

Implementation of the Sensitivities-Based Method (FRTB) with AAD: Application to an Equity Portfolio

Michele Ercole

University of Verona

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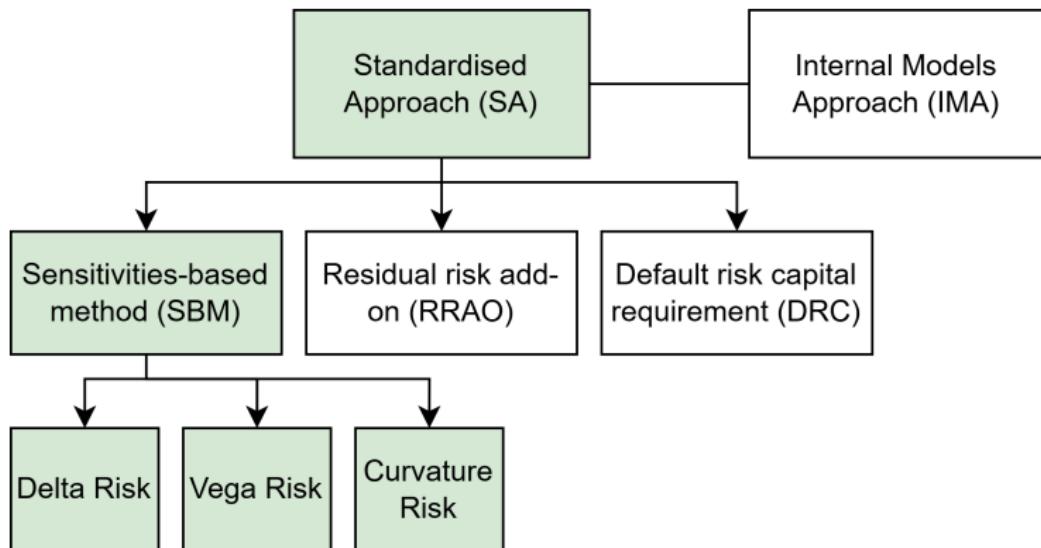
2 Sensitivities-Based Method Workflow

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Context and Objectives

- **Context:** The FRTB regulation, within the sensitivities-based method, involves using finite differences (\rightarrow derivative approximation) to calculate delta and vega.
- **Objective:** Build a Java project to replicate the sensitivities-based method by integrating Automatic Adjoint Differentiation (AAD, \rightarrow exact derivatives) for an equity portfolio.

FRTB – Components



Sample Portfolio

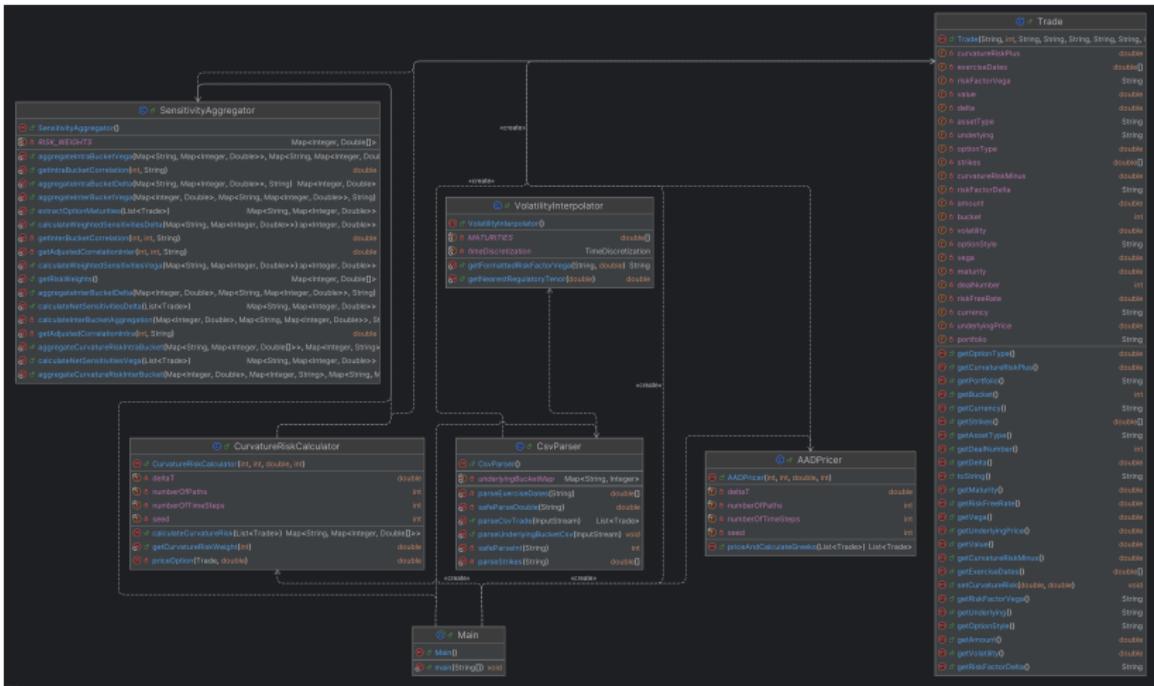
Composition:

