

Agents and Semantics for Human Decision-Making

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Agenda



Introduction

- Decision Making and Support Systems
- Agents and Semantic Technologies
- Agent-Based Semantic Decision Support
 - Showcases
 - Challenges





Human Decision-Making



Cognitive process of an **individual**, or a **group** which results in a final **choice**, or an opinion of choice **of a course of action** among **alternative** scenarios to accomplish given objective(s)



Decision-Making Model



Generic 7-Step Model (Brown, 2012)



Since the 1950s

- Detailed in more **specific (multi-criteria)** models based on **perspectives of human decision-making**
 - Psychological: Needs, preferences, values
 - Cognitive: Cont. deliberation integrated w/ environment)
 - Normative: Rational (logical, utility-driven) vs.
 Irrational choice
 - e.g. Recognition Primed Decision Model (Klein, 1989)

Decision support systems (DSS)

Early: Medical diagnosis, financial management, ...

Mathematical models, Statistical methods, data mining/OLAP, logical rule-based ...



Intelligent Decision Support Systems



Since late 1970s: Intelligent DSS

(Holsapple & Whinston 1977; Turban & Aronson 1998)

- Use of Al techniques to exhibit "intelligent behavior" of DSS
- Since 1990s: Use of agents in IDSS
- autonomous, proactive, reactive
 coordination of data, knowledge, services
 needed for personal and group decision support
 In distributed environments: collaborative (MAS)



Generic IDSS architecture

- GE Phillips-Wren (2008): Intelligent Agents in Decision Support Systems. Encyclopedia of Decision Making and Decision Support Technologies. IGI Global
- dto (2012): AI Tools in Decision Making Support Systems: A Review. Artificial Intelligence Tools, 21(2)



Examples



Prominent and mature applications of agent-based decision support include



Manufacturing, Transport & Logistics

MAS for production and fleet management DS via distributed planning, scheduling, etc.

✓ MAS for clinical decision support





E-Business

- Negotiation support for online market places, auctions
- ✓ Product recommendation
- ✓ Service **brokerage**, matchmaking



E-Health

Virtual Worlds

- User avatars for group decision
 support in virtual team meetings
- K Fischer, JP Müller (2014): Application Impact of Multiagent Systems and Technologies: A Survey. In: AOSE. Springer
- GE Phillips-Wren, LC Jain (2005). Intelligent Decision Support Systems in Agent-Mediated Environments. In: Frontiers in Artificial Intelligence and Applications Series, 115, IOS.





To better cope with problems of semantic interoperation and reuse

of decision support data, services and knowledge ...





Decision Support and Semantic Web



Since 2001: Intelligent DSS also use semantic Web technologies

- Modelling of decision domain semantics
 with W3C standardized and formal
 ontology and rule languages
 (OWL2, RDF/S, OWL-S, SA-WSDL, RuleML, SWRL)
- Supported with methods for ontology selection, evaluation, and alignment





Examples



Prominent and standard models of semantic decision domains include:

Domain ontologies (OWL)

- Negotiation [W3C]
- Medical [SNOMED, NIST]
- Sensor Networks [W3C SSN]
- Provenance [W3C PROV-O]

etc.

User profile / context ontologies

- GUMO (OWL), schema.org, ODP
- AmbiSense

Semantic service collections

OWLS-TC, SAWSDL-TC, hRESTS-TC @semwebcentral.org, >3k services

□ Various domains: Linked data sets (RDF/S)



Domain		Number of datasets	Triples	%	(Out-)Links	%
Media		25	1,841,852,061	5.82 %	50,440,705	10.01 %
Geographic		31	6,145,532,484	19.43 %	35,812,328	7.11%
Government		49	13 315 000 400	42.00%	10 242 510	3 94 %
Publications	More than 30 billion RDF triples as					
Cross-domain	linked factual knowledge available					
Life sciences	initice ractaal knowledge available					
User-generated content		20	134,127,413	0.42 %	3,449,143	0.68 %
		295	31,634,213,770		503,998,829	



Semantic Decision Support





... and a large number of **methods** for

ontology-based querying and reasoning:

- ✓ Semantic data
 - □ Facetted, Link-Traversal, P2P, Federated search (e.g. with SPARQL1.1)
 - □ Stream reasoning (e.g. with C-SPARQL)
 - Hybrid semantic and statistical analysis
 - Explanation of results
- ✓ Semantic services
 - Discovery, Selection (Matchmaking)
 - Composition planning

E Blomqvist (2012): Use of Semantic Web Technologies for Decision Support. Semantic Web Journal.





Agent-Based Semantic Decision Support

SHOWCASES IN INFOTAINMENT



Recommender Agents in the Web



Provide personal decision support:

What, where, why to buy, watch, read, listen to, etc. ?

by recommending **relevant** items to the user

pull and/or push mode



- content-based, social, or hybrid (profile, text similarity-based relevance)
- user/item context-sensitive (time, season, location, companion, etc.)

How can **semantics** help here?



