Globally Distributed Secure Data Exchange Fabrics

Cees de Laat

Systems and Networking Laboratory
University of Amsterdam

Contributions from:

Leon Gommans, Paola Grosso, Wouter Los, Yuri Demchenko, Lydia Meijer, Tom van Engers, Sander Klous, Rodney Wilson, Marc Lyonais, Inder Monga, Reggie Cushing, Ameneh Deljoo, Sara Shakeri, Lu Zhang, Joseph Hill, Lukasz Makowski, Ralph Koning, Gleb Polevoy, Tim van Zalingen, and many others!

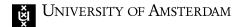




Mission

The Systems and Networking Lab conducts research on leading-edge computer systems of all scales, ranging from global-scale systems and networks to embedded devices.

• Across these multiple scales our particular interest is on extrafunctional properties of systems, such as performance, programmability, productivity, security, trust, sustainability and, last but not least, the societal impact of emerging systemsrelated technologies.

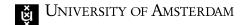




Broad spectrum of research

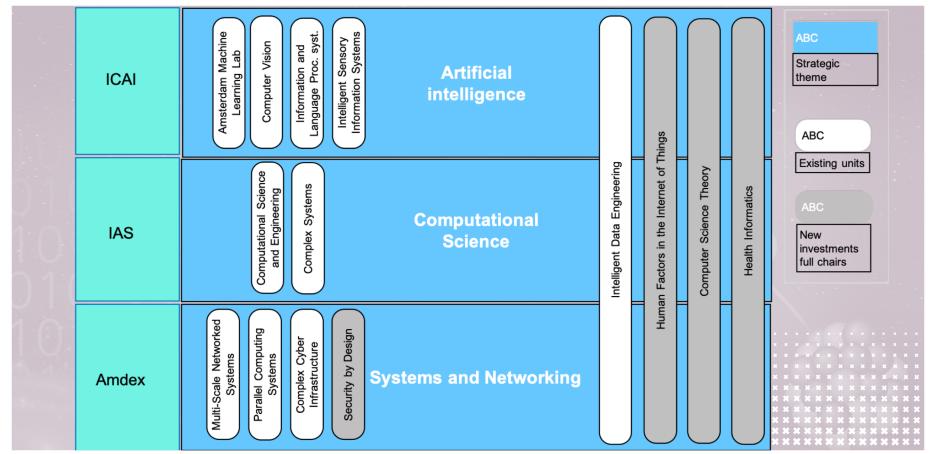
- Advanced networks / Internet architecture
- Network programmabillity / Overlays / Virtualisation
- Authorisation of Internet resources
- Quality of service for apps on Clouds
- Systems: Embedded / real time / parallel / design
- Performance & Compilers & ExaScale
- Safe Secure Data Sharing / Processing
- Data Sovereignty & Normative Agents & Trust → AmDEX
- Well funded accross the themes → ~ 10 Meuro in last 3 years, mostly from NWO / EU and smaller portion from SURF and Industry

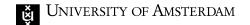




Position in the Instituut

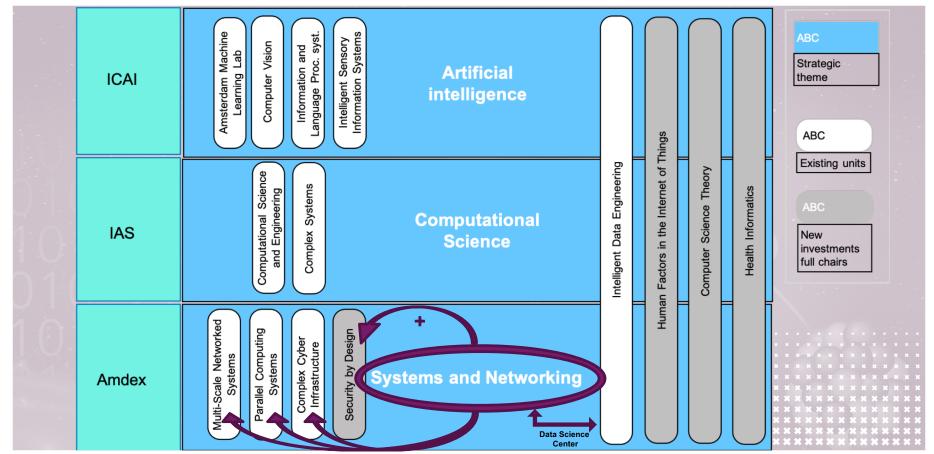


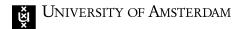




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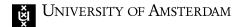




- Group leaders: prof.dr.ir. C. de Laat, dr. Paola Grosso, dr. Andy Pimentel
 - 1 full prof (CdL)
 - 4 associate professors (PG, AV, AP, CG)
 - 2+3 assistant professors (AB, ZZ, 3 open positions)
 - 3 part time professors (PA, PK, TvE)
 - 3 endowed professors (RvN, SK, LG)
 - 1 senior researcher (YD)
 - ~ 23 phd students, ~ 6 postdoc's
 - 2 tech support/scientific programmers
 - ~18 guests
 - Total about 65 people ~ 40 fte
- Yearly turnover ~ 3 MEuro research + similar education

Totals staffing in the SNE Lab







Research based Education

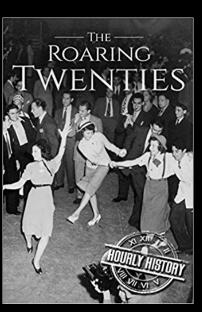
 The SNE laboratory is closely working together with the SNE Master (www.os3.nl), Software Engineering Master (SE) and the Computer Science Master (CS) programs to disseminate knowledge through education.

- Both SNE and SE masters score in the top nationwide
- SNE master Excellent in accreditation and



ICT to support the transformation of Science in the Roaring Twenties





From Wikipedia: The Roaring Twenties refers to the decade of the 1920s in Western society and Western culture. It was a period of economic prosperity with a distinctive cultural edge in the United States and Western Europe, particularly in major cities such as Berlin, Chicago, London, Los Angeles, New York City, Paris, and Sydney. In France, the decade was known as the "années folles" ('crazy years'), emphasizing the era's social, artistic and cultural dynamism. Jazz blossomed, the flapper redefined the modern look for British and American women, and Art Deco peaked....

This period saw the large-scale development and use of automobiles, telephones, movies, radio, and electrical appliances being installed in the lives of thousands of Westerners. Aviation soon became a business. Nations saw rapid industrial and economic growth, accelerated consumer demand, and introduced significantly new changes in lifestyle and culture. The media focused on celebrities, especially sports heroes and movie stars, as cities rooted for their home teams and filled the new palatial cinemas and gigantic sports stadiums. In most major democratic states, women won the right to vote. The right to vote made a huge impact on society.



In most applications, utilization of **Big Data** often needs to be combined with Scalable Computing.

