

Parallel VTFR

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Team HPCFaI

- Teamlead: Riccardo Gismondi
- Researchers:
 - Kujtim Avdiu
 - Bernhard Bruckner
 - Christian Kremnitzer
 - Peter Kreuzinger
 - Johannes Menzel
- start in June 2007

Our Projects

- PVTFR (Parallelized Valutazione Trattamento Fine Rapporto) - **to be presented at OR 2008 in Augsburg, September 2008**
- Application of Parallel Genetic Algorithms for the Calibration of Financial Models (FIRM presentation in June 2008)
- Model Risk and Calibration Risk
- Just-In-Time Implied Volatility Estimation of Stock Options applying Parallel Computing
- Asset Liability Management, Stochastic Optimization and Parallel Computing
- Optimal Trading Strategies through Genetic Programming
- Parallel SPES (Standard Pricing Estimator System)

today, we will provide an insight into PVTFR

An Introduction to VTFR

- VTFR: Valutazione Trattamento Fine Rapporto
- target of VTFR:
 - to calculate a robust estimation of future pension liability for a dedicated company for all its employees
 - in Italy, as well as in other European countries it is a legal obligation (see IAS 19)

Input Values of VTFR

- **interest rate**
- **inflation rate**
- basic salary growth rate
- salary growth rate dependent on employee qualification level
- **death probabilities**
- probabilities of invalidity
- **probabilities of dismissal**
- hypothesis for payments in advance

- we have to deal with a great number of input factors
- input factors are stochastic
- we do not know the joint distribution of the input factors

⇒ Monte Carlo Approach

Monte Carlo Approach in VTFR

- employees stay in a company and the pension liabilities increase, until an employee
 - dies,
 - becomes invalid,
 - leaves the company (dismissal) or
 - retires
- we estimate the pension liabilities by
 - generating a huge number of employee's life paths
 - estimate future pension liabilities
 - discount future liabilities to current time
 - => taking the average gives us the MC estimator

Specification of Input Parameters

Valutazione TFR

Test firm-19092007

Data acquisizione informazioni: Numero dipendenti:

Ipotesi Economiche

Tasso di inflazione (p.e.: 2,0) Tasso di incremento delle retribuzioni (p.e.: 3,0)

Tasso di rendimento atteso: Linee di carriera:

Ipotesi Demografiche

Probabilità di Morte: Ipotesi di pensionamento: minima
 intermedia
 massima

Probabilità di Inabilità:

Probabilità di dimissioni:
 anno di dimissioni
 anzianità di servizio

Ipotesi richiesta anticipazioni

No. massimo di anticipi (p.e.: 1) Percentuale anticipazione (p.e.: 70,0)

Probabilità di anticipazioni (p.e.: 3,0) Anni minimi tra 2 anticipi (p.e.: 99,0)

Ipotesi calcolo

Numero simulazioni (p.e.: 1000)
 automatico
 manuale

Data di valutazione (dd/mm/yyyy):

Figure: Specification of Input Parameters

Specification of Input Parameters II

Tasso di Rendimento Atteso

Inserisci valore Tasso (p.e. 4,0)

Inserisci valore Indice IBOXX (p.e. 4,0)

Inserisci vettore SWAP (p.e. 4,0 - in ogni campo)

Swap 1 anno:	<input type="text" value="1"/>	Swap 6 anni:	<input type="text" value="3"/>	Swap 12 anni:	<input type="text" value="6"/>
Swap 2 anni:	<input type="text" value="2"/>	Swap 7 anni:	<input type="text" value="4"/>	Swap 15 anni:	<input type="text" value="7"/>
Swap 3 anni:	<input type="text" value="2"/>	Swap 8 anni:	<input type="text" value="4"/>	Swap 20 anni:	<input type="text" value="8"/>
Swap 4 anni:	<input type="text" value="3"/>	Swap 9 anni:	<input type="text" value="5"/>	Swap 25 anni:	<input type="text" value="9"/>
Swap 5 anni:	<input type="text" value="3"/>	Swap 10 anni:	<input type="text" value="5"/>	Swap 30 anni:	<input type="text" value="10"/>

