

PANEL - CONTENT / PATTERNS 2013

Advances on Information Mining

Samuel Kosolapov, Hans-Werner Sehring, Alfred Zimmermann

Introduction

1. What are the Advances in Information Mining we have to discuss today?

1. Goal of our discussion:

We want to discuss about our experiences and our positions, coming from both academia and practice

Panelists

Self-introduction of panelists with short **information** about each **person**, their **work** and **expectations** to our **discussion**

- **Samuel Kosolapov – Israel**
- **Hans-Wener Sehring – Germany**
- **Alfred Zimmermann – Germany**

Definition of Information Mining

1. **Origin:** Data Mining or Knowledge Discovery from Databases
2. **Orientation:** towards high structured data
3. **Analysis of heterogeneous information sources**
4. **Often** combined with **AI technologies**, like fuzzy logic
5. **Classical methods** from **statistics, decision trees, neural networks**
6. **Challenges:** mining of texts, models, patterns, image & sound data, medical data, spacial data, other temporal data
7. **Information Mining:** Extension of Data Mining to identify understandable patterns in heterogeneous information sources

Tasks of Knowledge Discovery

1. **Deviation Filtering** (Anomaly Detection) of unusual instances
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5. **Regression**: find a function which models the data with least error
6. **Summarization**: compact representation for visualizations or reports
7. **Pattern Mining**: Find sets of instances that occur frequently together

Positions from Panelists

1. **Samuel Kosolapov:**

YourContributionTitle

2. **Hans-Wener Sehring:**

YourContributionTitle

3. **Alfred Zimmermann:**

**Extracting and Integrating Metamodel-Information from Capability
Maps of IT Enterprise Architectures**

Synthesis of Opinions

1. **Questions** from the **audience** and **answers** from **panelists**
2. **What** are our **main conclusions** and **hypotheses** for **future work** in **Information Mining**?
3. **What** are the **big challenges** we have to face?

School of Informatics

Business Informatics

ARL - Architecture Reference Lab



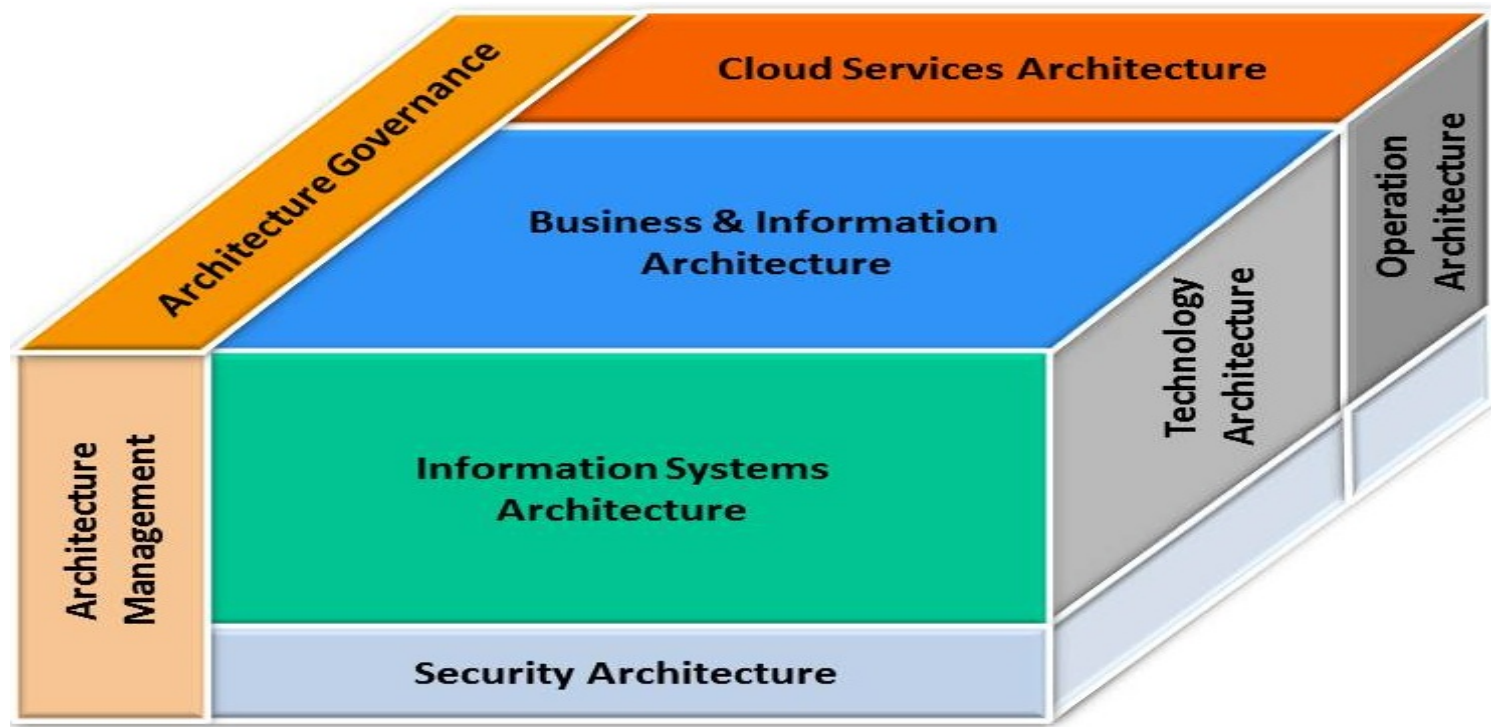
ESAMI – Enterprise Services Architecture Metamodel Integration