- 2 stocks
- 2 European options
- 2 Bermuda options

Portfolio	DealNumber	AssetType	OptionStyle	Underlying	OptionType	Currency	Amount	Volatility	Strikes	UnderlyingPrice	Maturity	ExerciseDates	RiskFreeRate
EQ_PORT	1	Stock		WMT		USD	1			90			
EQ_PORT	2	Stock		JPM		USD	1			100			
EQ_PORT	3	Option	European	AAPL	1	USD	1	0.32	100	100	1.5		0.02
EQ_PORT	4	Option	European	AMZN	1	USD	1	0.25	110	90	2		0.02
EQ_PORT	5	Option	Bermudan	MSFT	1	USD	1	0.3	100;120	100	2	1.0;2.0	0.02
EQ_PORT	6	Option	Bermudan	NFLX	1	USD	1	0.32	100;90	100	2	1.5;2.0	0.02

Input table: excerpt from sample_trades.csv

Project UML Diagram



Sensitivities-Based Method Workflow

- 1 Read input files
 - a) Bucket assignment
 - b) Identification of risk factors
- 2 Compute sensitivities with respect to risk factors
 - a) Net Sensitivities
 - b) Weighted Sensitivities
- 3 Aggregation
 - a) Intra-bucket → medium, high, low correlation
 - b) Inter-bucket → medium, high, low correlation
- 4 Final capital requirement calculation

1. a) Bucket Assignment

Input: sample_trades.csv and underlying_bucket_mapping.csv

Classes: CSVParser, Trade, Main

Each record in sample_trades.csv is assigned a bucket based on the underlying. For example:

Trade: 4 → Underlying: AMZN → Bucket: 5

Underlying	Bucket	Capitalizzazione	Economia	Settore
AAPL	8	Grande	Economia avanzata	Tecnologia
AMZN	5	Grande	Economia avanzata	Beni di consumo e servizi
BABA	1	Grande	Economia emergente	Beni di consumo e servizi

Input table: excerpt from underlying_bucket_mapping.csv

1. b) Identification of Risk Factors

Input: sample_trades.csv

Classes: CSVParser, VolatilityInterpolator, Trade, Main

- **For delta and curvature** → Spot price of the underlying
- **For vega** → Implied volatility mapped to regulatory tenors (0.5Y, 1Y, 3Y, 5Y, 10Y)

DealNumber	Underlying	Bucket	RiskFactorDelta	RiskFactorVega
1	WMT	5	Spot-WMT	null
2	JPM	8	Spot-JPM	null
3	AAPL	8	Spot-AAPL	ImpliedVol-AAPL-1.0Y
4	AMZN	5	Spot-AMZN	ImpliedVol-AMZN-3.0Y
5	MSFT	8	Spot-MSFT	ImpliedVol-MSFT-3.0Y
6	NFLX	5	Spot-NFLX	ImpliedVol-NFLX-3.0Y

Focus: Automatic Differentiation (AD)

Automatic differentiation \leftrightarrow chain rule:

$$\frac{\partial y}{\partial x_m} = \sum_I \frac{\partial y}{\partial x_I} \cdot \frac{\partial x_I}{\partial x_m},$$

Forward

$$\frac{\partial x_m}{\partial x_i} = \sum_{l=1}^{k(m)} \frac{\partial x_{i_l^{(m)}}}{\partial x_i} \cdot \frac{\partial f_m}{\partial x_{i_l^{(m)}}}(x_{i_1^{(m)}}, \dots, x_{i_{k(l)}^{(m)}})$$

Forward mode applies the chain rule starting from the inputs: for each node x_m (from n to N) and for each independent variable x_i , $\frac{\partial x_m}{\partial x_i}$ is computed propagating the information forward along the computational graph.