Ripristina Ipotesi Salvate Reset Ipotesi Salva Ipotesi

Indietro

Personalizzazione Probabilità di Dimissioni

	Maschi	Femmine		Maschi	Femmine
1	0,01	0,025	26	0,01	0,025
2	0,01	0,025	27	0,01	0,025
3	0,01	0,025	28	0,01	0,025
4	0,01	0,025	29	0,01	0,025
5	0,01	0,025	30	0,01	0,025
6	0,01	0,025	31	0,01	0,025
7	0,01	0,025	32	0,01	0,025
8	0,01	0,025	33	0,01	0,025
9	0,01	0,025	34	0,01	0,025
10	0,01	0,025	35	0,01	0,025
11	0,01	0,025	36	0,01	0,025
12	0,01	0,025	37	0,01	0,025
13	0,01	0,025	38	0,01	0,025
14	0,01	0,025	39	0,01	0,025
15	0,01	0,025	40	0,01	0,025
16	0,01	0,025	41	0,01	0,025
17	0,01	0,025	42	0,01	0,025
18	0,01	0,025	43	0,01	0,025
19	0,01	0,025	44	0,01	0,025
20	0,01	0,025	45	0,01	0,025
21	0,01	0,025	46	0,01	0,025
22	0,01	0,025	47	0,01	0,025
23	0,01	0,025	48	0,01	0,025
24	0,01	0,025	49	0,01	0,025
25	0,01	0,025	50	0,01	0,025

Dimissioni per Servizio

Valore da replicare (p.e.: 0,01) Applica a maschi Applica a femmine

Ripristina ipotesi salvate Reset Ipotesi Salva Ipotesi Indietro

Figure: Specifying Interest Rate and Probability of Dismissal

Results

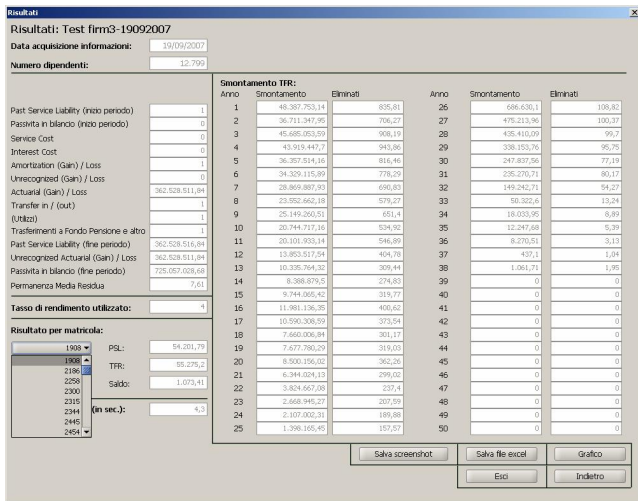


Figure: Results of VTFR Estimation

VTFR technology

JAVA

- primarily the implementation was based on JAVA, since
 - JAVA is fast, platform independent and today the de facto standard in programming

C

- computing intensive applications were sourced out to C, since
 - C is machine oriented and
 - especially for huge companies that have lots of employees, the computational time can be reduced substantially
 - using JNI (JAVA Native Interface)
 - this is where the Cluster@WU and PVTFR comes in

What factors mostly affect the results?

- question:
 - **What are the factors that most influence the results of VTFR?**
- answer:
 - **Interest Rate**
 - **Inflation Rate**
 - **Probabilities of Dismissal**
- also the death probabilities are time variant
- the population more and more grows older
- i.e. e.g. a 30-year old man in 2008 has generally a higher death probability than a 30-year old in 2010
- dynamic death tables considering future death probabilities are included

What factors mostly affect the results?

- include scenario analysis for the 3 risk factors
- this enables a more robust estimation of future pension liabilities
- we have to care about the restrictions of the Cluster@WU
 - since we still have $68 \times 2 = 136$ processors, we can not run as much scenarios as we want in a reasonable time
 - consequently, we have to apply adequate sampling techniques

