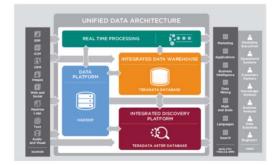


#### Many Data Science Related Initiatives...

# ... and out there, there are many different views So what should we do???!?

#### Teradata

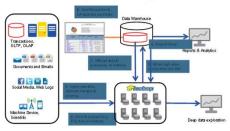


#### Open source big data tooling

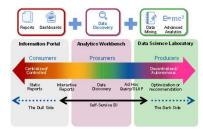


#### MapR Technologies





#### Gartner



## Which elements do we need?

	Research Area Network (RAN), Data Platform +		8
Tooling	<ul> <li>Analytical Workspaces/Datalabs/Data science Toolkit, memory/cpu/storage</li> <li>Cloud deployment; Data(platform) connectivity, other connectivity (open data, etc), quick scaling of datalabs</li> <li>Open source tooling (e.g. R,python, Git, Neo4J, MongoDB, SQLlight, MySQL,)</li> </ul>	People	<ul> <li>Appreciation of the scientific method</li> <li>Knowledge of statistics (descriptive, explorative, predictive, causal,)</li> <li>Knowledge of coding in 'interpreter' languages (Python, R, Julia,) and support (Anaconda, Jupyter Notebooks, Git,)</li> </ul>
Organisation	<ul> <li>Decentral vs. Central</li> <li>Governance (!!!) – data protection, deployment of analysis (KIV→KII)</li> <li>Agile, pilots, data science as a brand</li> <li>FTE's</li> </ul>	Culture	<ul> <li>Informal: knowledge networks, lunches, seminars</li> <li>Creating a community, many already do 'something' with datascience: Get-togethers, what do people need?, datascience 101 sessions, seminar with externals, deep-dive sessies (R, Python, Git, LAMP stack, Neo4J, MySQL, MongoDB, etc), show preliminary - results</li> </ul>

## Current tooling is inadequate

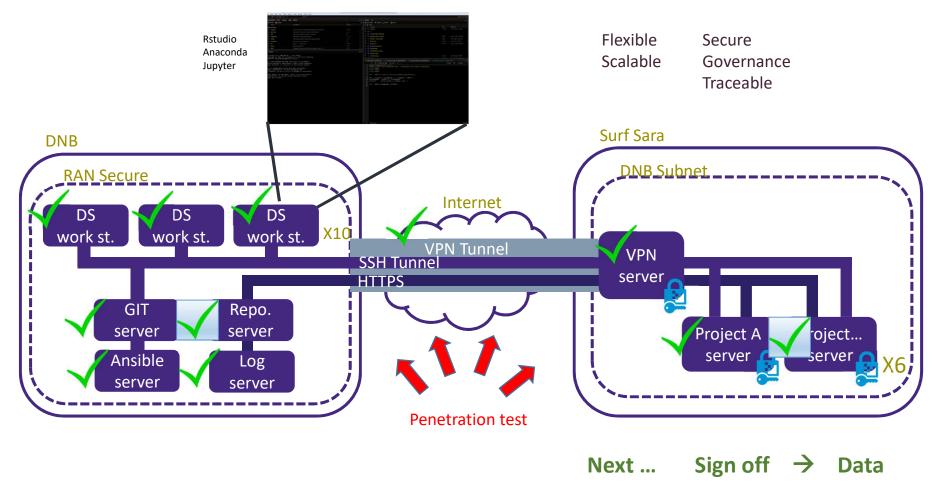
- Data not sufficiently standardized
- Data not sufficiently accessible
  - Application Programming Interfaces (API-layer)
- Research Area Network (RAN) too successful
  - Tedious to move information back and forth
  - Is limited in size and computational speed
  - Fails at a basic level (even simple code loops crash for no reason)

We need a propper analytical workbench ... but how ?? Let's just find out!!!