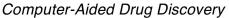


COMPUTING AT DIVERSE SCALES



"BIG" DATA

Enables dynamic data-driven applications



Smart Cities

Disaster Resilience and Response













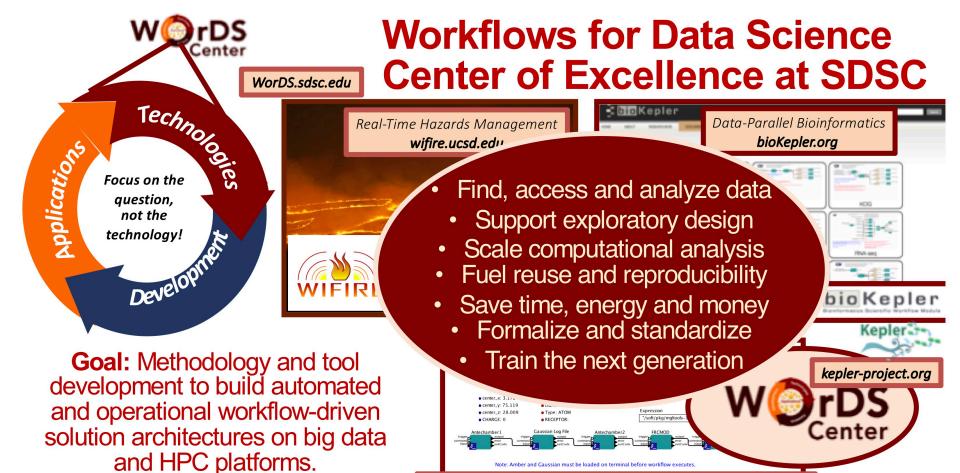


Smart Manufacturing

Personalized Precision Medicine

Smart Grid and Energy Management



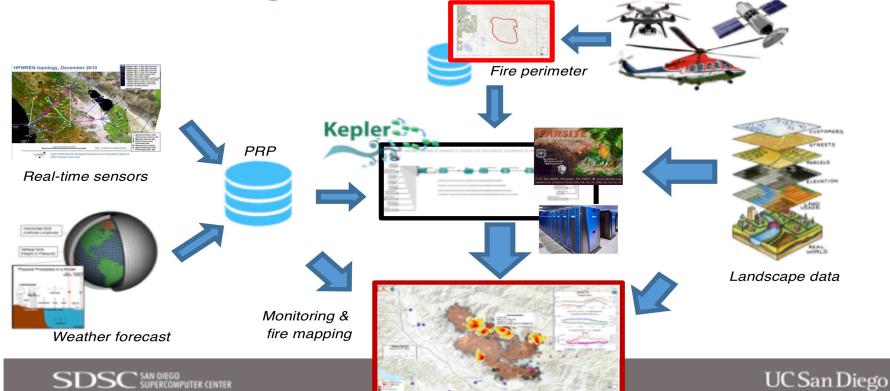




Scalable Automated Molecular Dynamics and Drug Discovery **nbcr.ucsd.edu**

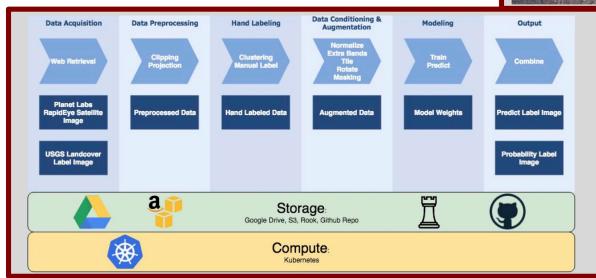
an Diego

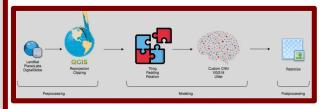
Fire Modeling Workflows in WIFIRE



One Piece of the Puzzle: Vegetation Classification using Satellite Imagery

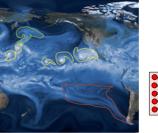


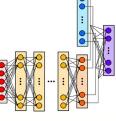


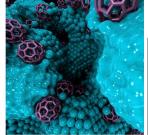








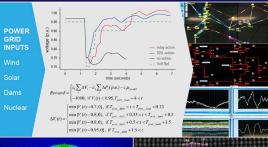


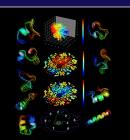


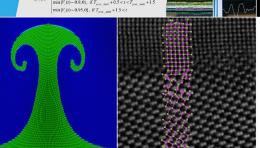
BASIC RESEARCH NEEDS FOR

Scientific Machine Learning

Core Technologies for Artificial Intelligence







Prepared for U.S. Department of Energy Advanced Scientific Computing Research

U.S. DEPARTMENT OF **ENERGY**

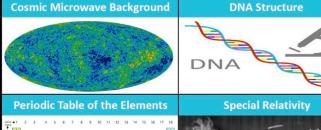
Scientific Machine Learning & Artificial Intelligence

Scientific progress will be driven by

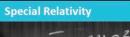
- Massive data: sensors, simulations, networks
- Predictive models and adaptive algorithms
- Heterogeneous high-performance computing

Trend: Human-Al collaborations will transform the way science is done.

EXEMPLARS OF SCIENTIFIC ACHIEVEMENT











Human-AI insights enabled via scientific method, experimentation, & Al reinforcement learning.



Office of Science

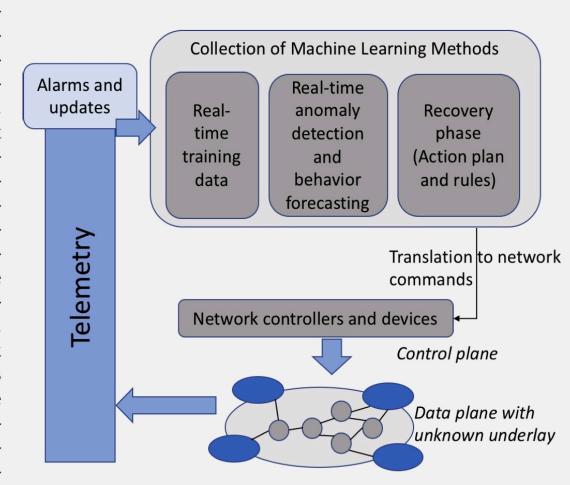
DOE Applied Mathematics Research Program Scientific Machine Learning Workshop (January 2018)

Workshop report:

https://www.osti.gov/biblio/1478744

Example 1: Optimizing Network Traffic with Machine Learning

Exascale and increasingly complex science applications are exponentially raising demands from underlying DOE networks, such as traffic management, operation scale, and reliability constraints. Networks are the backbone to complex science workflows, ensuring data are delivered securely and on time for important computations to happen. To optimize these distributed workflows, networks are required to understand end-toend performance needs in advance and be faster, efficient, and more proactive, anticipating bottlenecks before they happen. However, to manage multiple network paths intelligently, various tasks, such as pre-computation and prediction, must be done in near real time. ML provides a collection of algorithms that can add autonomy and assist in decision making to sup-



Change in computing

- Early days a few big Supercomputers
 - Mostly science domain
- Via grid to commercial cloud
 - AWS, Azure, Google Cloud, IBM, Salesforce
 - The big five: Apple, Alphabet, Microsoft, Facebook and Amazon
 - Computing has transformed into an utility
- Data => Information is the key







Main problem statement

- There is lots of data out there that is not shared (99%)
- FAIR is typically not fair ;-), but limited by policy and/or law
- the A in FAIR is about access, trust is hard to implement across domains
- Organizations that normally compete have to bring data together to achieve a common goal/benefit!
- The shared data may be used for that goal but not for any other!
- Expected use is fine but unexpected use/mission creep...
- Data processed by alien algorithms in foreign data centers... Hmmm...
 - How to organize data processing alliances?
 - How to enforce policy using modern Cyber Infrastructure?
 - How to translate law policy from strategic via tactical to operational level?
 - What are the different fundamental data infrastructure models to consider?



