## Iterative, Interactive Analysis of Agent-Goal Models for Early Requirements Engineering

#### RE Seminar DISI, University of Trento

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DISI, University of Trento June 14, 2012

### Iterative, Interactive Modeling

- Models are helpful in the development of software systems:
  - Means of abstraction
  - Communication
  - Convergent understanding
- Models support analysis:
  - Using the structure or contents of the model to answer questions
- It is challenging to know when a model is (sufficiently) accurate or complete
- We need: methods and tools to guide modelers in an interactive process of model iteration
  - Leading to more stable and complete models and improved analysis



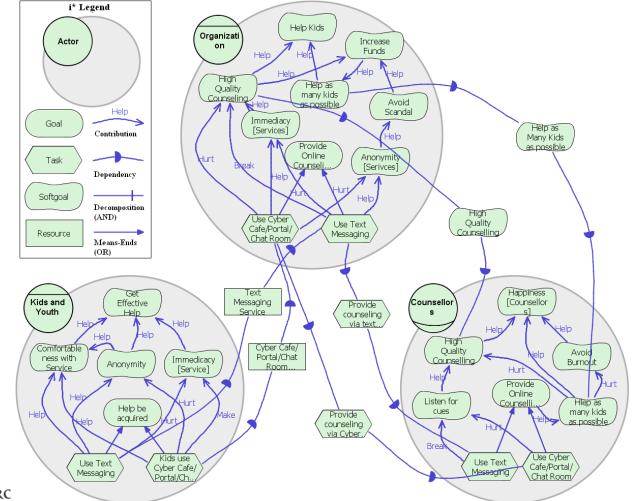
### Iterative, Interactive Modeling

- In Early Requirements Engineering (ERE), when models describe what will be, it is especially challenging to make models sufficiently complete and correct
- ERE focuses on understanding the domain and the (often conflicting) needs of the stakeholders enough to gain a high-level understanding of the required functionality for the system-to-be
  - Domain information in early project stages is often incomplete
  - Success of the system often relies on important non-functional success criteria
  - Involvement of key stakeholders is important
  - Key decisions concerning project scope or functionality are made
- Goal models are useful for ERE modeling and analysis
  - NFR, i\*, KAOS, Tropos, GRL, etc.
- Existing goal model analysis procedures are typically automated or require detailed information

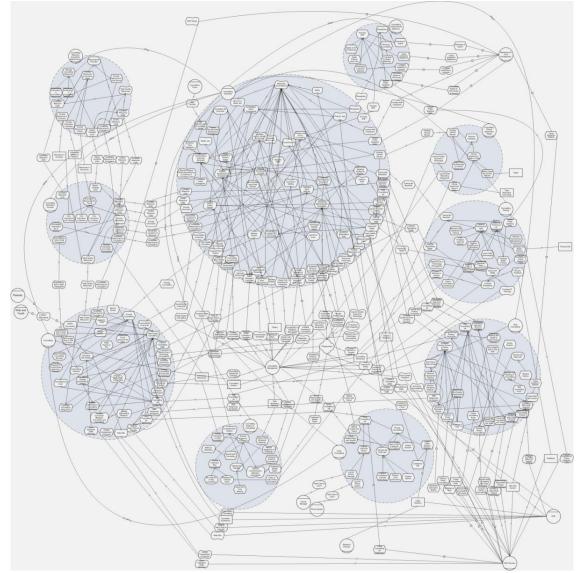


### Motivating Example: Youth Counseling Organization

A not-for-profit organization that focuses on counseling for youth over the phone, but must now expand their ability to provide counseling via the Internet



#### Motivating Example: Youth Counseling Organization



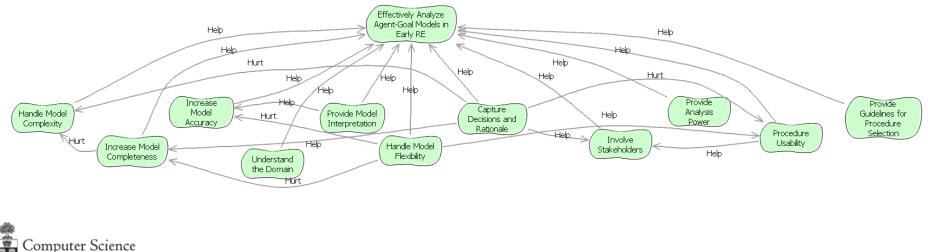
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## Challenges in Agent-Goal Model Analysis for Early Requirements Engineering

- Model Complexity
- Model Completeness
- Model Accuracy
- Domain Understanding
- Model Interpretation
- Model Flexibility

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- Decision Rationale
- Stakeholder Involvement
- □ Analysis Power
- Procedure Usability
- Procedure Selection



## Requirements for Early RE Agent-Goal Model Analysis

#### Model Complexity

- R1 Scalability
- R2 Analysis Comprehension
- R<sub>3</sub> Partial Automation
- Model Completeness & Accuracy
  - R4 Model Iteration
  - R5 Interactive Procedure
- Domain Knowledge
  - R6 Prompt Further Elicitation
- Model Interpretation
  - R7 Definition
- Model Flexibility