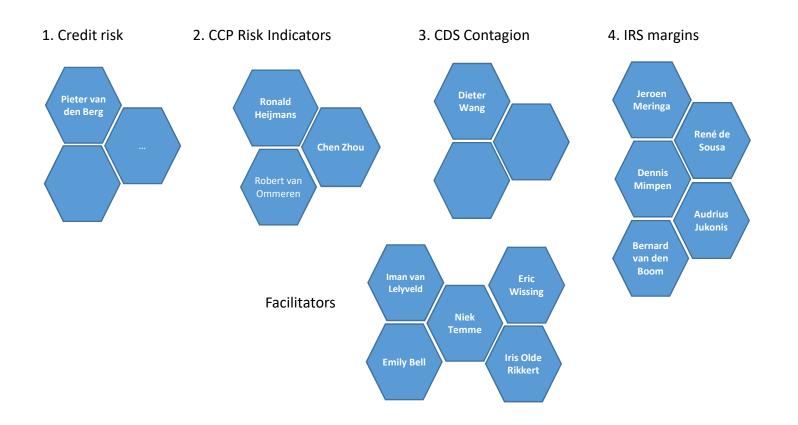
## Today's agenda

- 1. Bring confidential data to the cloud
- 2. Deliver 4 Proofs of Concepts
- 3. Build a Data Science Community

#### Bringing confidential data to the cloud



## 4 Proofs of Concept



#### Proofs of concepts I

#### PoC 1 Credit risk

Importance

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#### Pieter van den Berg

#### PoC 2 EURO CCP

#### **Eric Hogewoning**

#### Importance

- Central Counterparties (CCPs) are becoming more and more important. Post-crisis regulation and technical innovations draw increasing market volume to these central nodes.
  - With their increased systemic importance, it is imperative to understand the risks in CCPS. In particular, how CCP transaction data can support the CCP-oversight function.
  - In order to have a firm base for data driven oversight/supervision we need to develop risk
    indicators from transaction and initial margin data.

#### Analysis

•

- Take the transaction and initial margin data of EuroCCP for one year.
- Explore the data and develop risk indicators with R.
- Develop a method to set a threshold for medium or substantial change in an indicator with R.
- Write reusable code for the different CCP risk indicators and threshold method.

#### Deliverables

- A better understanding of the potential of the data for deriving risk indicators
- A selection of risk indicators linked to the Principles for Financial Market Infrastructures
- Know whether it is possible to derive a adequate threshold for medium or substantial risk of the indicators

#### Credit risk is the largest contributor to overall financial risk and required capital for all of the large banks. Large banks in the NL have permission to calculate minimum capital requirements using internal models.

Differences between risk estimates of different banks are due to different actual risks and risk
management practices (warranted variability) but also due to different default- and loss definitions,
modelling methodology, assumptions and other factors such as regulatory add-ons.

- Due to the variety of modelling practices allowed under Basel and the CRR, there is no clear view on what exactly would constitute an acceptable level of RWA variability.
- Explicitly estimating bank-specific effects could lead to a better understanding of the differences between minimum capital requirements of banks.

#### Analysis

- Get historical loan tape data (100Gbs) in the cloud; at least initially only mortgage data
- Extend existing code base of OSBE/IMK to build simple credit risk / pillar 1 shadow models
- Estimate bank-specific effects
- · Compare typical (scorecard) methods with more sophisticated ('machine learning') methods
- Nice to have:
  - Bayesian methods, with Stan or on a tensorflow backend
  - More advanced modelling techniques (state-space/graphical models)

#### Deliverables

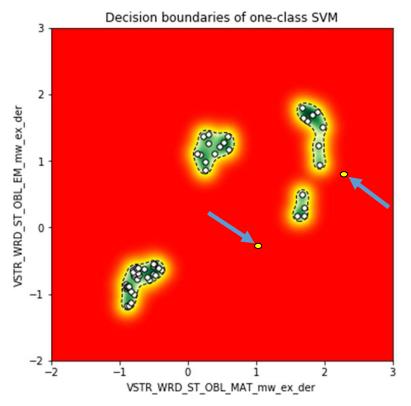
- Replication / extension of current RAN secure RStudio+MonetDB analysis environment in the cloud
- IRB challenger modelling methodology and results
- A better understanding of RWA variability for Dutch banks
- A more generally applicable methodology for risk-corrected RWA benchmarking