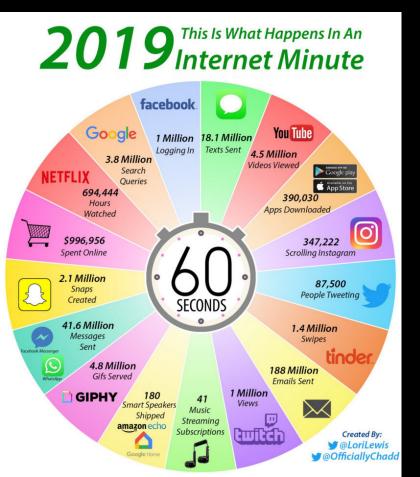
All for one and one for all



- All for one
 - Many infrastructures centered around compute and workflows
- One for all
 - Now we need to get a fluid data layer that frees data to be shared and used by (unforeseen) applications
- Efforts as FAIR and ScienceDMZ / DTN fabrics pave the way to solve the data problem that is also encountered by industry.



Now, how do we get and use data?

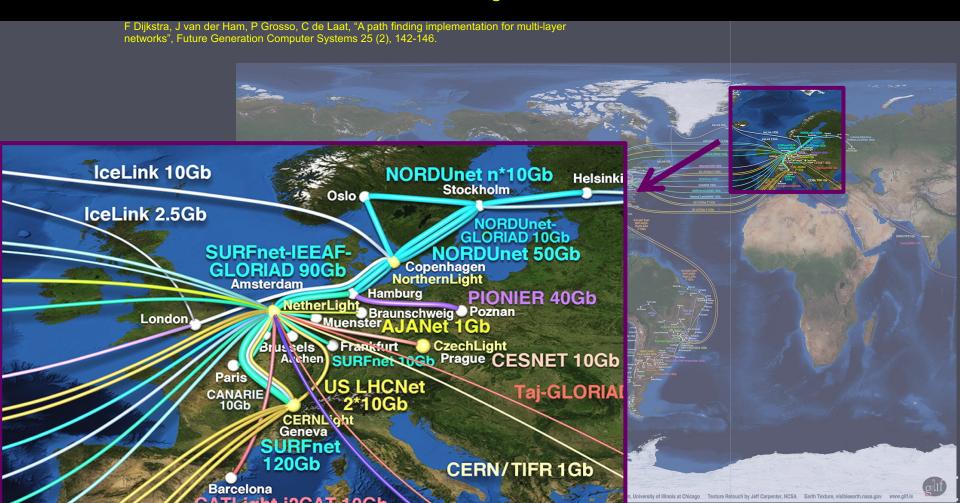


- Move towards streaming
 - Netflix
 - youtube
- Same in science world
 - SKA / LOFAR
 - Light Source
 - Environmental (Marine, Meteorology, ...)
- Data is not always huge
 - Sometimes it is very complex
 - Some example:
 - biodiversity