Backward (AAD)

$$\frac{\partial y}{\partial x_m} = \sum_{l \in I} \frac{\partial y}{\partial x_l} \cdot \frac{\partial f_l}{\partial x_m}(x_{i_1^{(l)}}, \dots, x_{i_{k(l)}^{(l)}})$$

Backward mode applies the chain rule backwards: for each node x_m , derivatives from dependent nodes are aggregated, propagating $\frac{\partial y}{\partial x_m}$ backward. Initial condition: $\frac{\partial y}{\partial y} = 1$

2. Computation of Sensitivities with Respect to Risk Factors

Classes: AADPricer, CurvatureRiskCalculator, Main

Delta

$$s_{i,k}^{FD} = \frac{V_i(1.01 \cdot EQ_k) - V_i(EQ_k)}{0.01}$$

↓

$$s_{i,k}^{AAD} = \frac{\partial V_i}{\partial EQ_k} \cdot EQ_k$$

Vega

$$s_{i,k}^{FD} = \frac{V_i(1.01 \cdot \sigma_k, x, y) - V_i(\sigma_k, x, y)}{0.01}$$

↓

$$s_{i,k}^{AAD} = \frac{\partial V_i}{\partial \sigma_k} \cdot \sigma_k$$

Curvature

$$CVR_k^{\pm} = - \sum_i \left\{ V_i \left(x_k^{RW^{\text{Curvature} \pm}} \right) - V_i(x_k) \mp RW_k^{\text{Curvature}} \cdot s_{ik} \right\}$$

2. Computation of Sensitivities – Example Output

Here is an excerpt of the **output**:

```
==== SENSITIVITY CALCULATION =====
Calculating trade: 4
Underlying AMZN | Bucket: 5 | AssetType: Option | OptionStyle: European
Value AAD: 7,301334 | Value FD: 7,301334 | Value Analytic: 7,261207
Delta AAD: 35,392896 | Delta FD: 35,857758 | Analytic Delta: 35,157002
Vega AAD: 12,281135 | Vega FD: 12,281502 | Analytic Vega: 12,214215
Time AAD: 594,264 ms | Time FD: 897,606 ms | Time Analytic: 0,000 ms

==== CURVATURE RISK CALCULATION =====
RiskFactor: Spot-AAPL |Bucket: 8 | CVR +: -8,885463 | CVR -: -14,019368
RiskFactor: Spot-NFLX |Bucket: 5 | CVR +: -2,724541 | CVR -: -4,186483
RiskFactor: Spot-MSFT |Bucket: 8 | CVR +: -9,869637 | CVR -: -15,638343
RiskFactor: Spot-AMZN |Bucket: 5 | CVR +: -4,085770 | CVR -: -4,227744
```

2. a) Net Sensitivities

Classes: Main, SensitivityAggregator

For delta and vega: Net Sensitivity is calculated for each risk factor:

$$s_k = \sum_i s_{i,k}$$

==== NET SENSITIVITIES DELTA =====

```
RiskFactor: Spot-JPM | Bucket: 8 | Net Delta: 100,000000
RiskFactor: Spot-AAPL | Bucket: 8 | Net Delta: 60,985109
RiskFactor: Spot-NFLX | Bucket: 5 | Net Delta: 71,012533
RiskFactor: Spot-MSFT | Bucket: 8 | Net Delta: 59,138392
RiskFactor: Spot-AMZN | Bucket: 5 | Net Delta: 35,392896
RiskFactor: Spot-WMT | Bucket: 5 | Net Delta: 90,000000
```

==== NET SENSITIVITIES VEGA =====

```
RiskFactor: ImpliedVol-AAPL-1.0Y | Bucket: 8 | Net Vega: 15,226626
RiskFactor: ImpliedVol-MSFT-3.0Y | Bucket: 8 | Net Vega: 14,659538
RiskFactor: ImpliedVol-AMZN-3.0Y | Bucket: 5 | Net Vega: 12,281135
RiskFactor: ImpliedVol-NFLX-3.0Y | Bucket: 5 | Net Vega: 15,626644
```