#### Proofs of concepts II

PoC 3 Contagion	Dieter Wang	PoC 4 IRS	Iman van Lelyveld	
Importance		Importance		
<ul> <li>Credit Default Swaps (CDS) prices reflect the perce bank. A higher CDS price indicates higher credit ris</li> </ul>		<ul> <li>Interest rate derivative markets financial an non-financial firms.</li> </ul>	are a key component in how interest rates are managed by Daily volume is USD 2.7 trillion.	
<ul> <li>To understand what drives these prices and risks, specific (stock prices, leverage ratio) or economy-v However, these variables explain CDS prices very p</li> </ul>	vide (stock index, bond yields) variables.	Recent regulatory chances force more and more firms to post margin. This reduces counterparty credit risk. But transforms it into liquidity risk.		
<ul> <li>Not only that, the part of credit risk that we cannot there is another hidden variable which failed to include the second second</li></ul>		• Especially for entities that are not used to posting margin (and without access to the discount window == pension funds) this might be an issue		
<ul> <li>In other words: Something is driving credit risk, th</li> </ul>	at we cannot explain!	Moreover, from an FS perspecti	ive, the sinkhole effect of unexpected IM calls might be an issue	
Analysis		Analysis		
<ul> <li>It is not too surprising, that the bank-specific and e risk of banks very well – we didn't account for con</li> </ul>		• Collect the data: take IRS data fi data, join. We do this for a limit	rom 3 TRs (DTCC, REGIS, ICE), apply cleaning steps, add auxiliary ed number of days.	
<ul> <li>Thus, we look at the networks between banks bas bank under stress starts a fire-sale of its assets, thi affected by the sale (namely those with similar hol</li> </ul>	s network will tell us who's likely to be	<ul> <li>Analyse the data in a mix of Stat</li> <li>Write modular code for the deli</li> </ul>	ta, R, Python, Gephi iverables: community detection, pricing, and stress testing	
<ul> <li>We use the resulting portfolio overlap network to spills over from one bank to another.</li> </ul>	capture how much of this stress or credit risk	modules. Deliverables		
<ul> <li>Lastly, we estimate the importance of the network more important during stress times than calm time</li> </ul>		1. Overview of the Dutch IRS markets have? What kind of contracts do these	How many observations/reporters/contracts do we e parties trade? Who is buying/selling risks?	
Deliverables		2 How does TR reporting match up w	ith prudential reporting Match TR with other prudential data:	
<ul> <li>We hope to find proof that the portfolio overlap n risk that eluded the other variables.</li> </ul>	etwork can capture the hidden part of credit	FINREP/COREP, Basel International Da		
<ul> <li>We would like to see how the importance of the n</li> </ul>	etwork varies over time.	<i>3. How to price IRS with TR data?</i> MtM in the data, we want to price ou	Given that we don't trust the rselves. What are the options? How can we estimate all of this?	
Ideally, we can use our resulting model to conduct stress tests in the system. I.e. a dashboard that tells us, how a shock will pass through the banking system.			of margin demands? Given the price model, what model can we at can reliably estimate margin demands across the entire market	
		5. A stress test of the Dutch IRS marke changing some of the parameters affer	et Given data, prices and margin demands: how would ect a) solvency and b) margin required?	

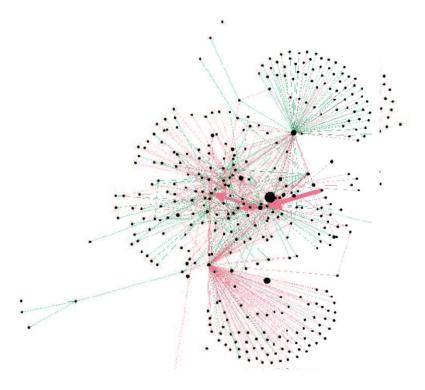
#### Dazzling Data Science ...

