quad (X + Y)(i,j) \equiv -1.$$

Then ES(X(i)) = ES(Y(i)) = 0, but $ES((X + Y)(i, j)) \ge 0$ is violated,

 $\mathsf{ES}((X+Y)(i)) = |\mathsf{ES}((X+Y)(i,j))| = |-1| = 1 > \mathsf{ES}(X(i)) + \mathsf{ES}(Y(i)).$

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We propose to floor each ES(X(i,j)) at zero (not required by FRTB)

$$\mathsf{ES}^+(X(i)) = \sqrt{\sum_{j=1}^5 \mathsf{ES}^+(X(i,j))^2},$$

where $ES^+(X(i,j)) = max\{ES(X(i,j)), 0\}.$

The resulting FRTB ES is sub-additivity and positive homogeneous.

Capital allocation

Consider a portfolio of N risk positions with losses L_1, \ldots, L_N . The total loss is $L = \sum_{n=1}^{N} L_n$.

 ρ is a risk measure.

An allocation is a map $Law(L_1, \ldots, L_N) \rightarrow \mathbb{R}^N$:

$$Law(L_1, \ldots, L_N) \mapsto \rho(L_n | L),$$
 for each n ,

such that

$$\sum_{n=1}^{N} \rho(L_n \mid L) = \rho(L).$$

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Banks need allocations to calculate return on risk-adjusted capital (RORAC):

$$\frac{-\mathbb{E}[L_n]}{\rho(L_n\,|\,L)}.$$

RORAC evaluates the capital efficiency of each position.

Euler allocation principle

Let v_1, \ldots, v_N be a sequence of numbers and $L^v = \sum_{i=1}^N v_n L_n$. Per-unit Euler allocation is

$$\rho(L_n \mid L)(\mathbf{v}) := \frac{\partial}{\partial \mathbf{v}_n} \rho(L^{\mathbf{v}}).$$

Setting all $v_n = 1$, we denote the allocation to L_n as $\rho(L_n | L)$.

if ρ is homogeneous of degree 1, Euler's theorem for homogeneous functions implies

$$\rho(L^{\nu}) = \sum_{n} \nu_{n} \frac{\partial}{\partial \nu_{n}} \rho(L^{\nu}).$$

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Setting v = 1, we have the full allocation property.

Pros and Cons of Euler allocation

Tasche (1999) shows that

$$\frac{\partial}{\partial v_n} \Big(\frac{-\mathbb{E}[L^v]}{\rho(L^v)} \Big) \left\{ \begin{array}{l} >0, \text{ if } \frac{-\mathbb{E}[L_i]}{\rho(L_i \mid L)(v)} > \frac{-\mathbb{E}[L^v]}{\rho(L^v)} \\ <0, \text{ if } \frac{-\mathbb{E}[L_i]}{\rho(L_i \mid L)(v)} < \frac{-\mathbb{E}[L^v]}{\rho(L^v)} \end{array} \right.$$

Denault (2001) uses corporative game (Shapley (1953) and Aumann-Shapley (74)) to show that the Euler allocation is the "fair" allocation.

There is no sub-portfolio whose total capital charge is less than the sum of the capital allocations of its components.

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There is no sub-portfolio whose total capital charge is less than the sum of the capital allocations of its components.

However,

- Euler allocation is unstable.
- Euler allocation induces large negative allocations.

Two steps allocation for FRTB

Given liquidity horizon adjusted risk profiles $\{X_n\}_{1 \le n \le N}$, Step 1: allocate to each $X_n(i, j)$

 $\rho(X_n(i,j)\,|\,X).$

Step 2: allocate to each $\tilde{X}_n(i, k)$

$$\rho(\tilde{X}_n(i,k)|X_n(i,j)), \quad k \geq j.$$

Then aggregate

$$\rho(\tilde{X}_n(i,k)|X) = \sum_{j=1}^k \rho(X_n(i,k)|X_n(i,j)).$$



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Euler allocation under FRTB

Let
$$v = \{v_1, \dots, v_n\}$$
 be real numbers
Let $X^{v,j}(i) = \sum_n X_n^{v_n,j}(i)$, where
 $X_n^{v_n,j}(i) = (X_n(i,1), \dots, X_n(i,j-1), v_n X_n(i,j), X_n(i,j+1), \dots, X_n(i,5)).$

For each RF_i , we define the Euler allocation for FRTB ES as

$$\mathsf{ES}(X_n(i,j) | X(i)) := \frac{\partial}{\partial v_n} \mathsf{ES}(X^{v,j}(i)) \Big|_{v=1},$$

where v = 1 means all $v_n = 1$.