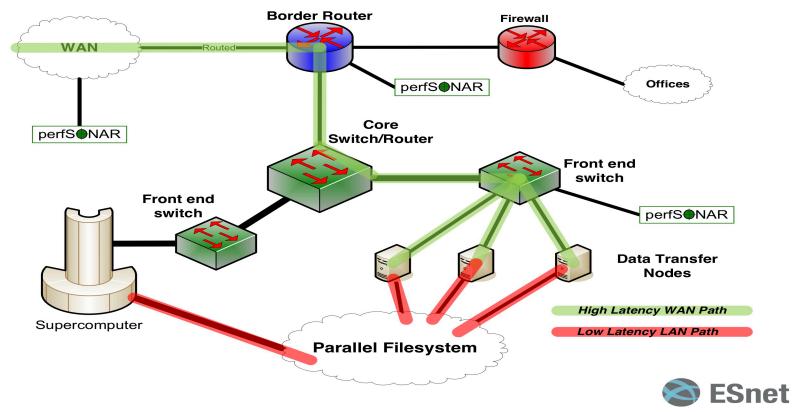
The GLIF – LightPaths around the World



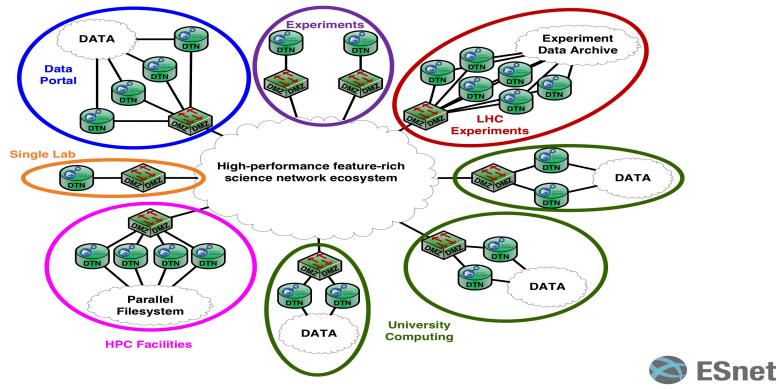
Amsterdam is a major hub in The GLIF



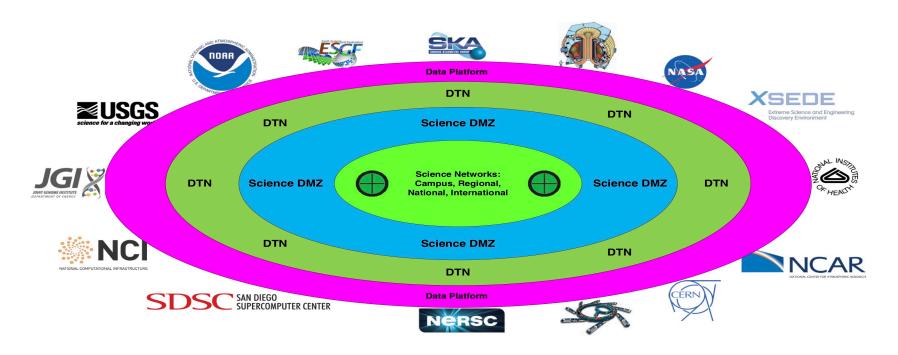
Science DMZ – HPC Center DTN Cluster



Science DMZs for Science Applications

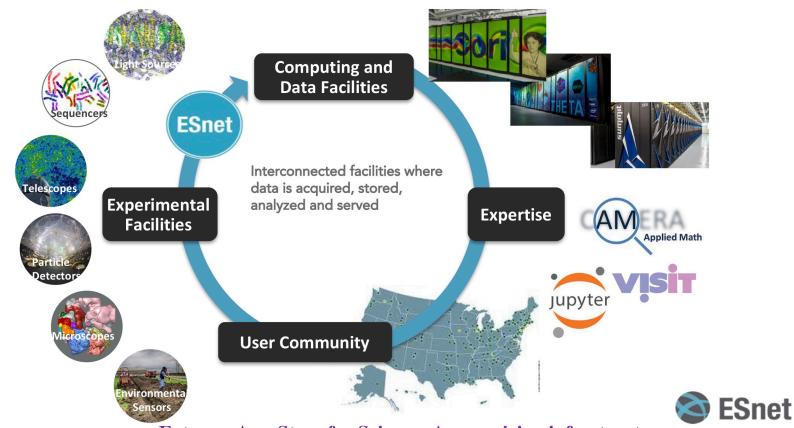


Data Ecosystem – Concentric View



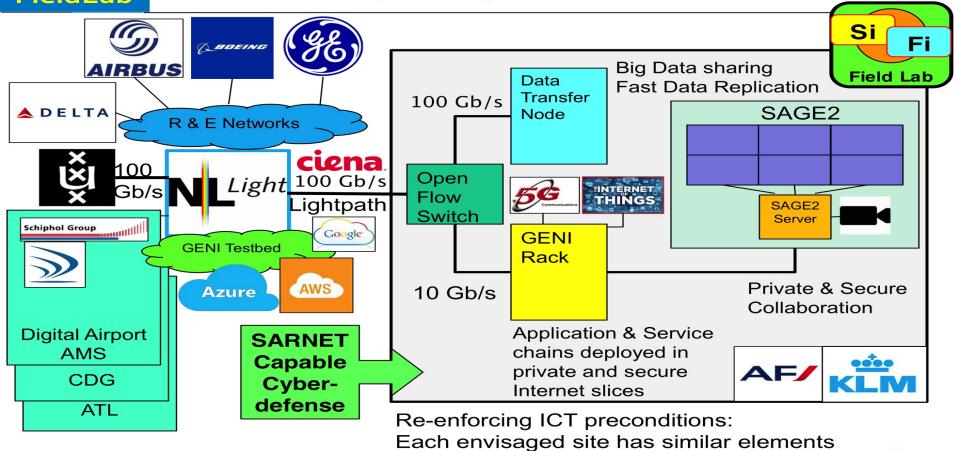


Superfacility Model for Productive, Reproducible Science



AF/KLM FieldLab

Ambition to put capabilities into fieldlab

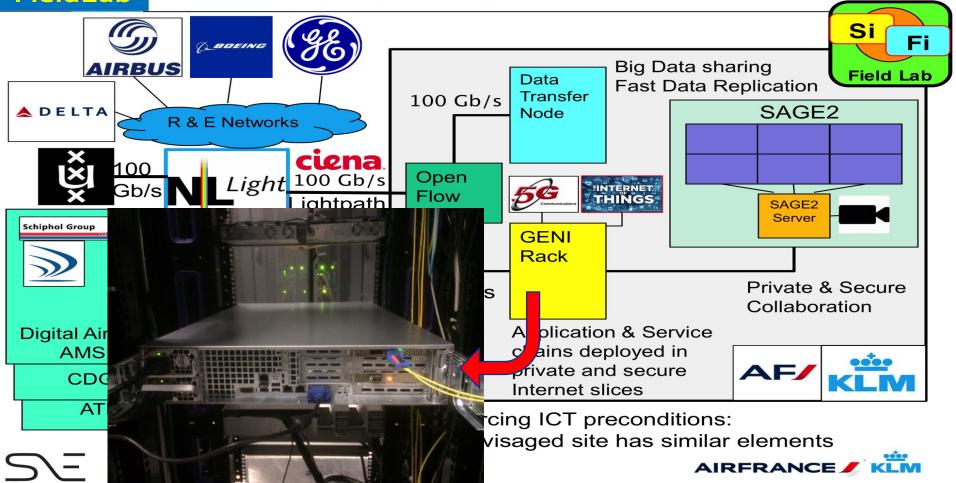


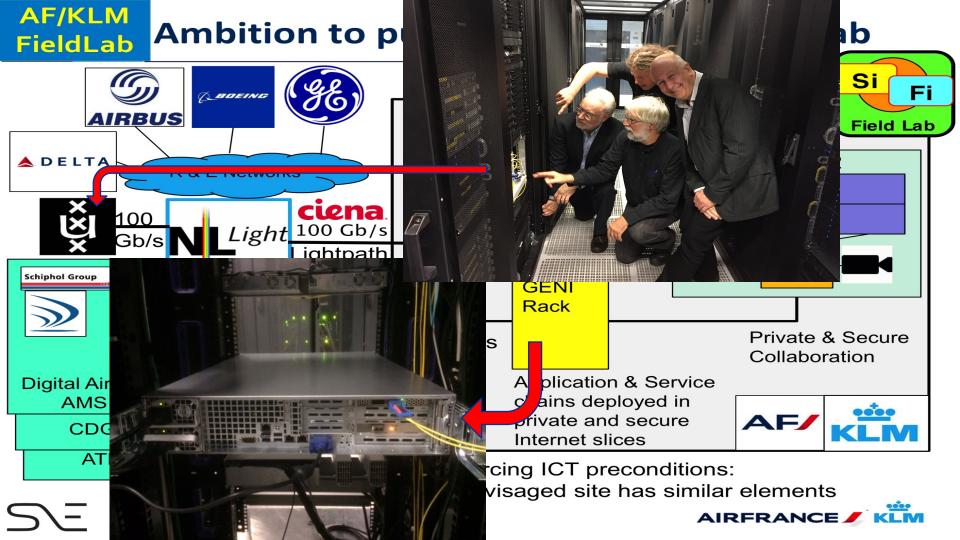




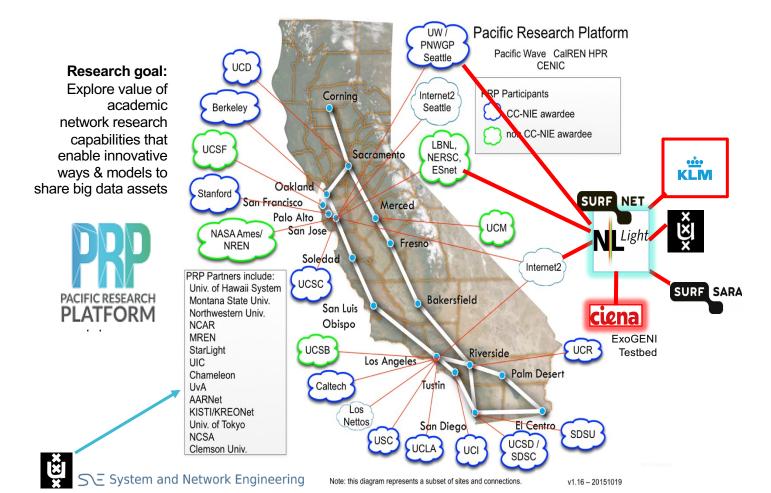
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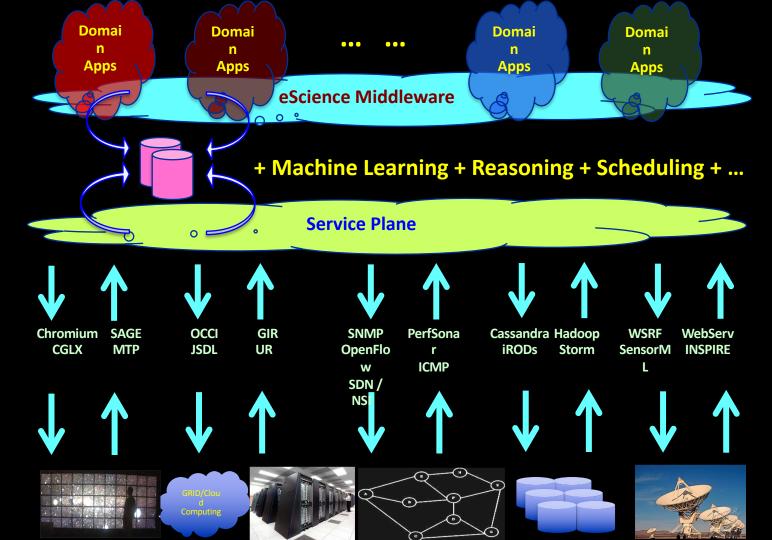
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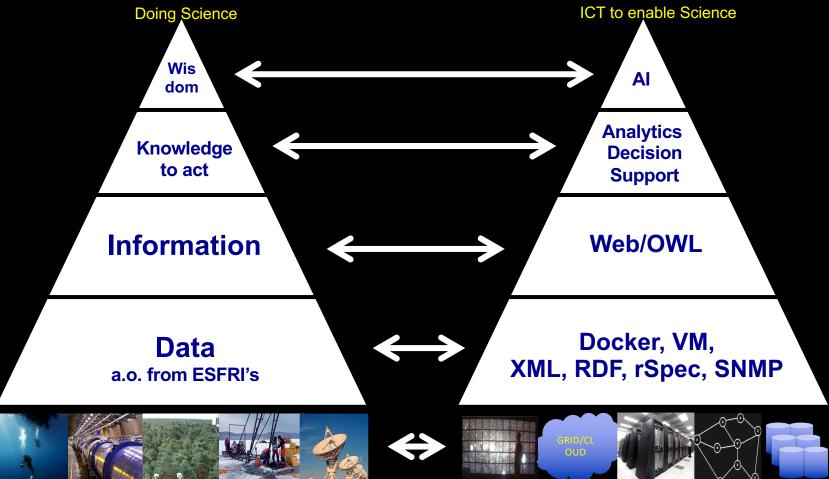


Pacific Research Platform testbed involvement



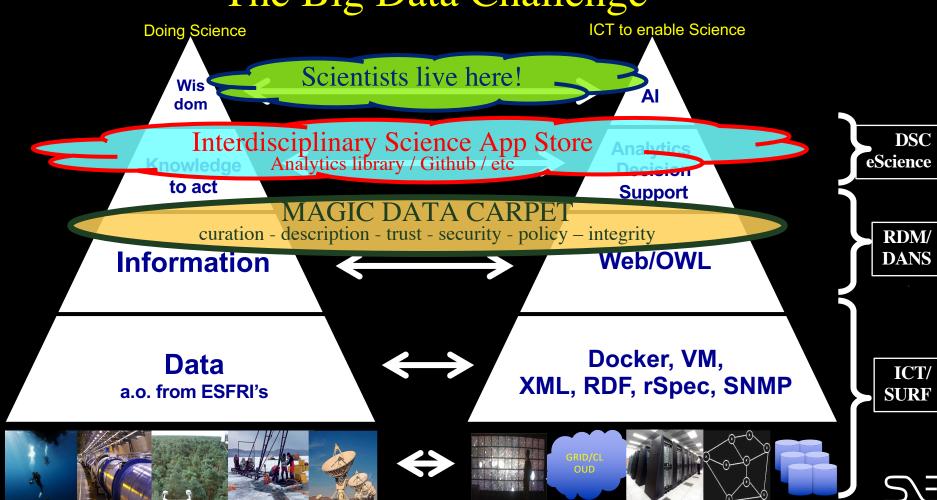


The Big Data Challenge





The Big Data Challenge



SARNET: Security Autonomous Response with programmable NETworks

Marc Lyonnais, Leon Gommans, Rodney Wilson, Lydia Meijer, Frank Fransen Tom van Engers, Paola Grosso, Gauravdeep Shami, Cees de Laat, Ameneh Deljoo, Ralph Koning, Ben de Graaff, Gleb Polevoy, Stojan Travanovski.













Big Data: real time ICT for logistics Data Logistics 4 Logistics Data (dl4ld)

Lydia Meijer (PI), Cees de Laat (Co-PI), Leon Gommans, Tom van Engers, Paola Grosso, Kees Nieuwenhuis.























EPI: Enabling Personalized Interventions

Cees de Laat(PI), Sander Klous (PL), Leon Gommans, Tom van Engers, Paola Grosso, Henri Bal, Anwar Osseyran, Aki Harma, Douwe Biesma, Peter Grünwald, Floortje Scheepers, Gertjan Kaspers.



