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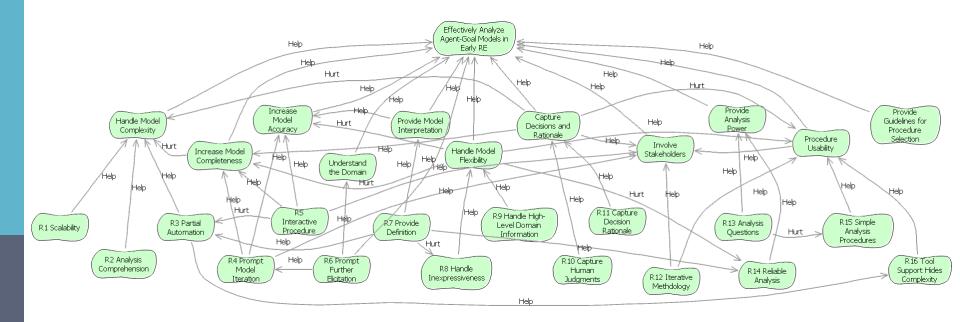
- R8 Handle Inexpressiveness
- R9 Handle High-Level Domain

#### Information

- Decision Rationale
  - R10 Human Judgments
  - R11 Decision Rationale
- Stakeholder Involvement
  - R12 Iterative Methodology
- Analysis Power
  - R13 Analysis Questions
  - R14 Reliable Analysis
- Procedure Usability
  - R15 Simple Analysis Procedures
  - R16 Tool Support Hides Complexity
- Procedure Selection

## Requirements for Early RE Agent-Goal Model Analysis

Requirements Agenta Gyar Madeh Anabysis for Early alysis
Requirements Engineerings





## **Thesis Statement**

- Framework for Iterative, Interactive Analysis of Agent-Goal Models in Early RE
  - Provide methods, algorithms, and tools
- Address challenges in Early RE Agent-Goal Model Analysis
  - Aim to provide analysis power, improve model quality, increase domain knowledge...
- Framework validation
  - Provide tools and methods which are usable in practice
- We claim that such contributions will ultimately lead to the development of more effective software systems

## Iterative, Interactive Agent-Goal Model Analysis Framework

- Goal Model Analysis Review
- Selection Criteria (SAC'11)
- Forward Satisfaction Detailed Comparison (REJ'11)
- Reflective Analysis and Definition of i\* (RiGIM'08, iStar'08)
- Forward Evaluation (Caise'09 (short), PoEM'09, IJISMD'10)
- Backward Evaluation (iStar'o8, ER'10)
- Analysis Visualization (REV'10)
- Human Judgment Checks (iStar'ıı)
- Suggested Methodology (Caise'09, PoEM'09, IJISMD'10, PoEM'10)
- Implementation (iStar'11)

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Framework Validation ((coauthor) HICSS'07, REFSQ'08, (first author) PoEM'09, IJISMD'09, PoEM'10, REV'10)

# **Related Work**

#### Existing approaches to goal model analysis

- Forward and backward satisfaction propagation: (Giorgini et al., 2004), (Amyot et al., 2010), (Letier & van Lamsweerde, 2004)...
- Metrics: (Franch, 2006)...
- Planning: (Bryl et al., 2007)...
- Simulation: (Gans et al., 2004)...
- Model Checking: (Fuxman et al., 2004)...

#### Other approaches

- Goal model development approaches
- Goal model visualization approaches
- Analysis methods in RE and business

#### Horkoff, Depth Oral'08; Horkoff & Yu, SAC'10

# **Procedure Selection Guidelines**

- □ Abundance of goal model analysis approaches
  - What are the differences?
  - When would I use one and not another?
- Have provided a survey of methods
  - What methods are available?
  - What types of analysis questions can these methods answer?
  - What types of goal modeling constructs do the procedures support?
  - What information is needed in order to use the methods?
- Mapping and Selection
  - What available methods can be applied to achieve which kinds of usage objectives?
  - How can we use this information to advise on selection?



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#### Horkoff & Yu, SAC'10

## Goal Model Analysis Approaches: Classification and Additional Information

	Approach							Analysis Results			Additional Notation Supported		Notation	Analysis Scope	
Approach	Satisf Forwds	Satisf Backwds	Human Interv	Metrics	Plan- ning	Simu- lation	Model Check	Qual	Quant	Binary	Depend- encies	Soft- goals	Contrib. Links	Global	Loca
Jureta et al., 2008, 2010	N	Y	N	N	N	N	N	N	N	Y	N	Y	N	Y	N
Chung et al., 2000	Y	N	Y	N	N	N	N	Y	N	Y	N	Y	Y	Y	N
Giorgini et al., 2002, 2004a	Y	N	N	N	N	N	N	Y	Y	Y	N	М	Y	Y	N
Giorgini et al., 2004b	Y	Y	N	N	N	N	N	Y	N	Y	N	М	Y	Y	N
Giorgini et al., 2005	Y	Y	N	N	N	N	N	Y	N	Y	М	Y	Y	М	Y
Ernst et al., 2010	Y	Y	М	N	N	N	N	Y	N	Y	N	М	Y	Y	Y
i* Evaluation, Horkoff, 2006	Y	N	Y	N	N	N	N	Y	N	Y	Y	Y	Y	Y	N
Maiden et al., 2007	Y	N	Y	N	N	N	N	N	N	Y	Y	Y	Y	Y	N
Z.151, 2008, Amyot et al., 2010, Pourshahid et al., 2011	Y	N	N	N	N	N	N	Y	Y	Y	Y	Y	Y	Y	Y
Asnar & Giorgini, 2006	Y	Y	N	N	N	N	N	Y	М	Y	N	М	Y	Y	N
Barone et al., 2011	Y	N	N	N	N	N	N	Y	Y	Y	N	М	Y	Y	N
Letier & van Lamsweerde, 2004	Y	Y	N	N	N	N	N	N	Y	Y	М	N	N	Y	N
van Lamsweerde, 2009	Y	N	N	N	N	N	N	N	Y	Y	N	Y	М	Y	N
Franch & Maiden, 2003	N	N	N	Y	N	N	N	N	Y	N	Y	Y	N	Y	N
Franch et al., 2004	N	N	N	Y	N	N	N	М	Y	N	Y	Y	N	Y	Y
Franch, 2006	N	N	Y	Y	N	Ν	N	Y	Y	N	Y	Y	Y	Y	Y
Kaiya et al, 2002	N	N	N	Y	N	N	N	N	Y	Y	N	N	М	Y	N
Tanabe et al., 2008	Y	N	N	Y	N	N	N	N	Y	Y	N	N	М	Y	N
Bryl et al., 2006a	N	N	N	Y	Y	N	N	N	Y	Y	Y	N	N	М	Y
Bryl et al., 2009a	N	N	Y	Y	Y	N	Y	М	Y	Y	Y	N	N	М	Y
Asnar et al., 2007	Y	Y	Y	Y	Y	N	N	Y	N	Y	Y	М	Y	Y	N
Liaskos et al., 2011	N	N	М	N	Y	N	N	Y	Y	Y	N	М	N	Y	N
Gans et al. 2002	N	N	Y	N	N	Y	Y	N	N	Y	Y	N	N	Y	N
G ( )	N	N	Y	N	Y	Y	Y	N	Y	Y	Y	Y	Y	М	М
Gans et al., 2001, 2003b	N	N	N	Y	N	Y	М	N	Y	Y	Y	N	N	Y	N
Wang & Lesperance, 2001	N	N	N	N	N	Y	N	N	N	Y	N	N	N	Y	N
Fuxman et al., 2001, 2003	N	N	М	N	N	N	Y	N	N	Y	Y	Y	N	Y	М
Giorgini et al., 2004c	N	N	N	N	N	N	Y	N	N	Y	Y	N	N	Y	N
	N	N	N	N	Y	N	Y	N	N	Y	Y	N	N	Y	N