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Lemma

$$ES(X_n(i,j) | X(i)) = \frac{ES(X(i,j))}{ES(X(i))} \frac{\partial}{\partial v_n} ES(X^v(i,j))\Big|_{v=1},$$

where $X^v(i,j) = \sum_n v_n X_n(i,j).$

Euler allocation of regular ES can be calculated by scenario-extraction (Trashe (1999)):

$$\frac{\partial}{\partial v_n} \mathsf{ES}(X^v(i,j))\Big|_{v=1} = \mathbb{E}[X_n(i,j) | X(i,j) \ge \mathsf{VaR}(X(i,j))].$$

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• If each X(i,j) is floored at zero, then

$$\mathsf{ES}^{+}(X_{n}(i,j) | X(i)) \\ = \begin{cases} \frac{\mathsf{ES}^{+}(X(i,j))}{\mathsf{ES}^{+}(X(i))} \mathbb{E}[X_{n}(i,j) | X(i,j) \ge \mathsf{VaR}(X(i,j))] & \text{if } \mathsf{ES}(X(i,j)) > 0 \\ 0 & \text{otherwise} \end{cases}$$

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▶ If each X(i,j) is floored at zero, then

$$\begin{split} \mathsf{ES}^+(X_n(i,j) \,|\, X(i)) \\ &= \begin{cases} \frac{\mathsf{ES}^+(X(i,j))}{\mathsf{ES}^+(X(i))} \, \mathbb{E}\big[X_n(i,j) \,|\, X(i,j) \geq \mathsf{VaR}(X(i,j))\big] & \text{ if } \mathsf{ES}(X(i,j)) > 0 \\ 0 & \text{ otherwise} \end{cases} \end{split}$$

Euler allocation of IMCC

$$\mathsf{IMCC}^{\mathsf{E}}(X_{n}(i,j) \mid X) := 0.5 \frac{\mathsf{ES}^{\mathsf{R},\mathsf{S}}(X(i))}{\mathsf{ES}^{\mathsf{R},\mathsf{C}}(X(i))} \mathsf{ES}^{\mathsf{F},\mathsf{C}}(X_{n}(i,j) \mid X(i)).$$

It is a full allocation.

Negative allocations

Hedging among different RFs or LHs does not lead to negative allocations.

Example:

Consider two loss Y with RF_i and Z with RF_k, $i \neq k$, $Y, Z \sim N(0, \sigma)$, and Y + Z = 0.

Euler of regular ES:

$$\mathsf{ES}^R(Y|Y+Z) + \mathsf{ES}^R(Z|Y+Z) = \mathsf{ES}^R(Y+Z) = 0.$$

Then one of two allocations must be negative, say $ES^{R}(Y|Y+Z)$.

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Euler of FRTB ES:

Let X be the risk profile containing X and Y. X(i) = Y, X(k) = Z, then

$$\mathsf{ES}(Y|X(i)) = \mathsf{ES}(Y) > 0, \quad \mathsf{ES}(Z|X(k)) = \mathsf{ES}(Z) > 0.$$

Even though Y + Z = 0, IMCC(X) > 0.

Motivated by Li, Naldi, Nisen, and Shi (2016), who combine Shapley and Aumann-Shapley allocations.

LH permutation matrix

$$\mathcal{L} := \begin{bmatrix} 10 & 20 & 40 & 60 & 120 \\ 10 & 20 & 40 & 120 & 60 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 120 & 60 & 40 & 20 & 10 \end{bmatrix}_{5! \times 5}$$

Let $\mathcal{L}^{-1}(r, j)$ be the column of \mathcal{L} in which LH_j locates. e.g. $\mathcal{L}^{-1}(2, 5) = 4$.

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Let $\mathcal{L}^{-1}(r,j)$ be the column of \mathcal{L} in which LH_j locates. e.g. $\mathcal{L}^{-1}(2,5) = 4$. Let $v = \{v_1, \dots, v_N\}$, $X^{v,r,j}(i) = \sum_n X_n^{v,r,j}(i)$,

where $X^{v,r,j}(i)$ is a row depending on when LH_j appears in *r*. For example,

$$X_n^{\nu,2,5}(i) = (X_n(i,1), X_n(i,2), X_n(i,3), 0, \nu_n X_n(i,5)).$$

We define the Constrained Aumann-Shapley allocation (CAS) in the permutation r as

$$\mathsf{CAS}(r, X_n(i, j)) := \int_0^1 \frac{\partial}{\partial v_n} \mathsf{ES}(X^{v, r, j}(i)) \Big|_{v=q} dq,$$

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Lemma

$$CAS(r, X_n(i, j)) = \eta(r, i, j) \frac{\partial}{\partial v_n} ES(X^v(i, j))\Big|_{v=1},$$

where

$$\eta(r,i,j) = \frac{\sqrt{\sum_{1 \le s \le \mathcal{L}^{-1}(r,j)} ES(X(i,\mathcal{L}(r,s)))^2} - \sqrt{\sum_{1 \le s < \mathcal{L}^{-1}(r,j)} ES(X(i,\mathcal{L}(r,s)))^2}}{ES(X(i,j))}$$