2. b) Weighted Sensitivities

Classes: Main, SensitivityAggregator

For delta and vega: Weighted Sensitivity is calculated for each risk factor using regulatory weights:

$$WS_k = RW_k \cdot s_k$$

==== WEIGHTED SENSITIVITIES DELTA =====

RiskFactor: Spot-JPM | Bucket: 8 | Weighted Delta: 50,000000

RiskFactor: Spot-AAPL | Bucket: 8 | Weighted Delta: 30,492555

RiskFactor: Spot-NFLX | Bucket: 5 | Weighted Delta: 21,303760

RiskFactor: Spot-MSFT | Bucket: 8 | Weighted Delta: 29,569196

RiskFactor: Spot-AMZN | Bucket: 5 | Weighted Delta: 10,617869

RiskFactor: Spot-WMT | Bucket: 5 | Weighted Delta: 27,000000

==== WEIGHTED SENSITIVITIES VEGA =====

RiskFactor: ImpliedVol-AAPL-1.0Y | Bucket: 8 | Weighted Vega: 11,843270

RiskFactor: ImpliedVol-MSFT-3.0Y | Bucket: 8 | Weighted Vega: 11,402189

RiskFactor: ImpliedVol-AMZN-3.0Y | Bucket: 5 | Weighted Vega: 9,552267

RiskFactor: ImpliedVol-NFLX-3.0Y | Bucket: 5 | Weighted Vega: 12,154403

3. a) Intra-Bucket Aggregation

Classes: Main, SensitivityAggregator

- For delta and vega →

$$K_b = \sqrt{\max(0, \sum_k WS_k^2 + \sum_k \sum_{l \neq k} (\rho_{kl} WS_k WS_l))}$$

- For curvature → $K_b = \max(K_b^+, K_b^-)$, where:

$$K_b^\pm = \sqrt{\max\left(0, \sum_k \max(CVR_k^\pm, 0)^2 + \sum_k \sum_{l \neq k} \rho_{kl} CVR_k^\pm CVR_l^\pm \psi(CVR_k^\pm, CVR_l^\pm)\right)}$$

==== INTRA-BUCKET AGGREGATION ======

--- Delta -----

Bucket: 5 | Delta (M: 39,593061, H: 40,442746, L: 38,724736)

Bucket: 8 | Delta (M: 72,665777, H: 74,326043, L: 70,966680)

--- Vega -----

Bucket: 5 | Vega (M: 16,370731, H: 16,590878, L: 16,147584)

Bucket: 8 | Vega (M: 17,433313, H: 17,672925, L: 17,190362)

--- Curvature -----

Bucket: 5 | Curvature (M: 0,000000, H: 0,000000, L: 0,000000)

Bucket: 8 | Curvature (M: 0,000000, H: 0,000000, L: 0,000000)

3. b) Inter-Bucket Aggregation

Classes: Main, SensitivityAggregator

- **For delta and vega** → $K_{\text{class}} = \sqrt{\sum_b K_b^2 + \sum_b \sum_{c \neq b} \gamma_{bc} S_b S_c}$
- **For curvature** → $k_{\text{class}} = \sqrt{\max(0, \sum_b K_b^2 + \sum_b \sum_{c \neq b} \gamma_{bc} S_b S_c \psi(S_b, S_c))}$

```
==== INTER-BUCKET AGGREGATION =====
--- Delta -----
(M: 90,181645, H: 91,520037, L: 88,823088)
--- Vega -----
(M: 25,754159, H: 26,118930, L: 25,384147)
--- Curvature -----
(M: 0,000000, H: 0,000000, L: 0,000000)
```

4. Final Capital Requirement Calculation

Classes: Main, SensitivityAggregator

The final capital requirement is given by:

$$K_{final} = \max\{K_{medium}, K_{high}, K_{low}\}$$

```
==== FINAL RESULTS =====
Final Capital Requirement Delta: 91,520037
Final Capital Requirement Vega: 26,118930
Final Capital Requirement Curvature: 0,000000
```

Conclusions

The implementation of the project described above led to the following conclusions:

- AAD timings are comparable to FD (the efficiency of the former method becomes more evident as input size increases).
- Greater accuracy of AAD in derivative calculation (taking into account Monte Carlo simulation errors).