Classification by analysis approach, result, notation supported, and scope

Summarize additional information needed (e.g., cost, preference, metrics)

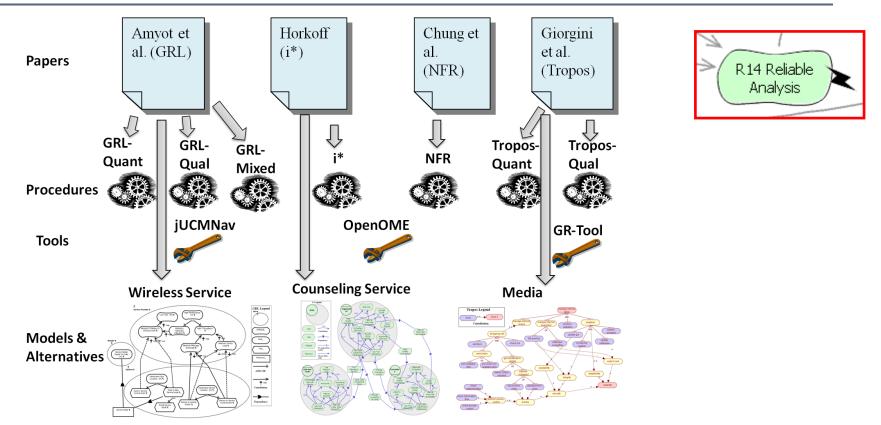
	Additional Information	Required by								
1	Goal Cost	Satisfaction Analysis: (Giorgini et al., 2005)(Asnar et al., 2007) (Giorgini et al., 2004b)(Asnar & Giorgini, 2006), Planning: (Bryl et al., 2006a)								
2	Risk	Satisfaction Analysis: (Asnar & Giorgini, 2006), Planning: (Asnar et al., 2007)								
3	Textual Arguments	Satisfaction Analysis:(Maiden et al., 2007) , Metrics, Model Checking: (Kaiya et al., 2002)								
4	Probabilistic Information	Satisfaction Analysis: (Giorgini et al., 2005) (Letier & Lamsweerde, 2004)								
5	Events and Treatments	Satisfaction Analysis: (Asnar & Giorgini, 2006)								
6	Importance/Priority	Satisfaction Analysis: (Asnar & Giorgini, 2006) Planning: (Liaskos et al., 2011)								
7	Actor Capabilities	Planning: (Bryl et al., 2006a, 2007) (Asnar et al., 2007), Model Checking: (Bryl et al., 2006a):								
8	(Pre/Post) Conditions/ Temporal Information	Planning: (Liaskos et al., 2011) Simulation: (X. Wang & Lespérance, 2001) (Gans et al., 2003a) (Gans et al., 2005) (Gans et al., 2003b), Model Checking: (Fuxman et al., 2001) (Fuxman et al., 2003)								
9	Delegation/Ownership	Model Checking: (Gans et al., 2002) (Bryl et al., 2006b):								
10	Trust	Planning: (Asnar et al., 2007), Simulation: (Gans et al., 2003b), Model Checking: (Giorgini et al., 2004c) (Bryl et al., 2006b):								
11	Speech Acts	Simulation: (Gans et al., 2003b)								
12	Confidence and Distrust	Simulation: (Gans et al., 2003b)								
13	Preferences	Satisfaction Analysis: (Jureta et al., 2008, 2010), (Ernst et al. 2010) Planning: (Liaskos et al., 2011) Model Checking: (Kaiya et al., 2002)								
14	Cardinalities	Simulation:(X. Wang & Lespérance, 2001), Model Checking: (Fuxman et al., 2003)								
15	Domain specific formula	Satisfaction Analysis: (A Pourshahid et al., 2008) (Barone et al., 2011) (Letier & Lamsweerde, 2004)								
16	KPIs/Metrics/Gauges	Satisfaction Analysis: (Pourshahid et al., 2008) (Pourshahid et al., 2011) (Barone et al., 2011) (Lamsweerde, 2009)								
17	Mandatory/Optional Requirements	Satisfaction Analysis: (Jureta et al., 2008, 2010), (Ernst et al., 2010) Planning: (Liaskos et al., 2011)								
Age	ent-Goal Models for Early	RE Horkoff et al. 13								