Prof. Dr. Alfred Zimmermann

ESARC® - Enterprise Software Architecture Reference Cube

Architecture Capability Diagnostics, Monitoring, and Optimization

- ESARC defines an original holistic classification scheme for cyclic diagnostics and optimization of eight types (viewpoints) of service-oriented enterprise software architectures
- ESARC substantiates the TOGAF standard with other models to provide a useful mapping foundation of a reference architecture, defining main architecture artifacts and their relationships



ESAMI – Enterprise Services Architecture Metamodel Integration

Integration Process

1. Analyze each **Base Model** with **Concept Maps**
2. Extract the **Base Capability Model** from each model:
Capability, Model, Element, [Example]
3. Initialize the **Reference Architecture Model** from Base Capability Models:
Capability, Model, Element
4. Analyze **Correlations** (Concept Mappings) between Base Models and Reference Architecture Model, and optionally conclude **transitive correlations** (Min)
5. Determine **Integration Options** for the resulting **Capability Integration Model**
6. Develop the **Synthesis Metamodel** from **Base Metamodels**
7. Consolidate the **Reference Architecture Model** according to the **Joined Metamodel**, and finally readjust **Correlations** and **Integration Options**
8. Develop **Capability Map** and **Ontology** of the **Reference Architecture Model**
9. Develop **Correspondence Rules** between **Model Elements**
10. Develop **Patterns** for **Architecture Diagnostics** and **Optimization**

Integration Scenario of Business Architectures from ArchiMate® 2.0 and TOGAF 9.1

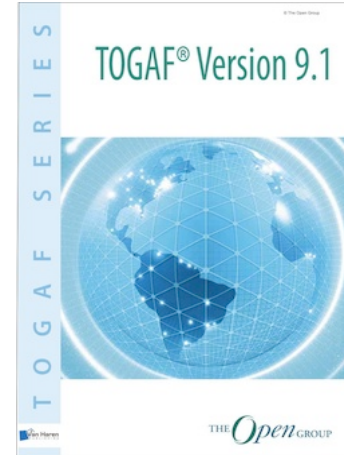


Open Group Standard

**ArchiMate® 2.0
Specification.**

The Open Group
2009-2012,

ISBN 1-937218-00-3



Open Group Standard

TOGAF 9.1.