Sketch of Program Sequence

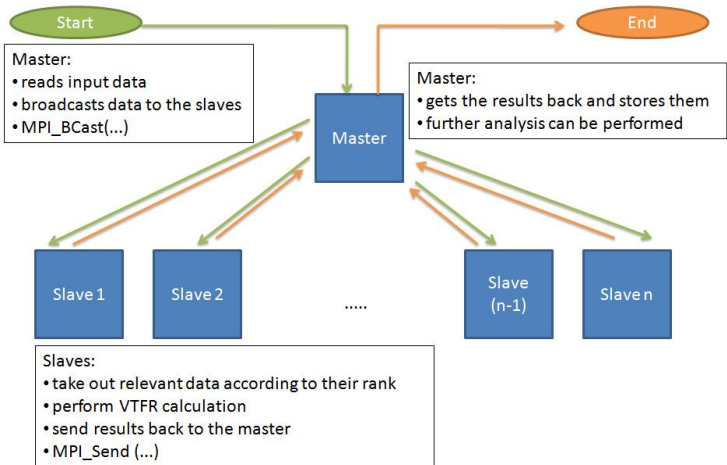


Figure: Sketch of Program Sequence

Short Explanation of Program Sequence

- currently, the application uses a 21 node framework (1 master + 20 slaves)
- the number of scenarios is equivalently shared between the slaves
- e.g. $100 \text{ scenarios} \div 20 \text{ slaves} = 5 \text{ scenarios per node}$
- the VTFR calculation lies completely on the slaves
- no communication between the slaves necessary
- each slave returns its results to the master
- the master bundles the results into one array and saves the results to the output file
- further analysis can be performed

Running VTFR in an MPI Environment

- an excerpt of our mpi.job file:

- `#$ -N vtfr`
`#$ -pe lam 21`
`#$ -q node.q`

```
mpirun -t -np 21 ./mm -t -pdata parameters.txt -pdata company.txt  
-sdata scenariodata.txt -e 12799 -s 100
```

```
lamtrace -mpi trace.ltr
```


Running and Debugging an MPI Task

- debugging parallel applications is a very challenging task
- visualization can help you
- XMPI - <http://www.lam-mpi.org/software/xmpi/>
- bar chart:
 - **RED**: blocked, waiting on communication
 - **YELLOW**: time spent inside an MPI function
 - **GREEN**: node is running
- additional analyse tools:
 - pie charts
 - traffic lights

Graphical Debugging Tools

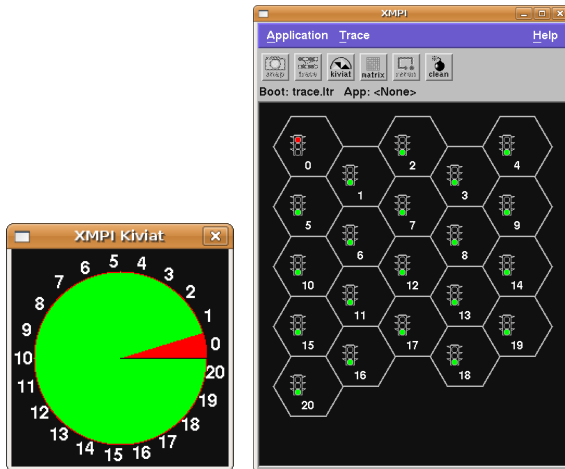


Figure: Pie Charts and Traffic Lights can be very helpful

Visualized Program Sequence observed with MPI Trace

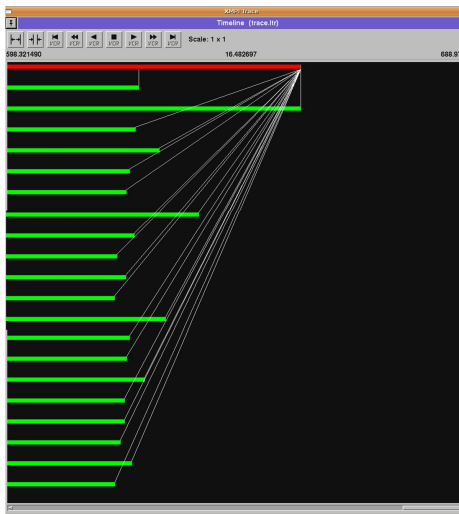


Figure: PVTFR Program Sequence

A theoretical computational example

- given a company (Italian bank) with 12.799 employees
- and let us assume that our cluster offers approximately 100 free nodes

	Computational_Time@Cluster@WU
1 simulation within a single scenario	0.03 sec.
10.000 simulations necessary for robust estimations	$10.000 * 0.03 \text{ sec.}$ $= 300 \text{ sec.} = 5 \text{ min.}$

A theoretical computational example

- we started out with

100 scenarios for	interest rate
100 scenarios for	inflation
100 scenarios for	probabilities of dismissal