Interactive Analysis of Agent-Goal Models for Early RE

## Mapping Procedures to Objectives

		Recommended Procedures Provide						
Understanding		Yes. Try: Agent Approaches: i*/GRL Satisfaction Anal ([11][12][13]) Tropos Metrics, Planning, or Model Ctr SNET([16][17][18]) Coloridation						
	QU2. Do you need to understand details of the system at this point? Do you have access to detailed information such as cost, probabilities, and conditions? Can you express necessary or desired domain properties?	([3][21][22][23]) KAOS Satisfaction Analysis ([31]) GRL Quar ([11][12][13]) Tropos Planning ([4][6][7][8]) Tropos Modeling Checking ([8][14][15][19]						
	Objectives	SNET([16][17][18][18]) i* Simulation([34]), or Model Checking: Tropos ([8][14][15][19] SNET([16][18])						
]	Justify recommendations?	Yes. Try: Forward Satisfaction Approaches: NFR([9]) Tropos([3][21][22][23]) KAOS([31 *([26][33]) GRL([1])						
Improvement	models?	No. Try: Interactive Approaches: NFR([9]) i*([26][27][33]) Tropos([4][7]) SNET([16][18]) Metrics([11])						
		Yes. Try: Model Checking: Tropos([8][14][15][19]) SNET([16][18])						
		Yes. Try: Agent Approaches: i*/GRL Satisfaction Analysis ([1][26] [27][33]) i* Metric ([11][12][13]) Tropos Metrics, Planning, or Model Checking ([4][6][7][8][14][15][19]) SNE ([16][18])						
Elicitation	further elicitation? Procedures	Yes. Try: Interactive Approaches: NFR([9]) i*([27][27][33]) Tropos([4][7]) SNET([16][18]) i Metrics([11])						
		Yes. Try: Quantitative or Detailed Information: Tropos Probabalistic Satisfaction Analysis ([3][21][22][23]) KAOS Satisfaction Analysis ([31]) GRL Quant. Analysis ([1]) i* Quant. Metric ([11][12][13]) Tropos Planning ([4][6][7][8]) Tropos Modeling Checking ([8][14][15][19] SNET([16][17][18][18]) i* Simulation([34])						
		Yes. Try: Approaches supporting softgoals or contributions: NFR([9]) i* Satisfaction Analys ([26][27][33]) Tropos Satisfaction Analysis ([3][21][22][23]) Tropos Model Checking([14][15 GRL([1]) i* Metrics([11][12][13]) SNET([16][17][18])						
	QE4. Do you need to capture domain assumptions?	Yes. Try: Approaches using Satisfaction Arguments: i* Satisfaction Arguments [33]						
Requirements	QR1. Are you working with a system where safety/security/ privacy/risks or other specific	Yes. Try: Analysis over Specific Constructs or Metric Approaches: KAOS([31])						
Improvement	properties are critical considerations?	Metrics([11][12][13]) AGORA([30]) Tropos Risk, Trust, and Security([3][4] [8][19]) SNE Trust([17])						
		Yes. Try: Model Checking: Tropos([8][14][15][19]) SNET([16][18])						
Design		<b>No. Try:</b> Agent, Planning, Forward and Backward Satisfaction Approaches: NFR([9]) i <sup>2</sup> Satisfaction Analysis ([26][27][33]) Tropos Planning([4][6][7][8]) KAOS([31]) GRL Forward Satisfaction Analysis([1]) SNET Planning([16][18])						
	QD2. Are you aware of a sufficient number of detailed design alternatives?	No. Try: Quantitative Planning, Forward and Backward Satisfaction Approaches: KAOS Satisfaction Analysis ([31]) GRL Forward Satisfaction Analysis([1]) Tropos Planning([6][7] SNET Planning([16][18])						
	QD3. Do you need to evaluate and choose between high-level design alternatives?	Yes. Try: Satisfaction Analysis, Metrics and Agent Approaches: KAOS Satisfaction Analysis([31]) i* Forward Satisfaction([26][33]) GRL Satisfaction Analysis([1]) i* Metrics([11][12][13]) Tropos Risk([4])						
	QD4. Do you need to evaluate and choose between detailed design alternatives?	Yes. Try: Quantitative or Detailed Information: Tropos Probabalistic Satisfaction Analysis ([3][21][22][23]) KAOS Satisfaction Analysis ([31]) GRL Quant. Analysis ([1]) i* Quant. Metric: ([11][12][13]) Tropos Planning ([4][6][7][8]) Tropos Modeling Checking ([8][14][15][19] SNET([16][17][18][18]) i* Simulation([34])						
	QD5. Do you need to find acceptable processes?	Yes. Try: Planning Approaches: Tropos Planning([4][6][7][8]) SNET Planning([16][18])						
	QD6. Do you need to test run-time operation before implementation?	Yes. Trv: Simulation Approaches: SNET([16][17][18]) i* Simulation([34])						
Computer								

# Forward Satisfaction Techniques: Detailed Comparison



Analysis results over "softer" models (dependencies, softgoals) can differ greatly between procedures with different model interpretations

- Goal model analysis should be used as a heuristic, not an oracle
  - Emphasizes the other benefits of goal model analysis

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Iterative, Interactive Analysis of Agent-Goal Models for Early RE Horkoff et al.