The Open Group 2011,

ISBN 978-9087536794



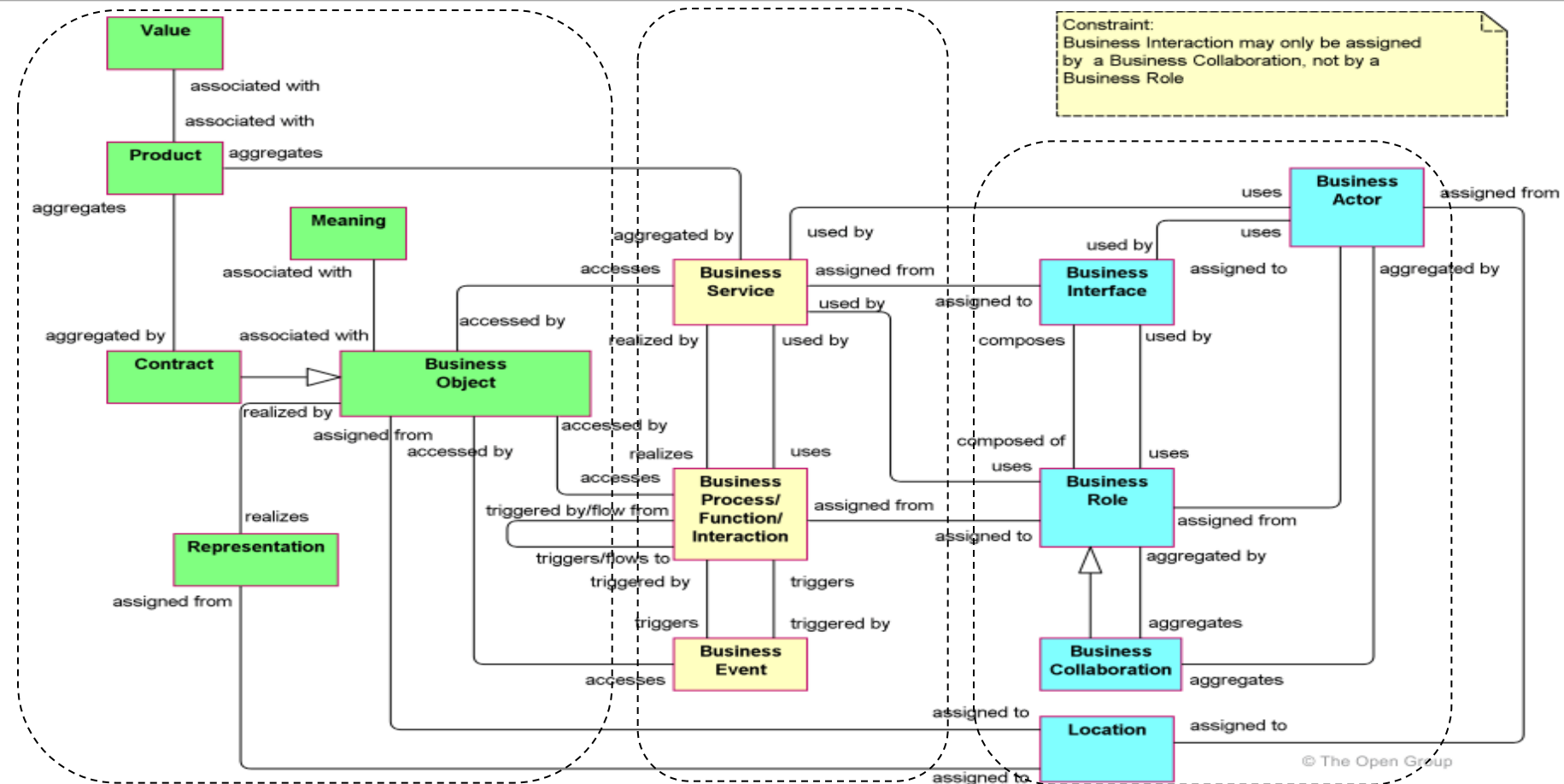
Capability	Model	Element	Example
BusinessActivator (Viewpoint)	<u>BusinessActor</u>	BusinessActor	BusinessActor = Language Insurance
	BusinessRole	BusinessRole	Department BusinessRole =
	Business Collaboration	Process	Travel Insurance Seller
	BusinessInterface	Service	Process = tale out travel insurance
	Location		Service = Offering travel insurance



Capability	Model	Element	Example
Organization (Viewpoint)	<u>Actor</u>	Actor Role	
	Role	OrganizationUnit Function	
	Location	Process	
		BusinessService Event Location	