- this leads to a computational time of

$$100 \text{ scenarios} \times 5 \text{ min. per scenario} / 100 \text{ nodes} = 5 \text{ min.}$$

- there is no problem running the interest-, inflation-, dismissal-scenarios separately
- even when fewer nodes are in use
(e.g. for 20 available nodes:
 $100 \text{ overall nodes} / 20 \text{ nodes available} \times 5 \text{ min/scenario} = 25 \text{ min}$)

A theoretical computational example

- the combination of all the $100 \times 100 \times 100 = 10^6$ scenarios will be a serious problem
- theoretically:

$$10^6 \text{ scenarios} \times 5 \text{ min/scenario} / 100 \text{ nodes} = \\ 50.000 \text{ min} = 833,33 \text{ hours} = 34,72 \text{ days}$$

- this is definitely too long
- coming from the other side and asking
 - “How long is the application allowed to run?”
and
 - “How many scenarios can be performed within this time?”

A theoretical computational example

- the reduction of e.g. 60 % of the relevant scenarios will lead to
- $40 \times 40 \times 40 = 64 \times 10^3$ scenarios and this will take
- theoretically:

$$64 \times 10^3 \text{ scenarios} \times 5 \text{ min/scenario} / 100 \text{ nodes} = \\ 3.200 \text{ min} = 53,33 \text{ hours} = 2,22 \text{ days}$$

- even not using all the nodes from the Cluster@WU, this will be a reasonable compromise
- consequence:
 - **provided that the underlying problem is designed badly, not even an expensive high performance computer can help you**

First Results of Estimation

- starting with 100 basic scenarios
 - inflation rate: 0.02
 - interest rate: EUR SWAP
 - probabilities of dismissal: observed data from customer
 - number of simulations: 10000

Mean	3.47339229×10^8
Std. Deviation	1.78273464×10^3
Skewness	-0.16631471
Kurtosis	3.52817135
Min.	3.47334258×10^8
Max.	3.47344773×10^8

Interest Rate Scenarios

- generate scenarios by applying a base point shift of ± 50 BP on the EUR SWAP curve

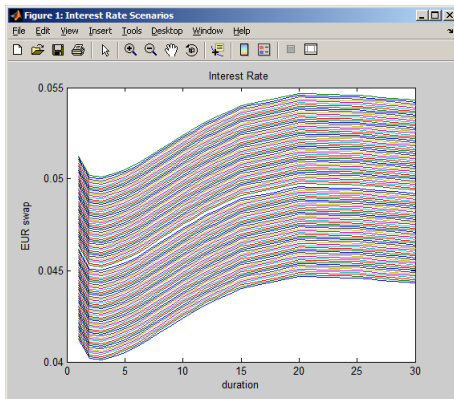


Figure: Generating Interest Rate Scenarios

Results for the Interest Rate Scenarios

- scenario: **Variant Interest Rate**
 - inflation rate: 0.02
 - interest rate: EUR SWAP \pm 50 BP
 - probabilities of dismissal: observed data from customer
 - number of simulations: 10000

Mean	3.47216065×10^8
Std. Deviation	6.34285683×10^6
Skewness	0.03457363
Kurtosis	1.78334055
Min.	3.36624497×10^8
Max.	3.58166489×10^8
Max. - Min.	21.541992×10^6

Impact of Scenario Implementation

- scenario P/L relative to the mean of our base scenario

- **best case:**

absolute difference: 336.62 Mio - 347.34 Mio. = -10.72 Mio.

relative difference: -3.08 %

- **worst case:**

absolute difference: 358.17 Mio - 347.34 Mio. = +10.83 Mio.

relative difference: +3.12 %

Inflation Scenarios

- base on historical CPI (Consumer Price Index) data
- modeling the inflation as Geometric Brownian Motion

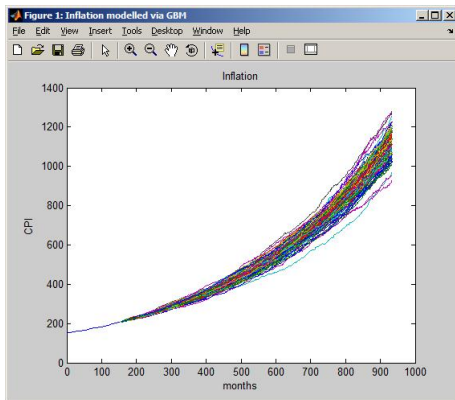


Figure: Inflation Scenarios

Results for the Inflation Scenarios

- scenario: **Variant Inflation Rate**
 - inflation rate: modeled via Geometric Brownian Motion
 - interest rate: EUR SWAP
 - probabilities of dismissal: observed data from customer
 - number of simulations: 10000