Horkoff & Yu, REJ'11

# Suitability of Existing Goal Model Analysis Approaches for Early RE

- Model Complexity
  - **Given Some argue for scalability**
  - Few address analysis comprehension
  - Almost all are fully automated
- Model Completeness & Accuracy
  - Only a few mention iteration over models
  - Most are not interactive
- Domain Knowledge
  - Not addressed explicitly
- Model Interpretation
  - Several formal interpretations
- Model Flexibility

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- **Most are formal or quantitative**
- Many require additional information

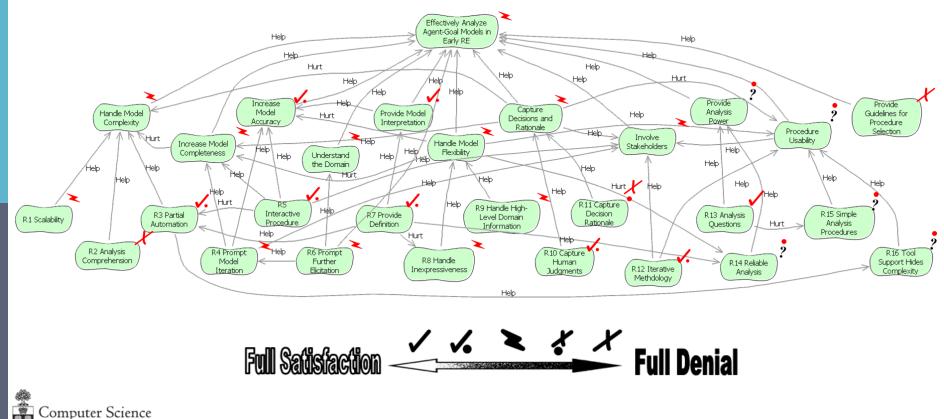
- Decision Rationale
  - Only a few capture human judgment or design rationale
- Stakeholder Involvement
  - Many do not provide methodology
  - Existing methodologies mostly focus on analysis and not iteration
- Analysis Power
  - Many analysis questions can be answered
  - Reliability addressed indirectly via examples and case studies
- □ Procedure Usability
  - Most do not address usability of procedure or tool support

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- Procedure Selection
  - Not addressed

# Suitability of Existing Goal Model Analysis Approaches for Early RE

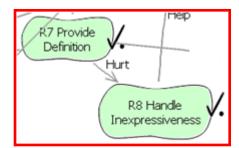
 Satisfaction Analysis for the Requirements for analysis of Agent-Goal Models in Early RE based on a Combination of Existing Work



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# Reflective Analysis and Formal Definition of i\*

- As i\* has been adopted and used, it's use has evolved
- Survey 15 student projects and 15 academic papers using i\*
- Looking for variations from "U of T" style
- Lead to the creation of strict and loose i\* syntax
  - List of i\* sytnax errors and warnings



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- Create more precise definition of i\* which to help remove ambiguity in the syntax and support semi-automated analysis
  - E.g., **Definition: agent-goal model.** An *i*\* model is a tuple M = <I, R, A>, where I is a set of intentions, R is a set of relations between intentions, and A is set of actors.
- Definition aims to support common variations (warnings)

Horkoff et al., RiGIM'08; Horkoff & Yu, ER'10; Grau et al., iStar'08, i\* Wiki, 2010 Computer Science UNIVERSITY OF TORONTO Iterative, Interactive Analysis of Agent-Goal Models for Early RE Horkoff et al.

## Iterative, Interactive, Forwards Satisfaction Analysis

- An analysis alternative is given in the model and its effects are propagated "forward" through model links
- Propagation defined using qualitative labels through model links
- Interactive: user input (human judgment) is used to decide on partial or conflicting evidence "What is the resulting value?"
- Adapted from Horkoff (2006) (MSc Thesis)
  - Described in new formalism
    - **•** Forward propagation axioms
  - Updated algorithm, O(n)
  - Readdress convergence & termination

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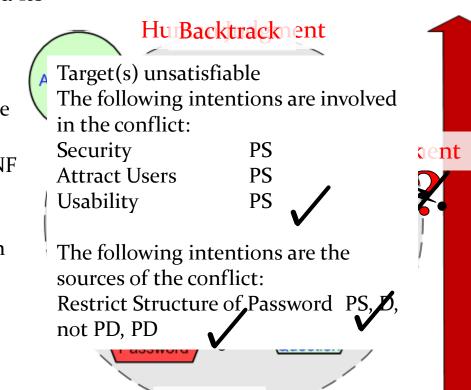
Iterative, Interactive Analysis of Agent-Goal Models for Early RE Horkoff et al.

#### Human Judgment Attract Application Users Help Human Judgment molement Usability Security Password System Restrict Ask for Secret Structure of Questio Password What if...? Full Satisfactio



## Iterative, Interactive, Backward Satisfaction Analysis

- A target is placed on the model and its affects are propagated "backward" through model links
  - Asks for human judgment "What incoming values could produce the target value?"
  - Model is iteratively encoded in CNF form and passed to a SAT solver
    - Backward propagation axioms
  - Modify and expand encoding from (Giorgini et al., 2004b)
    - Single analysis value per goal
    - More analysis values
    - Extra i\* syntax
  - Model restrictions: no mixture of links, no cycles
  - Algorithm worst case O(6<sup>q</sup>(ln<sup>2</sup> + n(zChaff)))



Conflic

### Is this possible...? How?

Horkoff & Yu, iStar'o8, ER'10

R13 Analysis

Ouestions

## A Methodology for Agent-Goal Model Creation and Analysis

#### Apply the following steps iteratively:

- Stage 1: Purpose and Elicitation
  - Identify scope or purpose of the modeling process.
  - Identify modeling participants and/or model sources.