ArchiMate 2.0

Business Layer Metamodel



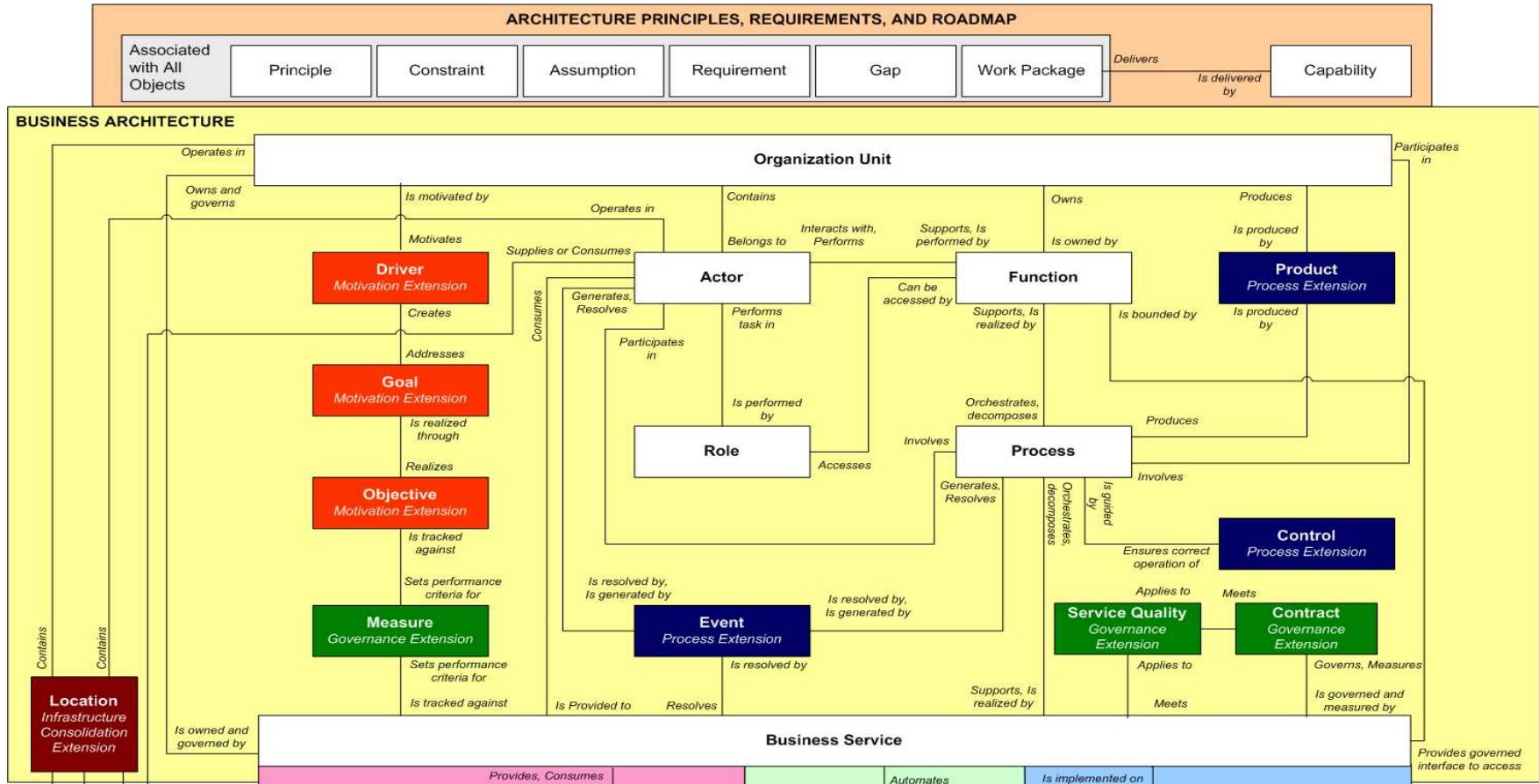
ArchiMate 2.0

Base Capability Map for BusinessActivator: BusinessActor

Capability	Model	Element	Example
BusinessActivator	BusinessActor	BusinessActor	BusinessActor = Luggage Insurance Department
		BusinessRole	BusinessRole = Travel Insurance Seller
		Process	Process = take out travel insurance
		Service	Service = Offering travel insurance

TOGAF

Business Architecture Metamodel



TOGAF 9.1**Base Capability Map for Organization: Actor**

Capability	Model	Element	Example
Organization	Actor	Actor Role OrganizationUnit Function Process BusinessService Event Location	

ESAMI – Enterprise Services Architecture Metamodel Integration

Consolidated EAM Reference Architecture: Analysis and Integration

Reference	EAM Reference Architecture			Correlation Index		Integration Options		Documents		
Origin	Capability	Model	Element	ArchiMate	TOGAF	ArchiMate	TOGAF	File	Pages	Authors
ArchiMate Specific. and TOGAF Standard	Business Activator	ActorRole	Actor	2	2	p	m		25-32 87-88	
			Role	3	2	m	m			
			Collaboration	3	0	m	r			
			Organiz. Unit	1	3	p	m			
			Business Function	2	3	p				
			Business Event		3	m	m			
	Organization	Organization Location	Location							
			Location							
			Location							
			Location							
			Location							