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Cyber security program SARNET

Research goal is to obtain the knowledge to create ICT systems that:

Adapt

Serve

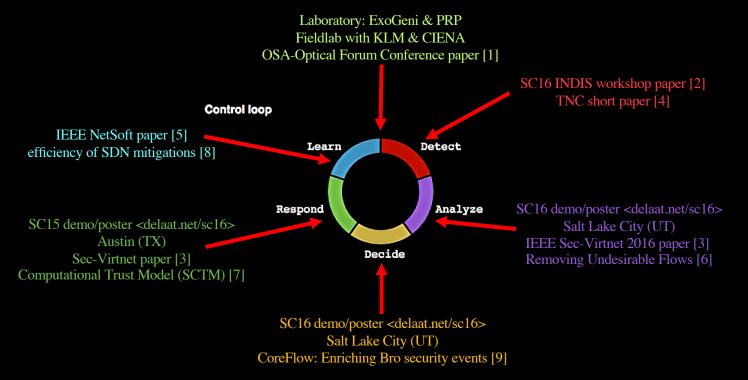
Observe

Observe

- model their state (situation)
- discover by observations and reasoning if and how an attack is developing and calculate the associated risks
- have the knowledge to calculate the effect of counter measures on states and their risks
- choose and execute one.

In short, we research the concept of networked computer infrastructures exhibiting SAR: Security Autonomous Response.

SARNET Publications (subset)



^{1.} Paper: R. Koning, A. Deljoo, S. Trajanovski, B. de Graaff, P. Grosso. L. Gommans, T. van Engers, F. Fransen, R. Meijer, R. Wilson, and C. de Laat, "Enabling E-Science Applications with Dynamic Optical Networks: Secure Autonomous Response Networks", OSA Optical Fiber Communication Conference and Exposition, 19-23 March 2017, Los Angeles, California.

^{2.} Paper: Ralph Koning, Nick Buraglio, Cees de Laat, Paola Grosso, "CoreFlow: Enriching Bro security events using network traffic monitoring data.", Special section on high-performance networking for distributed data-intensive science, SC16", Future Generation Computer Systems,

^{3.} Paper: Ralph Koning, Ben de Graaff, Cees de Laat, Robert Meijer, Paola Grosso, "Analysis of Software Defined Networking defenses against Distributed Denial of Service attacks", The IEEE International Workshop on Security in Virtualized Networks (Sec-VirtNet 2016) at the 2nd IEEE International Conference on Network Softwarization (NetSoft 2016), Seoul Korea, June 10, 2016.