#### **Stage 2: Model Creation**

- Identify relevant actors and associations.
- Identify relevant dependencies.
- Identify actor intentions.
- Identify relationships between intentions.

Stage 3: Analysis

Alternative Effects (Forward Analysis)

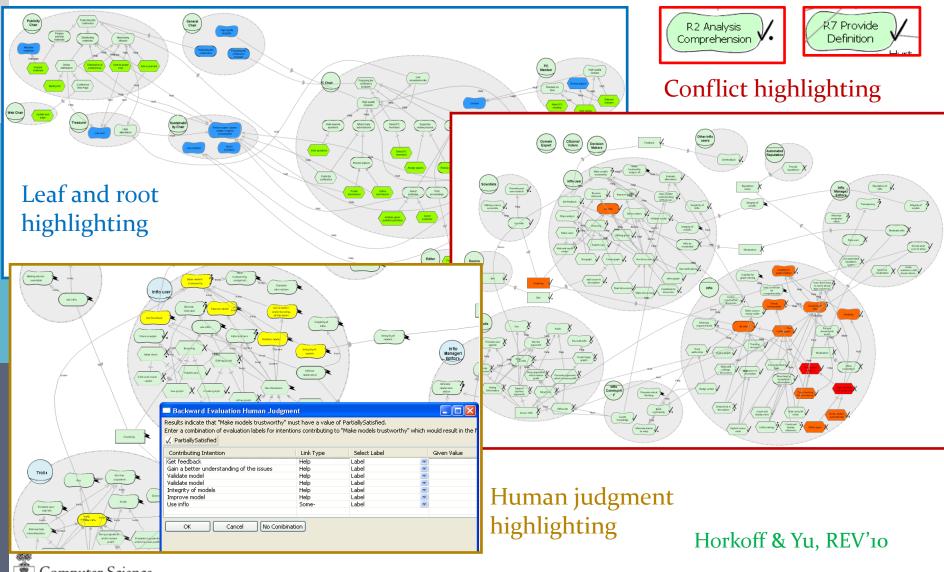
Iteration

- Identify all leaf intentions in the model, evaluate:
  - Implementing as much as possible.
  - Implementing as little as possible:
  - Reasonable Implementation Alternatives.
- Achievement Possibilities (Backward Analysis)
  - Identify all roots in the model, evaluate:
    - Maximum targets.
    - Minimum targets.
    - Iteration over minimum targets.
- Domain-Driven Analysis (Mixed)
  - Use the model to answer interesting domain-driven questions.

#### Horkoff & Yu, PoEM'09, IJISMD'10

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## Visualization Techniques for Analysis

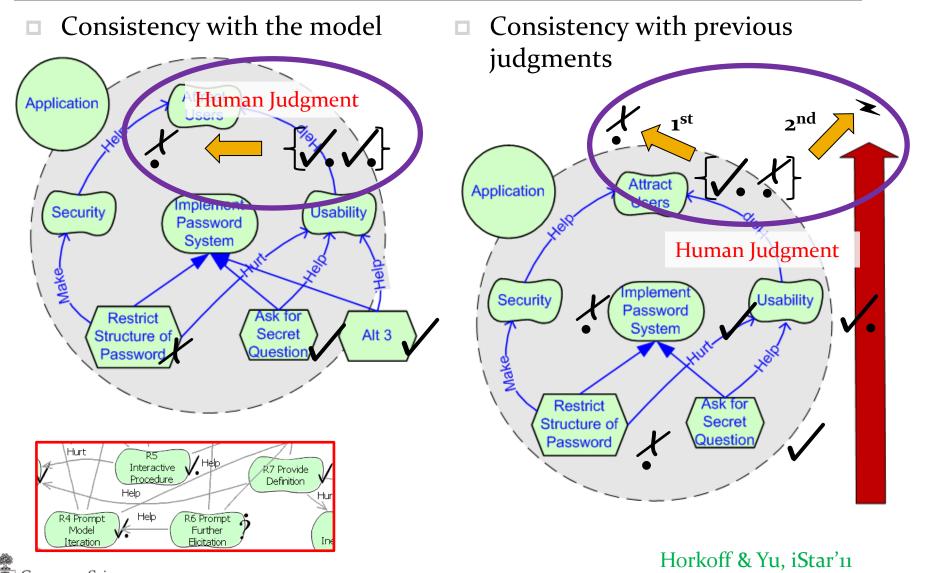


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Iterative, Interactive Analysis of Agent-Goal Models for Early RE

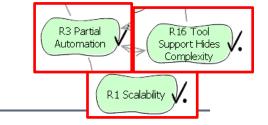
Horkoff et al.