0 no correlation

1 low correlation

2 medium correlation

3 strong correlation

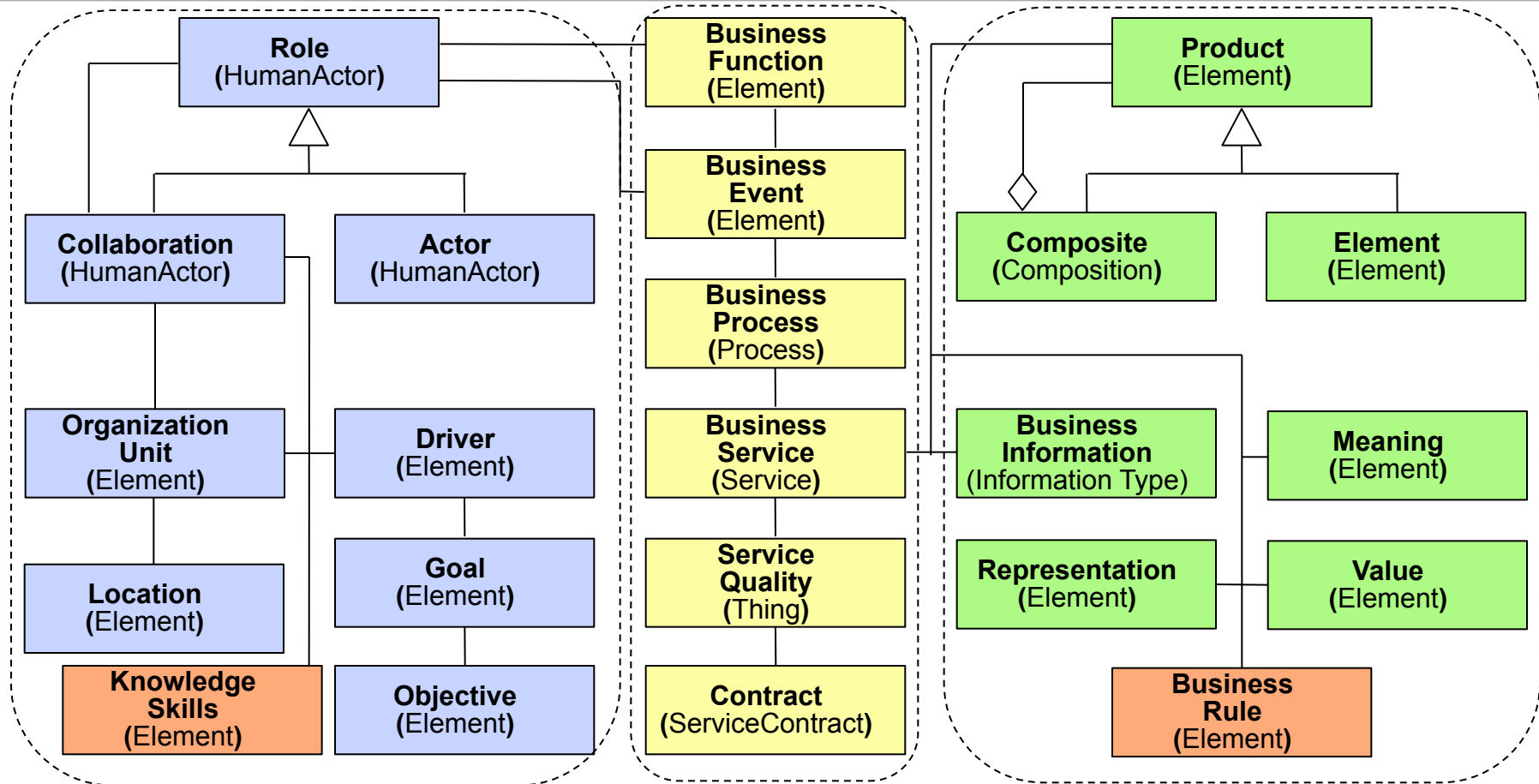
r reject

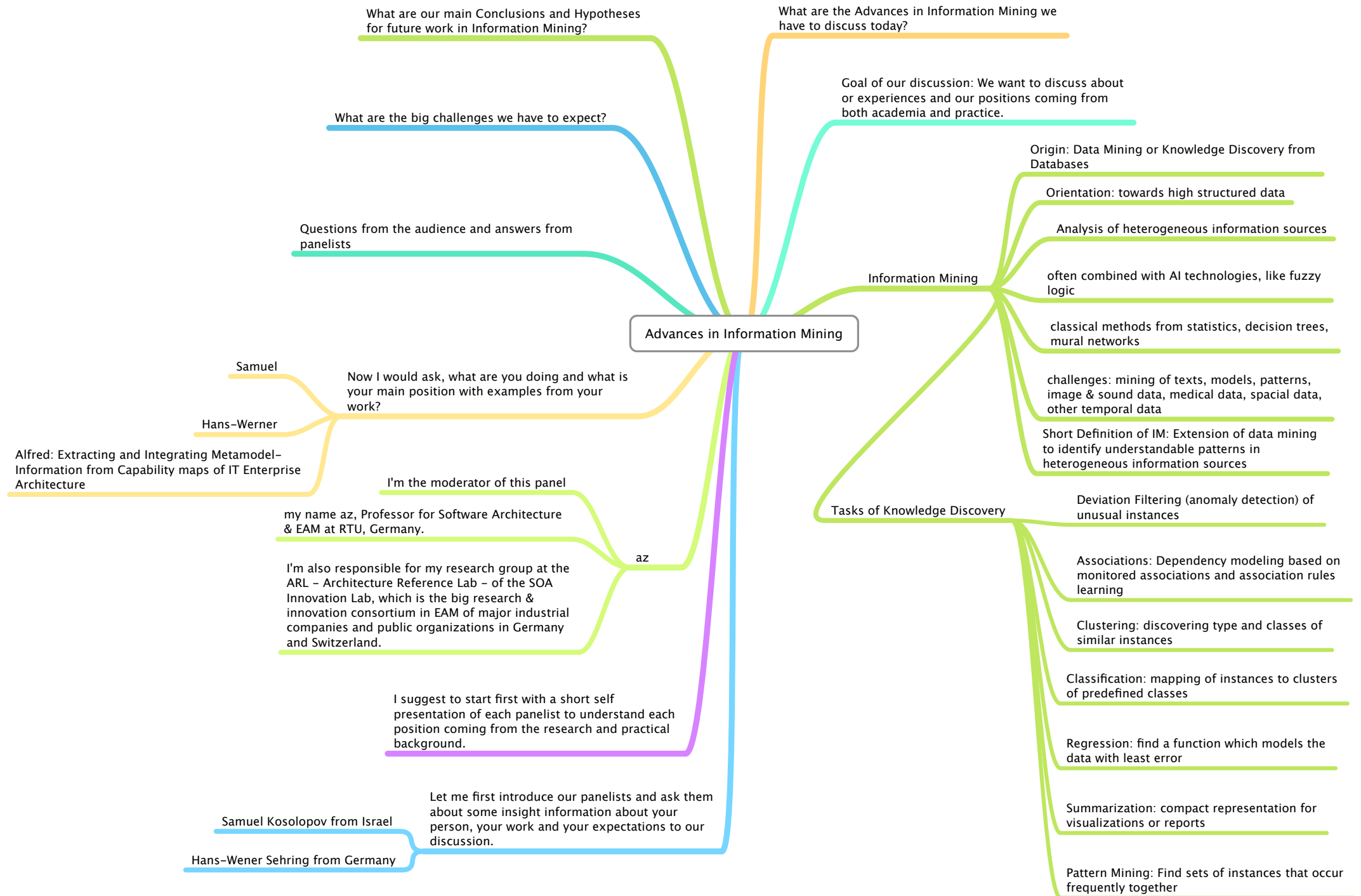
p partially

m mandatory (leading model)

ESAMI – Enterprise Services Architecture Metamodel Integration

SOA Ontology Typed Metamodel of Business Reference Architecture







www.iaria.org

**The Fifth International Conference on Creative Content
Technologies
CONTENT 2013**

May 27 - June 1, 2013 - Valencia, Spain

<http://www.iaria.org/conferences13/CONTENT13.html>

PANEL - CONTENT / PATTERNS 2013

Advances on Information Mining

Samuel Kosolapov, Hans-Werner Sehring, Bijan Raahemi, Alfred Zimmermann

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Positions from Panelists

1. Samuel Kosolapov

Addition of Rich Data from Smartphone Sensors to Image Metadata