#### **Machine Learning**



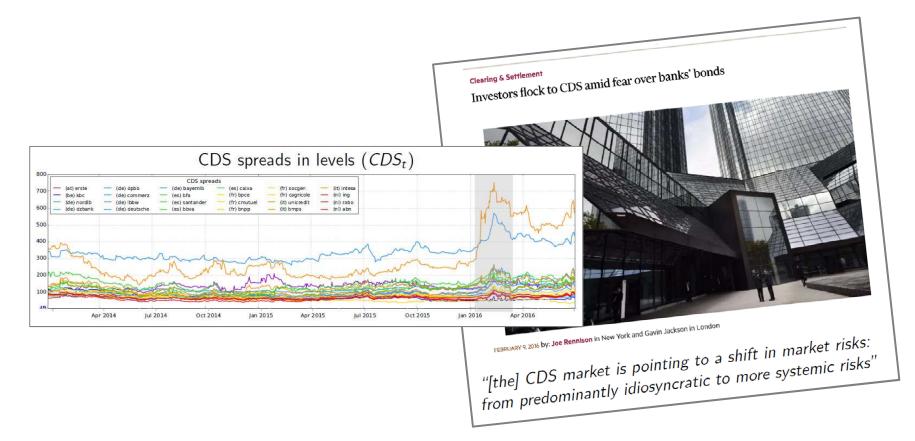
**Detecting outliers in regular reporting** 

#### **Stress testing IRS**



How big will margin stress be if interest rates rise?

### PoC 3 - Contagion in bank CDS



## The Credit Spread is a Puzzle

- Credit Default Swaps (CDS) prices reflect the (perceived) credit risk of the underlying entity (ie. bank)
- To understand what drives these prices and risks, researchers usually analyze the role of *bank-specific* (stock prices, leverage ratio) or *economy-wide* (stock index, bond yields) variables. However, these structural variables explain poorly.
- Not only that, the part of credit risk that we cannot explain is not random. Instead, we know that there is another *hidden* variable which we failed to include.
- Something is driving credit risk, that we cannot explain!

## Main hypothesis

- It is not too surprising, that the bank-specific and economy-wide variables do not explain credit risk of banks very well – we didn't account for overlapping business models!
- Our hypothesis

The credit spread puzzle is a result of the commonality in banking business

# Why is modelling contagion so difficult?

1. Spillovers and feedback loops are hard to grasp: Who will affect whom?

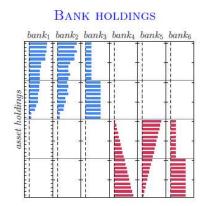
2. Spillovers and feedback loops are nonlinear: Small effects can explode!

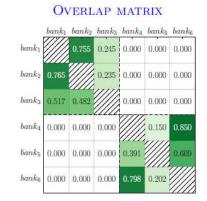
3. Spillovers and feedback loops change over time: Calm times vs Calamity times

## Capturing the underlying network

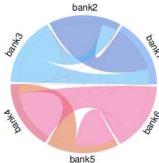
• Banks with similar business models (holdings) likely affect each other in stress times

→ What is the portfolio overlap network?



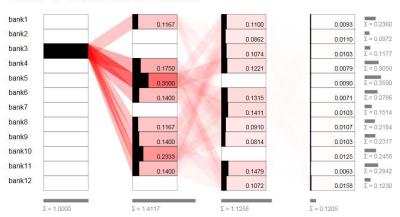






## Network stress-testing

• Once the network is available, we can use our model to conduct stress-tests



CHAIN OF CONTAGION SELECT A BLOCK!

http://dieter.wang/contagionchain

# Why do we need the DataScienceHub?

Massive computing needs

In the 1.5 years of development, the main bottleneck was computation.
Old machine: 4 core @2.4GHz, 16GB RAM
➔ Not only slow but also crashed

Cannot be moved to the cloud

E.g. Amazon Web Services, Microsoft Azure, Google Cloud Computing, etc. because of confidential data.