Semantic Recommender Agents



Use of semantic model of decision domain

- User profile and domain ontologies (OWL)
- Knowledge graph with linked data sources (RDF/S)

to compute semantic relevance of items

□ Logical (item) concept subsumption relations

□ Knowledge graph analytical heuristics, etc.



- Improved accuracy compared to non-semantic approaches
- Better explanation of recommendations
- J Pazos Arias et al. (2012): Recommender systems for the social Web. Springer
- R De Virgilio et al. (2012): Semantic search over the Web. Springer
- E Peis et al. (2008): Semantic recommender systems. Hypertext.net, vol 6



Example: Semantic Relevance



Path-based item relevance heuristic SPrank:Frequency of different user-item path types (features) j



$$\vec{x}_{ui}(j) = \frac{\# path_{ui}(j)}{\sum_{d \in D} \# path_{ui}(d)} \in R^{L}$$

Semantic relevance feature vector of item i₁ for user u₃:

x₃₁ = (2/5, 2/5, 1/5), |p|<5

- collaborative
 #path(1) = (likes,likes,likes): 2
- content-based
 #path(2) = (likes,p₂,p₁): 2
 #path(3) = (likes,p₂,p₃,p₁): 1
- hybrid, |p|>=5
 #path(4) = (likes,p₂,p₁,likes,likes)

Regression-based learning of rank f(x_{ui}): Higher accuracy than common standards (up to 0.6 recall >> BPRLin, SLIM, SMRMF with test data from MovieLens, Last.fm)

T Di Noia et al. (2013): Top-N recommendations from implicit feedback leveraging Linked Open Data. Proc. ACM 7th RecSys Conf.



For Best Performance: Semantic or Non-Semantic ?



The 1M\$ Netflix Prize





Fixed combination of **non-semantic** predictors of movie ratings

- Won the contest but with **low precision** (accuracy RMSE_n 0.8567)
- Not suitable for top-N recommendation



Since 2011: Open contest on learning the best combinations (Corr.: multi-armed bayesian bandit problem solving)

Current research and development of recommenders:

Make use of semantic relations to avoid cold start problem, and

to compensate for non-semantic misclassifications.

Search for best hybrid recommender for given context is still ongoing ...

- R Yan et al. (2013): Using semantic technology to improve recommender systems based on Slope One. Springer
- JW Ha et al. (2014): EPE An embedded personalization engine for mobile users. IEEE Internet Computing, 18(1)



Example: Semantic Explanation





Use of **semantic relevance** methods to **summarize** the **most relevant properties** of top-ranked items in **multimedia** panel

... as you know it also from e.g. dbpedia,

Google or Facebook knowledge graph search



Example: Semantic Explanation (2)





For N-Item relational queries:

Display the **shortest item-item paths in the knowledge graph**

(Corr. NP-hard Steiner tree problem solution)



🖒 📭 please give Horst a thumb for this path - he is willing to learn!





Example: Context-Sensitive



BMW SmartCarAgent provides personal decision support to a driver based on his current context:

That looks nice, is it worth to visit?



Yes, this **church** can be of **interest** to **you**. Here is why: **[facts].**



- Recognize object, cognitive activity: Eye tracking system eye-viu, Semantic image retrieval
- Semantic fact search in associated knowledge graph (profile, LD sources)
- Cognitive activity level constrains presentation (voice, text on panel, extent of information): Reduced distraction, cognitive overload in driving situation

M Moniri, M Feld, C Müller (2012): Personalized In-Vehicle Information Systems. Proc. 8th IEEE Conf. on Intelligent Environments. Best Demo Award.



We Have That Already in Practice ?



State of the art example for in-car entertainment, since mid 2013



Apple's **Siri** Eyes Free system allows drivers "to make calls, dictate text messages, play music or podcasts, and **access** everything else the voice-controlled **virtual assistant** has control over."

Uses **semantic** knowledge graph search for selection of registered data and services ... but **no personal context-sensitive** action, **no explanation** of results ..



Example: TIFF MyMedia Agent



Provides personal and group decision support in social media

- (1) Semantic P2P search of annotated media data and services
- (2) Network-adaptive real P2P live streaming of videos



Trailer of (1) Any SciFi movies on play ? "Gravity" Movie "Gravity" [Trailer]. Sheila @CinemaZ Playing next @CinemaZ Pete Mobile P2P network (2) Shall we all go watch Festival WLAN/WiFi it at CinemaZ? Share & jointly watch trailer with buddies Sheila and Carl ✓ Share & jointly watch Carl So, yes, we like to - but only in a nice cinema live recording of CinemaZ with Pete and Carl



Semantic P2P Search in MyMedia

- Each peer observes semantics of query q and traversing items during k-walker search
- Semantic routing of item query q:
 Shortest path (within TTL) with maximal # of peers with semantic expertise for the topic of q
 - Logic-based semantic relevance of media data (OWL)
 - *Hybrid semantic* media service selection (OWL-S)
- Dynamic semantic replication of items (optional)
 - Demand-driven in query topic-based peer groups
 - Maximized utility of replication (semantic gain vs. traffic costs)
- Avg. Precision: 0.82 (1M peers, RPLG networks, random/Zipf), robust
- Real P2P live streaming with MPEG-DASH: 4s latency (4 Nexus7, G2)

X Cao, M Klusch (2013): Proc. 15th IEEE Conference on High-Performance Computing and Communications.B Rainer et al. (2013): Real-time Multimedia Streaming in Unstructured Peer-to-Peer Networks. Proc. 11th IEEE CCNC Conf.