2. Hans-Wener Sehring

Adapting Content Based on Users' Behavior

3. Bijan Raahemi

Machine Learning with Applications in Intrusion Detection

4. Alfred Zimmermann

**Extracting and Integrating Metamodel-Information from Capability
Maps of IT Enterprise Architectures**

Synthesis of Opinions and Discussion

1. **Questions** from the **audience** and **answers** from **panelists**
2. **What** are our **main conclusions** and **hypotheses** for **future work** in **Information Mining**?
3. **What** are the **big challenges** we have to face?

**Addition of rich environment,
location, motion, and orientation data
from smartphone sensors to Image Metadata
will provide better scene analysis.**

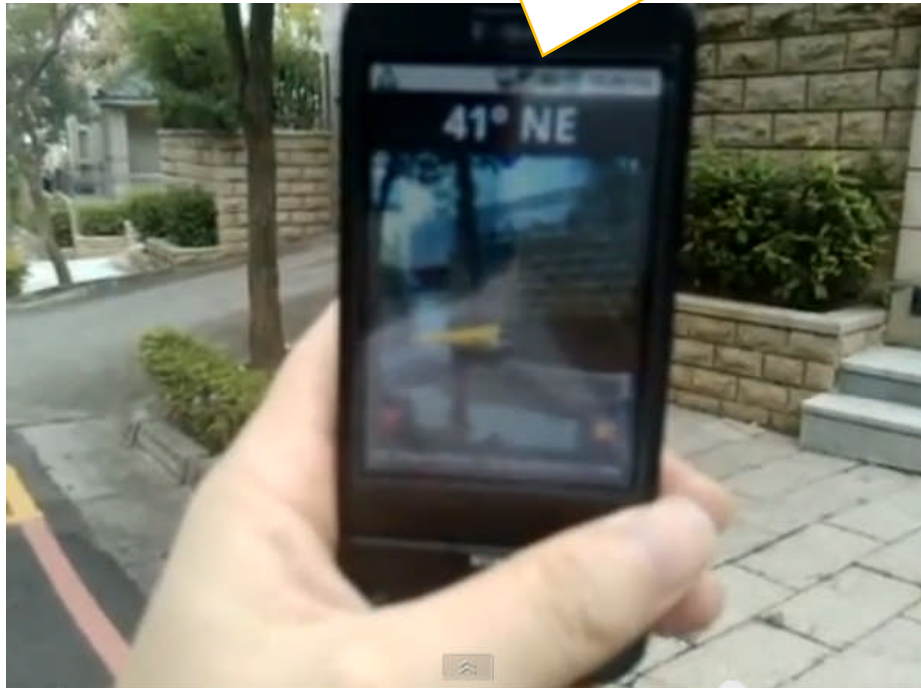
**Are current Camera and Image Analyzing
applications
ready for the change?**