### Detecting Judgment Inconsistencies to Encourage Model Iteration in Analysis



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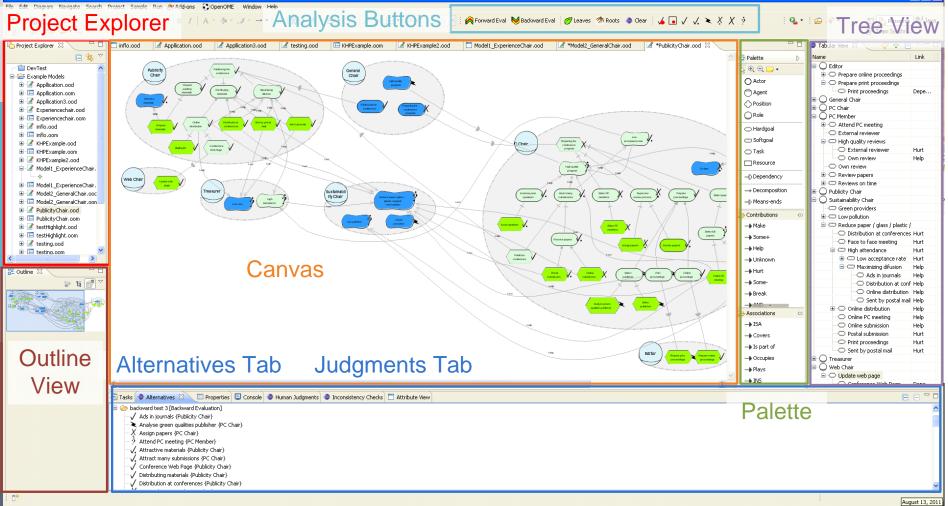
# Tool Support: OpenOME



#### Resource - Example Models/PublicityChair.ood - Eclipse Platform

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#### Horkoff, Yu &Yu, iStar Tool Fair'ıı

## Framework Validation: Manual Forward Procedure Application

- Analyzing effects of Trusted Computing
- Modeling strategy documents (FIS Strategy, USA Security Strategy)
- Counseling Service
  - Stage 1 (Horkoff, 2006): Understanding organization and technology options
  - Stage 2 (Horkoff, 2006): Requirements for revised online counseling system
  - Stage 3: Knowledge management needs and options
    - □ Analyzing effectiveness of knowledge transfer agents (Strohmaier et al., HICSS'07)
    - Analysis as part of pattern application (Strohmaier et al., REFSQ'08)
- Contributions and Lessons
  - Demonstrated cognitive scalability, answered domain questions, described viewpoints
  - Analysis reveals disconnected models, incomplete strategies, strategy achievability
  - Analysis helped to evaluate and compare technology options, alternative system features, and knowledge management interventions
  - Evidence of model and domain knowledge iteration
  - Models were too large importance of scoping
- □ Exploratory Experiment comparing ad-hoc vs. systematic forward analysis
  - Systematic analysis provoked model changes and questions beyond ad-hoc analysis

# Framework Validation: Forward and Backward Implementation

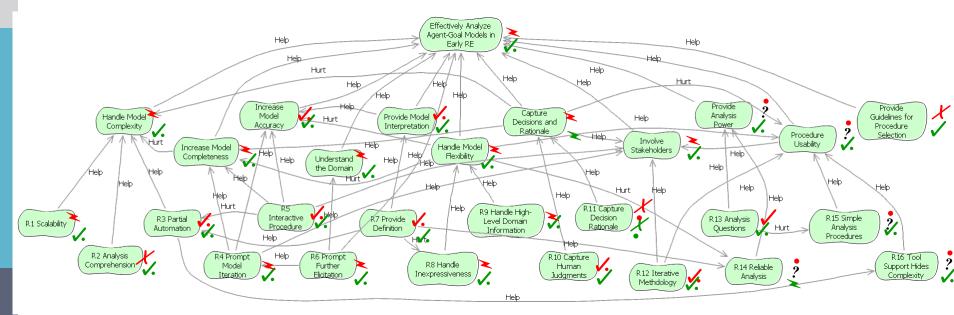
- □ Individual Case Studies (10 students with i\* experience)
  - Compare effects of ad-hoc vs. systematic forward and backward analysis
  - Five follow-up studies to test visualizations
- Group Case Study: inflo modeling tool
- **Quantitative and qualitative analysis of model changes, audio recordings, and video**
- □ Results
  - Participants had some difficulty analyzing large models created by others
  - No significant difference in the model changes and questions asked using ad-hoc vs. systematic analysis
  - Systematic analysis was well-received and generally usable
  - Systematic analysis increased analysis coverage, interpretation consistency, and revealed model incompleteness
  - Group analysis lead to some interesting discussion and model iteration
  - Needed domain-driven questions lead to creation of methodology (sanity checks)
- Conclusion
  - Analysis is most effective when participants are motivated by realistic situations and driving domain questions



Horkoff, Ghose & Yu, PoEM'10, REV'10 ent-Goal Models for Early RE Horkoff et al.

# **Framework Contributions**

Recluiting Syta Engine Action Conflicts



Combination of existing approaches

Contributions of current Framework

mputer Science

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# Limitations

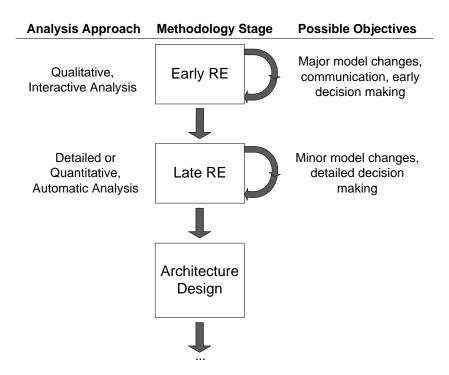
- Goal Modeling Limitations
  - Scalability, comprehension, learning curve
- □ Generalizability
  - Applied to i\*. Other goal modeling frameworks?
- Validation Results
  - Mixed results
  - Model iteration and elicitation primarily occurred in realistic studies with experienced modelers
  - Results due to interactive analysis or careful examination of model?
  - Fully-automatic provoke same benefits? Level of automation/interactivity
- Validation Study Design