## Practical uses and deliverables

- Knowledge transfer: High-performance computing
- Knowledge transfer: Confidential data in the cloud
- Dashboard to examine the spillover level in the Eurozone
- Same technology can be applied by other central banks.

## How do we work?

- Organisation
  - Sprint: 3 weeks, 2 week break
  - Physical location
- Agile: goals are clear for each PoC
  - Success factors for PoCs as defined in user stories
- Working towards responsible data use
  - Coding Hygiene document
  - GIT: code repository

PoC 1 Credit risk	Pieter van den Berg	PoC 2 EURO CCP	Ronald Heijmans
• Opzetten infrastrubij SuSa te kunne	uctuur om bestaande functionalite n draaien	<ul> <li>Data geladen         <ul> <li>2 miljard transacties</li> <li>Nog enkele kleine technis</li> </ul> </li> <li>Robert van Ommeren (sta         <ul> <li>Onderzoek naar crowded</li> </ul> </li> </ul>	giair) is begonnen
PoC 3 Contagion	Dieter Wang	• PoC 4 IRS	Iman van Lelyveld
<ul> <li>Presentaties         <ul> <li>EUBA</li> <li>World Bank, Fed Bo</li> <li>FSB Case study</li> </ul> </li> </ul>	oard, IMF, OFR	Prijsmodel werkt:     Aansluiting met EINBEP n	ag in onderzoek

- Aansluiting met FINREP nog in onderzoek ٠
- Presentatie methodologie bij ECB en ESRB
  - Project met ECB opgestart
- Verdere versterking van het team in bespreking
  - FM en TP\_ECFRP

## Community

- Python & R lunches
  - Purpose: get to know each other, exchange ideas
  - Frequency: every 6 weeks
  - Big succes: 30 people on average
- Training
  - Overview of training possibilities
  - Open source (Coursera) complemented with bespoke training

#### Questions?

## **Our Projects**

- 1. Credit Risk
- 2. <u>CCP Risk Indicators</u>
- 3. CDS Contagion
- 4. IRS Margin Requirements
- 5. <u>Tekst Analyse Tool Toezicht</u>

## What will we deliver to you?

#### • Deliverables as described in each PoC

- A presentation at the end of each Sprint
- Report / dashboard at the last
- Note: this is PoC, not an implemented product
- Training
  - Applying Python, R and GIT
  - Requirements for better analytical workspace

## Credit Risk: Importance

- Credit risk is the largest contributor to overal financial risk and required capital for all of the large banks. Large banks in the NL have permission to calculate minimum capital requirements using internal models.
- Differences between risk estimates of different banks are due to different actual risks and risk management practices (warranted variability) but also due to different default- and loss definitions, modelling methodology, assumptions and other factors such as regulatory add-ons.
- Due to the variety of modelling practices allowed under Basel and the CRR, there is no clear view on what exactly would constitute an acceptable level of RWA variability.
- Explicitly estimating bank-specific effects could lead to a better understanding of the differences between minimum capital requirements of banks.

## Credit Risk: Analysis

- Get historical loan tape data (100Gbs) in the cloud; at least initially only mortgage data
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- Compare typical (scorecard) methods with more sophisticated ('machine learning') methods Nice to have:
- Bayesian methods, with Stan or on a tensorflow backend
- More advanced modelling techniques (state-space/graphical models)

# Credit Risk: Results and Deliverables

- Replication / extension of current RAN secure RStudio+MonetDB analysis environment in the cloud
- IRB challenger modelling methodology and results
- A better understanding of RWA variability for Dutch banks
- A more generally applicable methodology for riskcorrected RWA benchmarking

## CCP Risk Indicators: Importance

- Central Counterparties (CCPs) are becoming more and more important. Post-crisis regulation and technical innovations draw increasing market volume to these central nodes.
- With their increased systemic importance, it is imperative to understand the risks in CCPS. In particular, how CCP transaction data can support the CCP-oversight function.
- In order to have a firm base for data driven oversight/supervision we need to develop risk indicators from transaction and initial margin data.