D knows expert B (not F) for q. C knows expert F (not B) for q. F stands in for G (TTL=0) for q. Walker q returns from F to D.



Example: Semantic Service Mashup



TIFF-MyMedia Agent 2.0 with personal action planner:

Where to best have lunch on my way to watch movie A in which cinema in time?



- Uses profile, context and mobile semantic service planner OWLS-Xplan 2.2
- **KPI-driven offloading** of planning processes to (private/TIFF) cloud

Other examples @aimashup.org

2013: 50.000 web services [sousuo 4/2013]

2017: 10 billion mobile connected devices

2050: 50 billion things as services [CISCO 2013, W3C XG SSN]





Selected Challenges



 Usability & Scalability: Multimodal, contextual user interaction for dynamic, energy-efficient service selection/planning on mobiles
 HCI, Mobile cloud computing [Fernando+ 2013; Ha+ 2014]



- Interoperability across mobile telco providers
 WebRTC (W3C, IETF) e.g. for web-based (mobile) P2P apps
- Security & privacy of data, processes, user/group profiles that are offloaded to cloud or distributed in peer group: Semantic inference problem + technical security threats
- Semantic-empowered trusted recommendation
 - → Semantic item/user relevance for (social) trust computation [Martin-Vicente et al., 2012]





Agent-Based Semantic Decision Support

SHOWCASES IN MANUFACTURING



ISReal Framework for Web-Based Virtual Engineering





Provides **personal decision support** for the commissioning or revision of (parts of) a production line before or during its physical development:

- Functional simulation of physical line in its annotated virtual 3D model
- Representation of user as intelligent avatar (BDI agent) in 3D scene





Integrated Technologies





P Kapahnke, et al. (2010): Proc. 9th Intern. Semantic Web Conference (ISWC), LNCS 6414. Springer



Example: Decision Support Scenario



Interactive query-answering on functionality of machines:



Can I produce 20 pills with this machine within 30 secs ?





- → Planning of machine services to reach goal
- → Time verification of plan
- → Plan **execution** (functional simulation)
- → Explanation (text2speech, animate plan execution)

Yes, this pill filling station can do it. I show you how to do it.



Example: Decision Support Scenario (2)



Agent identifies and explains failures of machines or their handling, and creates alternative(s) for its user:



Why did my handling of this machine fail ?





- \rightarrow Time-based **verification** of plan execution
- → Visualization of failure trace (**explanation**)
- \rightarrow Building of **alternative plan** with revised state

You lifted the carriage stopper too late (11s > 2.5s). Here is your updated control plan for handling this machine.



C3D Framework for Web-Based Collaborative Engineering



Provides group decision support for the design of virtual production lines.

Multiple designers represented by avatars in shared, annotated XML3D scene.



- High-precision local and
 P2P search of 3D models
- ✓ Synchronized joint view
 and design of 3D objects
- Functional verification of designed objects

BMBF project Collaborate3D (2011 – 2015): c3d.dfki.de

I Zinnikus et al. (2013): A Collaborative Virtual Workspace for Factory Configuration and Evaluation. Proc. 9th IEEE CollaborateCom.



Selected Challenges



Scalability



- → Web-based simulation of very large-sized, annotated 3D scenes (>>20 interacting user agents)
- Collaborative hybrid BDI planning agents under uncertainty
- Virtual condition monitoring of (designed) machines
 - → Semantic stream reasoning with CEP and verification of realworld sensor data for fault detection/diagnosis in virtual world
 Klusch et al. (2013): ICM-Wind: Semantics-empowered CM of wind turbines. ACM SAC
 BMBF project INVERSIV (2015 – 2018), EU IP OSMOSE (2013 – 2016)





Agent-Based Semantic Decision Support

SHOWCASES IN SMART RETAIL



CARAMEL: Virtual 3D Design of Store Layouts



Provides **personal decision support** for choosing the right layout of retail stores: How to best position the shelves to increase sales for which types of customers ?





ADIGE: Smart Re-Ordering of Goods



9 software

Provides **personal decision support** for manager for re-ordering of goods:

When to re-order from which provider ?





at any time due to changes in services performance, availability, prices

✓ ADIGE agent adapts the annotated process model (OWL) to these changes
 by semantic selection and re-planning of actual process services (OWL-S)





DemoVid

Take-Home Messages





For decision support, agents can make use of semantics to
✓ facilitate semantic data and service interoperation
✓ perform more precise search for relevant items
✓ create and recommend more accurate alternatives
in a more context-sensitive, explanatory way.

Many agent-based semantic decision support apps in various domains already available or in development.

Main challenges: Scalability, usability, privacy and trust



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Thank you for your attention ! Questions ?



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