Samuel Kosolapov

CONTENT/PATTERNS Panel.

CONTENT 2013. Valencia. 2013-05-29

**Google Street View in Compass Mode:
Compass that displays your location
and direction**



~ Standard JPG Image metadata

IMG_2301.JPG

JPEG image



Date taken:	2012-11-25 13:35
Tags:	Add a tag
Rating:	☆☆☆☆
Dimensions:	4000 x 3000
Size:	2.33 MB
Title:	Add a title
Authors:	Add an author
Comments:	Add comments
Camera maker:	Canon
Camera model:	Canon IXUS 220HS
Subject:	Specify the subject
F-stop:	f/2.7
Exposure time:	1/30 sec.
ISO speed:	ISO-640
Exposure bias:	0 step
Focal length:	4 mm
Max aperture:	2.875
Metering mode:	Pattern ▾
Flash mode:	No flash, compulsory ▾
Date created:	2013-02-12 17:11
Date modified:	2012-11-25 13:35

I want Sensors Info to be included into the Image Metadata
but in a standardized way.

Yes, everyone can add a TAG,
but in order to provide metadata search some standardization is required

Galaxy Nexus. List of Sensors

- **Camera** (Frons & Back) + Optical proximity sensor: GP2A
- **Microphone**
- **TouchScreen sensor**: *Melfas MMSxxx touchscreen*
- **Triaxial acceleration sensor** *BOSCH BMA250*
Measurement of accelerations in 3 perpendicular axes
Used to sense tilt, motion, shock and vibration. (range +/-2g... +/-16g; 10 bit !)
- **Triple Axis MEMS Gyroscope**: *InvenSense MPU3050*
3-axis gyroscope with an embedded Digital Motion Processor™
Digital-output X-, Y-, and Z-Axis angular rate sensors (gyros)
range of ± 250 to $\pm 2000^\circ/\text{sec}$
Used to sense: gesture recognition, panning, zooming, scrolling,
zero-motion detection, tap detection, and shake detection

+ Digital-output **temperature sensor**
- **Geomagnetic sensor**: tri-axial *Yamaha YAS530*
Resolution: 0.15uT (X,Y), 0.3uT (Z)
Used to sense: Absolute Direction (compass)
- **Barometric pressure sensor**: BOSCH BMP180
(300...1100 hPa, first of its kind in a smartphone)
Used to sense: height over the sea level

Why would I want more sensors data inside the image ?

- We have a huge amount of photos and videos in the cloud.
But search images by content is still very slow and is not always feasible.
- Metadata can provide additional info which can be used to narrow the search.
- Example: Leaning Tower of Pisa
“Engineers” announced that the Tower had been stabilized
such that it had stopped moving for the first time in its history.
They stated it would be stable for at least 200 years

I want to analyze set of images
taken from the same camera position, direction and orientation
in order to check this claim.

- Actually, sensor information may help to reconstruct 3D scene
from a plurality of 2D Images



Adapting Content Based on Users' Behavior.

Panel on Advances on Information Mining.
Computation World 2013, Valencia, Spain.

Dr. Hans-Werner Sehring, T-Systems Multimedia Solutions GmbH, Germany.



Scope of Information Mining Used in this Proposition.

- Information/content: purposeful data.
- Mining: Detect implicit relationships in data (similar: find structure in data).
- My field: Content Management.
 - Manually managed content (data).
 - Structure and relationships are well-know/prescribed.
- Observation: Content is perceived in ...
 - a context (“Context is King”) ...
 - by a user/consumer/use/situation/...

..T..Systems.....