Mean	3.56176609×10^8
Std. Deviation	3.53186509×10^6
Skewness	-0.04253140
Kurtosis	2.47621072
Min.	3.46543358×10^8
Max.	3.63930861×10^8
Max. - Min.	17.387503×10^6

Impact of Scenario Implementation

- scenario P/L relative to the mean of our base scenario

- **best case:**

absolute difference: 346.54 Mio - 347.34 Mio. = -0.80 Mio.

relative difference: -0.23 %

- **worst case:**

absolute difference: 358.17 Mio - 347.34 Mio. = +16.59 Mio.

relative difference: +4.78 %

Scenarios for Probabilities of Dismissal

- based on observed data from our customer
- dependent on the number of years an employee is with the company, the probability of dismissal reaches from 3.5 to 0.5 %
- adding some noise $N(0, 0.15)$ to the data generates our scenarios

Probabilities of Dismissal

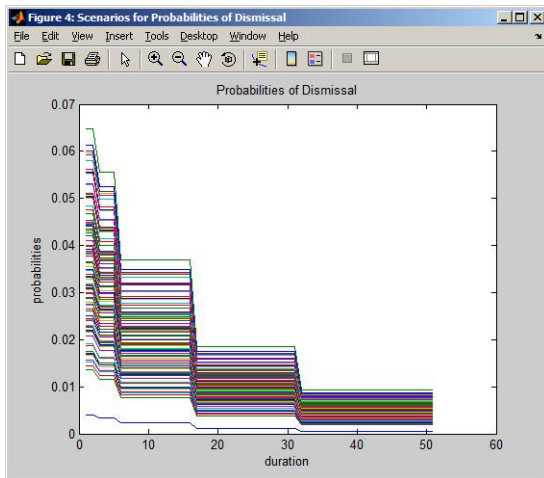


Figure: Scenarios for Probabilities of Dismissal

Basic and Extreme Scenarios

- just to point out the **basic**, **lowest** and **highest** scenario

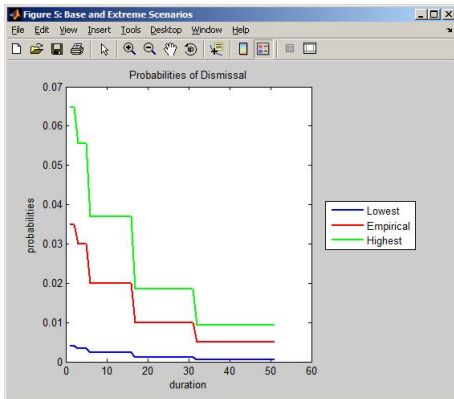


Figure: Showing the Extreme Scenarios Explicitly

Results for the Dismissal Scenarios

- scenario: **Variant Probability of Dismissal**
 - inflation rate: 0.02
 - interest rate: EUR SWAP
 - probabilities of dismissal: observed data + a noisy component $N(0, 0.15)$
 - number of simulations: 10000

Mean	3.46967848×10^8
Std. Deviation	2.76540192×10^6
Skewness	-0.24027199
Kurtosis	3.08889392
Min.	3.40099136×10^8
Max.	3.54701025×10^8
Max. - Min.	14.601889×10^6

Impact of Scenario Implementation

- scenario P/L relative to the mean of our base scenario

- **best case:**

absolute difference: $340.10 \text{ Mio} - 347.34 \text{ Mio.} = -7.24 \text{ Mio.}$

relative difference: -2.08%

- **worst case:**

absolute difference: $354.70 \text{ Mio} - 347.34 \text{ Mio.} = +7.36 \text{ Mio.}$

relative difference: $+2.12 \%$

Next steps

- combining scenarios
- applying adequate sampling techniques on multivariate stochastic processes
- combine sampled scenarios and
- perform a robust estimation of future pension liabilities

Thank you for your attention!