- CAS is a scaled version of Euler
- Losses with the same LH need to be added to the same portfolio at the same time, to ensure computational efficiency

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CAS allocation of IMCC

We define the CAS allocation for IMCC as

$$\mathsf{IMCC}^{\mathsf{C}}(X_n(i,j) \mid X) := 0.5 \frac{\mathsf{ES}^{\mathsf{R},\mathsf{S}}(X(i))}{\mathsf{ES}^{\mathsf{R},\mathsf{C}}(X(i))} \frac{1}{5!} \sum_{r=1}^{5!} \mathsf{CAS}^{\mathsf{F},\mathsf{C}}(r, X_n(i,j)).$$

It is a full allocation.

Stress scaling adjustment

In the previous two methods, the $X_n(i,j)$ induced risk contribution is not considered in the stress scaling factor $\frac{\text{ES}^{\text{R},\text{S}}(X(i))}{\text{ES}^{\text{R},\text{C}}(X(i))}$.

We define the Euler allocation with stress scaling adjustment as

$$\mathsf{IMCC}^{\mathsf{E},\mathsf{S}}(X_n(i,j) \mid X(i)) := 0.5 \frac{\partial}{\partial v_n} \Big[\frac{\mathsf{ES}^{\mathsf{R},\mathsf{S}}(X^{v,j}(i))}{\mathsf{ES}^{\mathsf{R},\mathsf{C}}(X^{v,j}(i))} \mathsf{ES}^{\mathsf{F},\mathsf{C}}(X^{v,j}(i)) \Big] \Big|_{v=1}$$

Lemma

$$IMCC^{E,S}(X_{n}(i,j) | X(i)) = 0.5 \Big[\frac{ES^{R,S}(X(i))}{ES^{R,C}(X(i))} ES^{F,C}(X_{n}(i,j) | X(i)) + \frac{ES^{F,C}(X(i))}{ES^{R,C}(X(i))} ES^{R,S}(X_{n}(i,j) | X(i)) - \frac{ES^{R,S}(X(i))ES^{F,C}(X(i))}{ES^{R,C}(X(i))^{2}} ES^{R,C}(X_{n}(i,j) | X(i)) \Big]$$

Simulation analysis 1

 $ilde{X}(i,j)$, normal with mean 0 and annual volatility 30% Risk profiles are simulated for 250 days

Independence among different days

The following correlation structure in the same day:

- 1. Independence
- 2. Strong positive correlation among RFs and LHs
- 3. Strong positive correlation among RFs, independent among LHs
- 4. Independent among RFs, strong positive correlation among LHs

Longer LH leads to larger percentage of allocation



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Simulation analysis 2

Three hedging structures:

- $1.\ \mbox{Strong}\ \mbox{hedging}\ \mbox{between EQ}\ \mbox{and}\ \mbox{IR}$
- 2. Strong hedging between LH_1 and LH_2
- 3. Strong hedging between two risk positions in the same bucket

Simulation analysis 2

Three hedging structures:

- $1. \ {\rm Strong} \ {\rm hedging} \ {\rm between} \ {\rm EQ} \ {\rm and} \ {\rm IR}$
- 2. Strong hedging between LH_1 and LH_2
- 3. Strong hedging between two risk positions in the same bucket



FRTB allocations produce

- no negative allocations for hedging among different bucket
- some negative allocations for hedging in the same bucket, but with smaller magnitude

FRTB allocations are more stable

Histograms and fitted kernel densities for allocations in Case 3:



∽ Q (~ 30 / 32

Simulation analysis 3

Loss in EQ with 40-days LH and CM with 60 days LH have 9 times of volatility in the stress period than the normal period.

Two reduced set of risk factors

- Set A : Include both RFs with large variations
- Set B : Exclude both RFs with large variations

Simulation analysis 3

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	Set A	Set A	Set B	Set B
	(Adj)	(Without adj)	(Adj)	(Without adj)
CM.60	4.00%	2.24%	1.43%	1.43%
EQ.40	5.04%	3.26%	2.11%	2.11%

Table: IMCC(Set A)=11.55 and IMCC(Set B)=3.14

The choice of reduced set of risk factors has large impact on allocations.

Conclusion

Two allocation methods reduce FRTB allocations to Euler allocations

- Computational efficiency
- Easy to adapt to the current system

Simulation analysis shows

- Longer LH leads to more allocation
- Much less negative allocations
- More stable allocations
- Sensitive to the choice of reduced set of risk factors

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Thanks for your attention!