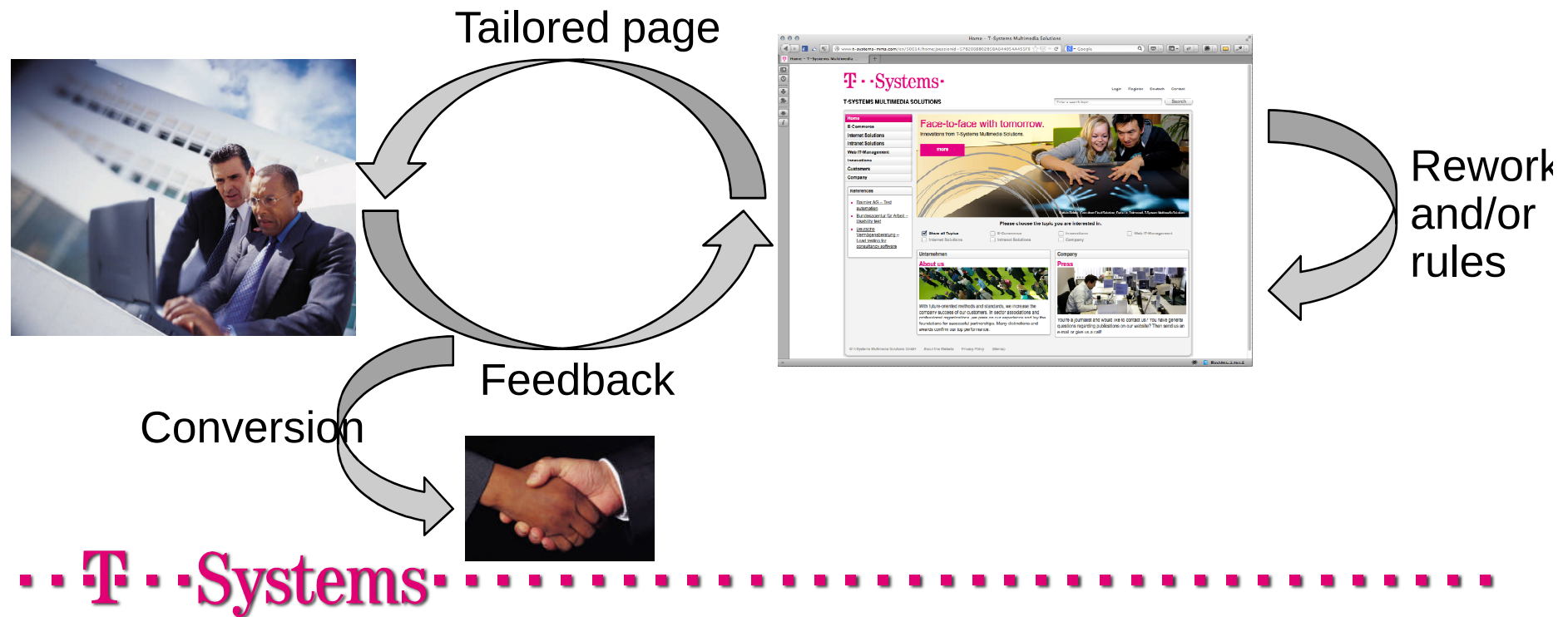
Learning About Users.

- Information = purposeful data, in a context; here: content targeted at a person.
- Information mining here: learning about users and their use of content.
 - Need to describe persons.
 - No explicit data available.
 - Not willing to give information: consumer, ...
 - Not possible to give info: diverse interest, subjective (non-explicable) views, ...
 - Known, e.g., from personalization based on tracking/profiling.
- Known, e.g., from personalization based on tracking/profiling.
 - Tracking actions.
 - Allowing registered users to give information about themselves.
 - Observing registered users, eventually including anonymous tracking history.

.. **T** .. **Systems** ..

Information Mining on the Web (I).

- In field of content management: Web Content Management now named
 - Web Experience Management (WCM)
 - Customer Experience Management (CEM, CXM)
- Currently of particular interest: marketing automation.



Information Mining on the Web (II).

- Topics, e.g.:
 - Web tracking (what did the customer do?).
 - Learn about customer journeys (how did the customer come to us?).
 - Segmentation (cluster/classify users).
- With the goals to ...
 - Move focus from Point of Sale (PoS) to Point of Contact.
 - Iteratively measure success and improve a site's appearance.
 - Personalization (for groups of users).
 - A/B testing.
 - Compute recommendations (based on segments).
 - Predictive marketing (based on user's history).
- Overall goal of course: increase conversion rate by attracting the customers in the “right” context.

.. **T** .. **Systems** ..

Information Mining on the Web (III).

- Challenges:
 - Getting to know about users.
 - How to identify anonymous users?
 - With different/changing devices?
 - Across different services provided by different service providers?
 - Within legal constraints?
 - How to distinguish different usage scenarios? E.g.,
 - Buy a new washing machine,
 - Having bought a new washing machine,
 - Looking for a washing machine to recommend to the mother in law.
 - How to incentivize users to give information about themselves?
 - ...?

Drawing conclusions from knowledge about users.

... **IT-Systems** ...