^{4.} Short paper: Nick Buraglio, Ralph Koning, Cees de Laat, Paola Grosso, "Enriching network and security events for event detection", Conference proceedings TNC2017, https://truct/7.geant.org/core/presentation/30

^{5.} Paper: Ralph Koning, Ben de Graaff, Robert Meijer, Cees de Laat, Paola Grosso, "Measuring the effectiveness of SDN mitigations against cyber attacks", IEEE Conference on Network Softwarization (Netsoft 2017 - SNS 2017), Bologna, Italy, July 3-7, 2017.
6. Paper: Gleb Polevoy, Stojan Trajanovski, Paola Grosso and Cees de Laat, "Removing Undesirable Flows by Edge Deletion.", COCOA'2018 conference, December 15 - 17, 2018, Atlanta, Georgia, USA, Springer-Verlag.

^{7.} Paper: Ameneh Deljoo, Tom van Engers, Leon Gommans, Cees de Laat, "Social Computational Trust Model (SCTM): A Framework to Facilitate Selection of Partners". In: Proceedings of 2018 IEEE/ACM Innovating the Network for Data-Intensive Science (INDIS), Dallas, TX, USA, 2018

^{8.} Paper: R. Koning, B. de Graaff, G. Polevoy, R. Meijer, C. de Laat, P. Grosso, "Measuring the efficiency of SDN mitigations against attacks on computer infrastructures", Future Generation Computer Systems 91, 144-156.

9. Ralph Koning, Nick Buraglio, Cees de Laat, Paola Grosso, "CoreFlow: Enriching Bro security events using network traffic monitoring data.", Special section on high performance networking for distributed data-intensive science, SC16", Future Generation Computer Systems

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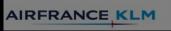
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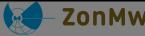








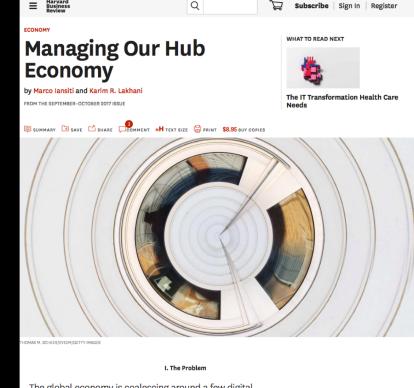


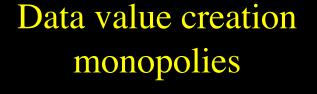






Harvard Business Review







Create an equal playing field



BUSINESS PUBLISHING

The global economy is coalescing around a few digital superpowers. We see unmistakable evidence that a winner-take-all world is emerging in which a small number of "hub firms"—including Alibaba, Alphabet/Google, Amazon, Apple, Baidu, Facebook, Microsoft, and Tencent—occupy central positions. While creating real value for users, these companies are also capturing a disproportionate and expanding share of the value, and that's shaping our collective economic future. The very same technologies that promised to democratize business are now threatening to make it more monopolistic.

Sound Market principles

https://hbr.org/2017/09/managing-our-hub-economy

Data Sharing: Main problem statement

- Organizations that normally compete have to bring data together to achieve a common goal!
- The shared data may be used for that goal but not for any other!
- Data or Algorithms may have to be processed in foreign data centers.
 - How to organize alliances?
 - How to translate from strategic via tactical to operational level?
 - How to enforce policy using modern Cyber Infrastructure?
 - What are the different fundamental data infrastructure models to consider?



Big Data Sharing use cases placed in airline context

Global Scale

National Scale

City / regional Scale

Campus / Enterprise Scale



Cargo Logistics Data
(C1) DaL4LoD
(C2) Secure scalable
policy-enforced
distributed data
Processing
(using blockchain)

NLIP iShare project



Aircraft Component Health Monitoring (Big) Data NWO **CIMPLO project** 4.5 FTE

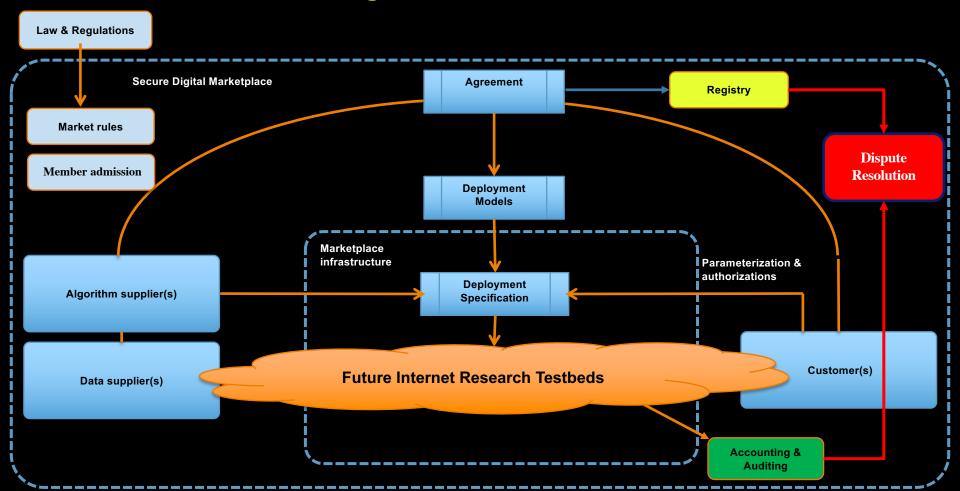


Cybersecurity Big Data NWO COMMIT/ SARNET project 3.5 FTE





Secure Digital Market Place Research



Application Application domain domain Data objects & methods AmDex Data & Algorithms service

FAIR / USE

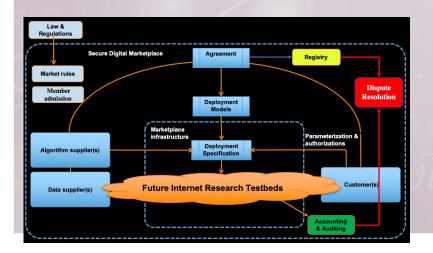
Routers - Internet – ISP's - Cloud AmsIX IP packet service Layer 2 exchange service Ethernet frames

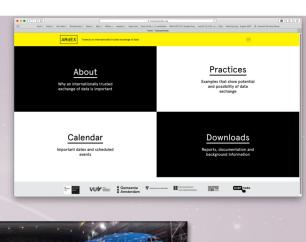
IP / BGP ETH / ST



AMdEX.eu

- Competing organisations, share data for common benefit
- Trust, Risk, data ownership & control
 - Industry: AF-KLM, Health, etc
 - Science: European Open Science Cloud
 - Society: Amsterdam Economic Board







Health: Enabling Personal Interventions



SARNET: Security Autonomous Response with programmable NETworks

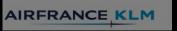
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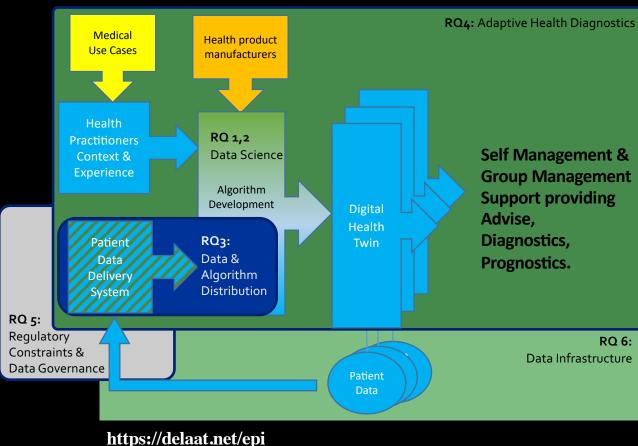








Health Use Case Enabling Personal Interventions



The overall aim of this project is to explore the use and effectiveness of data driven development of scientific algorithms, supporting personalized self- and joint management during medical interventions / treatments.