## CCP Risk Indicators: Analysis

- Take the transaction and initial margin data of EuroCCP for one year.
- Explore the data and develop risk indicators with R.
- Develop a method to set a threshold for medium or substantial change in an indicator with R.
- Write reusable code for the different CCP risk indicators and threshold method.

# CCP Risk Indicators: Results and Deliverables

- A better understanding of the potential of the data for deriving risk indicators
- A selection of risk indicators linked to the Principles for Financial Market Infrastructures
- Know whether it is possible to derive a adequate threshold for medium or substantial risk of the indicators

## CDS Contagion: Importance

- Credit Default Swaps (CDS) prices reflect the perceived credit risk of the underlying entity, e.g. a bank. A higher CDS price indicates higher credit risk.
- To understand what drives these prices and risks, researchers usually analyze the role of bank-specific (stock prices, leverage ratio) or economy-wide (stock index, bond yields) variables. However, these variables explain CDS prices very poorly.
- Not only that, the part of credit risk that we cannot explain is not random. Instead, we know that there is another *hidden* variable which failed to include.
- In other words: Something is driving credit risk, that we cannot explain!

## CDS Contagion: Analysis

- It is not too surprising, that the bank-specific and economywide variables do not explain credit risk of banks very well – we didn't account for *contagion*!
- Thus, we look at the networks between banks based on asset holdings similarity. Why? In case a bank under stress starts a fire-sale of its assets, this network will tell us who's likely to be affected by the sale (namely those with similar holdings).
- We use the resulting portfolio overlap network to capture how much of this stress or credit risk spills over from one bank to another.
- Lastly, we estimate the importance of the network over time, because contagion is likely to be more important during stress times than calm times.

# CDS Contagion: Results and Deliverables

- We hope to find proof that the portfolio overlap network can capture the hidden part of credit risk that eluded the other variables.
- We would like to see how the importance of the network varies over time.
- Ideally, we can use our resulting model to conduct stress tests in the system. I.e. a dashboard that tells us, how a shock will pass through the banking system.

## **IRS Margin Requirements**

#### Importance

- Interest rate derivative markets are a key component in how interest rates are managed by financial an non-financial firms. Daily volume is USD 2.7 trillion.
- Recent regulatory chances force more and more firms to post margin. This reduces counterparty credit risk. But transforms it into liquidity risk.
- Especially for entities that are not used to posting margin (and without access to the discount window == pension funds) this might be an issue
- Moreover, from an FS perspective, the sinkhole effect of unexpected IM calls might be an issue

#### Analysis

- Collect the data: take IRS data from 3 TRs (DTCC, REGIS, ICE), apply cleaning steps, add auxilary data, join. We do this for a limited number of days.
- Analyse the data in a mix of Stata, R, Python, Gephi
- Write modular code for the deliverables: community detection, pricing, and stress testing modules.

Results and Deliverables

# IRS Margin Requirements: Analysis

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- Analyse the data in a mix of Stata, R, Python, Gephi
- Write modular code for the deliverables: community detection, pricing, and stress testing modules.

#### I.P.P. va1 check

Lelyveld, I.P.P. van (Iman) (STAT\_OFIS); 17-1-2018

## IRS: Results and Deliverables

Overview of the Dutch iRS markets
 How many observations/reporters/contracts do we have?

What kind of contracts do these parties trade?

who is buying/selling risks?

2. How does TR reporting match up with prudential reporting

Match TR with FINREP/COREP

3. How to price IRS with TR data?

Given that we don't trust the MtM in the data, we want to price ourselves. What are the options? How can we estimate all of this?

4. How to come to sensible estimate of margin demands?

Given the price model, what model can we use to have a simple robust model that can reliably estimate margin demands across the entire market

5. A stress test of the Dutch IRS market

given data, prices and margin demands: how would changing some of the parameters affect a) solvency and b) margin required?