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- Realistic domain and motivation hard to measure benefits
- Controlled study hard to emulate realistic domain and motivation

## Future Work

- Additional Framework Features
  - Judgment Rationale and Assumptions
  - Varying Levels of Automation
  - Handling Iteration over Models and Analysis Results
  - Further visualizations
- Future Directions
  - From Early to Late RE
  - Confidence in Analysis Results
  - Analysis of uncertain models





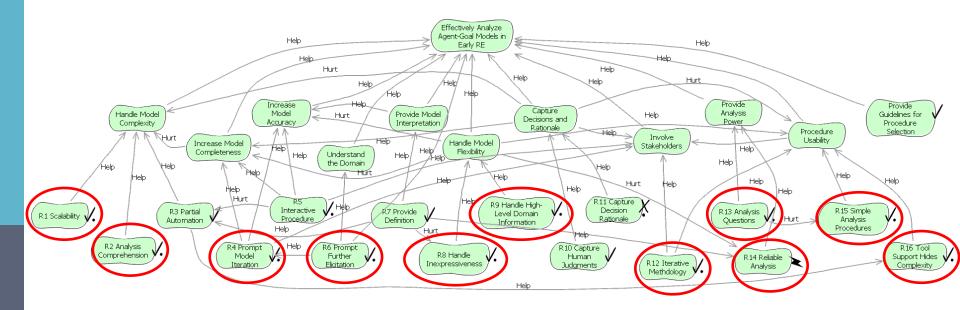
# Thank you

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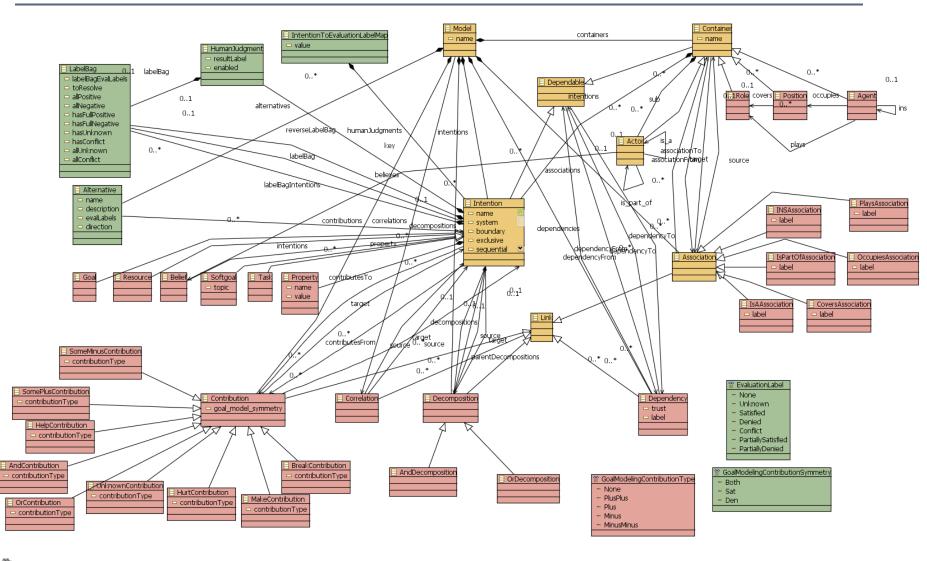
# Validation Contributions

Contributions of validation studies to the Requirements for analysis of Agent-Goal Models in Early RE





# Tool Support: Metamodel



Computer Science UNIVERSITY OF TORONTO

# **Tool Support: Scalability Tests**

Time Measures		Model 1		Model	2	Model 3				
	Ī	Alt 1	Alt 2	Alt 1	Alt 2	Alt 1	Alt 2		imin	
Num Judgments in Analysis		2	2	15	15	23	22		nd St	
Num Intentions receiving Judgments		2	2	9	9	16	16		esult	
Max Judgment Time		4.109	4.875	5.813	6.390	19.734	15.078		orwa	
Min Judgment Time		2.750	4.297	2.531	2.141	2.718	2.969	K	uns	
Average Judgment Time		3.429	4.586	4.328	3.930	8.048	6.296			
Total Judgment Time		6.859	9.172	64.922	58.954	185.106	138.517			
Total Computation Time		0.25	0.156	1.547	3.499	3.347	3.436			
Total Analysis Time	Ti	me Meas	ures			Model 1		Model 2	Model 2	
						Alt 1	Alt 2	Alt 1	Alt 2	
	Nι	ım Judgm	ents in A	nalysis	5	3	4	2		
Timing (Seconds)	Nι	um Intenti	ons recei	ving Judg	gments	2	2	1	2	
and Statistic	M	ax Judgm	ent Time			9.594	13.078	145.453	36.219	
Results for	M	in Judgme	ent Time			3.047	2.062	2.032	12.813	
Backward Analysis	verage Jud	lgment Ti	me		7.187	25.906	55.523	24.516		
Runs	То	tal Judgm	ent Time		35.937	8.635	222.094	49.032		
	Nι	ım Non-jı	Idgment	Messages	2	2	4	1		
	Total Time for Non-judgment Messages						9.077	72.220	2.265	
Computer Science	То	tal Comp	utation T	ime	0.579	17.616	30.905	1.047		
UNIVERSITY OF TOPONTO	1						1			

Total Analysis Time

g (Seconds) tatistic ts for rd Analysis

Model 3

Alt 2

40.547

4.438

18.162

90.814

49.984

150.765

291 563

4

5

2

Alt 1

9.766

9.766

9.766

9.766

3.437

2.391

41 312 35 328 325 219 52 344 15 594