The key objective is to use data science promoting health practically with data from various sources to formulate lifestyle advice, prevention, diagnostics, and treatment tailored to the individual and to provide personalized effective real-time feedback via a concept referred in this proposal as a digital health twin.

Research questions

- RQ1: Dynamically Analyzing Interventions based on Small Groups: how can we determine, based on as little data as
 possible, whether an intervention does or does not work for a small group or even an individual patient?
- RQ2: Dynamically Personalizing the Group: how can we identify effective intervention strategies and optimize
 personalization strategies applicable for different patient and lifestyle profiles via dynamic (on-line) clustering of
 patients? Can those clusters be adapted as new data about patients and results of interventions come in and as other
 data may be removed or modified?
- RQ3: Data and Algorithm Distribution: what are the consequences of a distributed, multi-platform, multi-domain, multi-data-source big data infrastructure on the machine learning algorithms and what are potential consequences on performance?
- RQ4: Adaptive health diagnosis leading to optimized intervention: how can we enhance self- / joint management by
 dynamically integrating updated models generated from machine learning from various data sources in state of the
 art health support systems that based on personal health records, knowledge of health modes and effective
 interventions?
- RQ5: Regulatory constraints and data governance: how can we create scalable solutions that meet legal requirements
 and consent or medical necessity-based access to data for allowed data processing and preventing breaches of these
 rules by embedded compliance, providing evidence trails and transparency, thus building trust in a sensitive big data
 sharing infrastructure?
- RQ6: Infrastructure: how can the various requirements from the use-cases be implemented using a single functional ICT-infrastructure architecture?

SC16 Demo

DockerMon Sending docker containers with search algorithms to databases all over the world.

http://sc.delaat.net/sc16/index.html#5

Container-based remote data processing

UNIVERSITEIT VAN AMSTERDAM

Łukasz Makowski, Daniel Romão, Cees de Laat, Paola Grosso System and Networking Research Group, University of Amsterdam

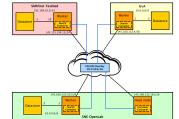


Problem Description

- Scientific datasets are usually made publicly availablebut data cannot always leave the organization premises
- · On-site data processing can be challenging because of incompatibility of systems or lack of manpower
- · Can a container-based system perform remote on-site data processing efficiently?
- · What are the networking issues to solve?



Underlay and Overlay



Main features:

- · Networked containers
- · VXLAN overlay
- · Containers that perform data retrieval and computation
- · Containers built on-demand
- · On-site data processing
- · Distributed data source
- · Multiple sites with datasets

The Game

Our SC16 demo is a gamification of the remote dataset processing architecture.

How many different animal species can you find? You have a fixed budget and each function and processing will cost you money!

In our game you will:

- Select a correlate function to combine the results of the different sites.
- Pick different search functions, represented as tools, to find animals in the remote datasets.
- Build containers with the search and correlate
- · Execute the containers on the sites of your choice.

Will you have the best score?







SC17 Posters and proof of concepts & demo's

http://sc.delaat.net/sc17

Unlocking the Data Economy via Digital Marketplaces

Researching governance and infrastructure patterns in airline context

Use Case: Sharing Aircraft Data to develop a Maintenance Credit System



- A Digital Twin estimates time before maintenance is needed after data is received from a corresponding aircraft system.

Algorithm quality increases when data, owned by different airline operators, can be shared during its development.

Sharing data assets carries risk (e.g. non-compliancy).

Research Question: "Can Digital Marketplace concepts organize trust amongst its stakeholders to enable common benefits no single organization can achieve, whilst observing economic principles?".

Digital Marketplace as a means to organize trusted data asset sharing

A Digital Marketplace is a membership organization identified by a common goal: Share data to enable development of a Maintenance Credit System.

Membership organization is institutionalized to create, implement and enforce membership rules.

Market members create digital agreements.

Agreements are translated into different software defined infrastructures using infrastructure patterns offered by a Digital Marketplace Ecosystem.



Examples of infrastructure patterns offered by a Digital Marketplace







Public cloud model

Container model

Turntable model



















DEMONSTRATION: LIGHT PATHS AND DATA TRANSFER NODES FOR AIRCRAFT MAINTENANCE

Via this open exchange, Data Transfer Nodes (OTNs) of Air France-KLM in the Netherlands and iCAIR – present in Chicago at Statugist – connect to each other using light paths on their links, in this demonstration, users at SCT7 in Deriver will experience the difference in file transfer rates within over their links, or this OTNs.

LISE CASE: AIDCDAFT MAINTENANCE

technical statistics and sensor readings. These data tell pilots and engineers if the aircraft's critical systems are doing their job safety. When data are transferred and analyzed rapidly defects can be solved more quickly, possibly even while the aircraft is waiting at the gate. When receiving the data within minutes, expert engineers in a remote airport can rapidly or by a falling sensor

INTERNET VS LIGHT PATH

Air France-KLM uses a 100 Gbit/s light path and researches its benefits. Using light paths hame. Transferring a tensityte of engine date via the current internet connections would take

AIR FRANCE-KLM CONNECTED TO NETHERLIGHT

\$2,000's European but for international light matter in Amoreviam, \$2,000 remotes the \$00. demonstration, the location is Startisht in Chicago, a hub similar to NetherLight.

DATA TRANSFER NODES

amounts of data. The interconnects between these systems exist of high-capacity dedicated bandwidth, removing network boffienecks within the mesh of global DTNs. To date, DTNs. are present on a small scale, e.g. a couple per continent. By copuing a file from an end user system directly into the nearest DTN, the global DTN system sends the file to the DTN nearest to the final file destination, optimizing the process of high-latency international transfers.

LIVE DEMONSTRATION

Neither system will be optimized for long range transfers, however each will have access between the systems via the DTNs with graphs showing various performance metrics. As long distance transfers can benefit from using nearby DTNs to facilitate the transfer and decreasing file transfer time

DESEADON IN OTHER INDUSTRIES

r research in other industries as well. For example, fundamental research on data transfer protocols suitable for these bandwidths can also help excel diagnosis by doctors when they

How information was perfective 200 G-An-Drance R M

Data Transfer Node (DTN) Workflows

Joseph Hill, Gerben van Malenstein, Cees de Laat, Paola Grosso, Leon Gommans

Why Data Transfer Nodes (DTNs)

- . DTNs can act as an interface to a high performance link
- · Configured to maximize performance for a given workflow
- · Simplifies configuration of client systems
- · Multiple clients may share a DTN
- · DTNs strategically placed to best benefit clients
- · DTNs can be compared to specialized high speed transport systems of the past



Example: Entry Point for High Speed Transport

A typical use case for DTNs is as a high speed file transfer service. A computer system's configuration may allow for the utilization of all available bandwidth in a LAN environment. However, it is often the case that in a WAN environment with high latency or packet loss the same system performs poorly. A DTN could be tuned to maximize performance on a high latency path. It could also use specialized transfer protocols to mitigate high packet loss. The DTN may also have access to an optimized path such as a light path. Files destined for a distant receiver would be first sent to a DTN located on the same LAN as the sender. That DTN would then forward it at high speed to a DTN near the receiver. That DTN would then forward it to the final destination



Example: Storage Access Point

Another possible use case for DTNs is to be used to access distributed data from remote locations. In this scenario a system located at a compute facility requests the data from the local DTN as it is required. That DTN would then transparently retrieve the data from multiple remote sites as needed. In contrast to the first example here block level access is provided by the DTNs. To the system performing the computations the nearby DTN appears to be the actual and only storage system. This hides both the remote and distributed nature of the data. While the compute side DTN may perform some caching, there need not be permanent storage of data at the compute facility.











SC18 – Dallas TX

Training AI/ML models using Digital Data Marketplaces

owned by multiple organizations in a trusted, fair and economic way

The more data - the better: an aircraft maintenance use-case



- AI/ML algorithm based Decision Support Systems create business value by supporting real-time complex decision taking such as predicting the need for gircraft maintenance.
- Algorithm quality increases with the availability of aircraft data.
- Multiple airlines operate the same type of aircraft.
- Research Question: "How can AI/ML algorithm developers be enabled to access additional data from multiple airlines?"

- Approach: Applying Digital Data Marketplace concepts to facilitate trusted big data sharing for a particular purpose.

Digital Data Marketplace enabling data sharing and competition

A Digital Data Marketplace is a membership organization supporting a common goal: e.g. enable data sharing to increase value and competitiveness of AI/ML algorithms.

Membership organization is institutionalized to create, implement and enforce membership rules organizing trust.

Market members arrange digital gareements to exchange data for a particular purpose under specific conditions.

Agreements subsequently drive data science transactions creating processing infrastructures using infrastructure patterns offered by a Data Exchange as Exchange Patterns.



Researching Exchange Patterns to support Digital Data Marketplaces





Research Infrastructure

Research Elements















Dataharbours: computing archetypes for digital marketplaces

Reginald Cushing, Lu Zhang, Paola Grosso, Tim van Zalingen, Joseph Hill, Leon Gommans, Cees de Laat, Vijaay Doraiswamy, Purvish Purohit, Kaladhar Voruganti, Craig Waldrop, Rodney Wilson, Marc Lyonnais

The problem

How can competing parties share compute and data? The architecture of a digital marketplace is an active research field and has many components to it. Here we investigate a federated computing platform which is molded into different archetypes based on trust relationships between organizations.



The components

Consortium: is an initial document which brings together organizations that wish to collaborate. It defines static information such as keys to identify parties.

Infrastructure: A single domain organization infrastructure that securely hosts data, compute containers and, optionally, compute infrastructure. We dub this infrastructure a data harbour. A harbour implements a set of protocols that allows it to interact with other harbours.

which describe how objects (data, compute) can be traded between harbours and who can process data. In its simplest form is a 7-tuple which binds a user,

contracts. The combination of application and contract defines the archetype of the computation i.e. how data and compute are moved to effect computation.

rule enforcemen Contracts: Are a set of rules that are shared amongst participating harbours mpute infrastructure data registry data object, compute container, contract, consortium, harbour, and expiry date. An application: Is a distributed pipeline which can make use of several organizations trust standards Auditor: A trusted entity that collects audit trails for use in litigation of

In action

Federated computing on 3 distributed data harbours. Here we illustrate one archetype where KLM and Airfrance do not trust each other and employ a trusted 3rd party to send the data and compute for

For the scenario to succeed the different harbours need to effect several transactions which are governed by contractual rules The transaction protocol involves first ... identifying both parties are who they say they are through pub/priv key challenges and secondly, that at least

a contract rule is matched to allow the transaction. Important steps of the transactions are audit logged i.e. signed and published to and audit log collector.







SC19 – Denver CO https://sc.delaat.net

A secure network overlay for tracking and enforcement of data transaction rules.

Ralph Koning, Reginald Cushing Lu Zhang, Cees de Laat, Paola Grosso, University of Amsterdam



Competing companies can, together generate value from collaborating on data and compute. Examples include airlines industry, ports, healthcare,

Clearly this poses a challenge of how to echnology. Here we look at one piece of the puzzle i.e. setting up distributed such parties to facilitate the running of

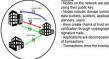


Activation

Multi-domain distributed applications need to share data and compute under

Build an infrastructure that facilitates these applications. Control sharing of data and compute.

Audit activity of the network. Minimize risk of policy/security



Overlay

Nodes on the network are addressed using their public key.

Nodes include: domain controllers data buckets, auditors, application planners, users.

- Keys create chains of trust and verification through cryptographic Applications are decomposed to a set



Securing bucket-to-bucket nunication through transaction Bucket node key address used as Opening connection endpoints on

Overlay allows for a distributed

signature trails and trust chains

attack vectors on data transfers.

Network of auditors provide rubber-stamping of actions/transactions

Control layer enforces security using

. Users continue to use the tools they are familiar with without any need to consider the location of demand. Bucket containers have no the recipient network interface. Interfaces are only created and attached per signed

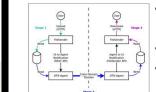
· Data transfers between domains are performed faster with both the sender and the recipient using their preferred data transfer application.

. Within a domain the DTN provides an API that is

utilized by user facing data transfer applications

· There is no change in how data transfers between users within the same domain are performed.







n Stage 1: User Unload

o Stage 2: Cross Domain Transfer

between stages

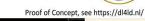
· FileSender need not be run on the



AIR FRANCE KLM



















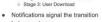


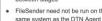
- · A worldwide network of DTNs supports users by allowing long distance transfers to be done by specialized systems on optimized paths.
- Data transfer domains are determined based on geographical region or organization.
- DNS service records are used to find the appropriate DTN for a recipients domain with support for user specific entries
- . Between domains the DTNs provide an API that allows for peer DTNs to choose from a selection of protocols designed for high speed data transfers.





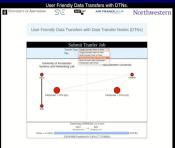














AIR FRANCE KLM

A secure network overlay for monitoring and enforcement of multi-domain applications.

Conclusions, Info, Acknowledgements, Q&A

- Data hindered by risk of unexpected use, lack of trust
- Using market principles, enforcement and determining incentives and value in the data life cycle to make data flow
- More information:
 - http://delaat.net/epi
 - https://www.esciencecenter.nl/project/secconnet
 - https://towardsamdex.org

