#### **Tutorial:**

Procesamiento de Lenguaje Natural en Ingeniería de Requisitos: Contribuciones Potenciales y Desafíos de Investigación

CIBSE 2015 - Lima, Peru

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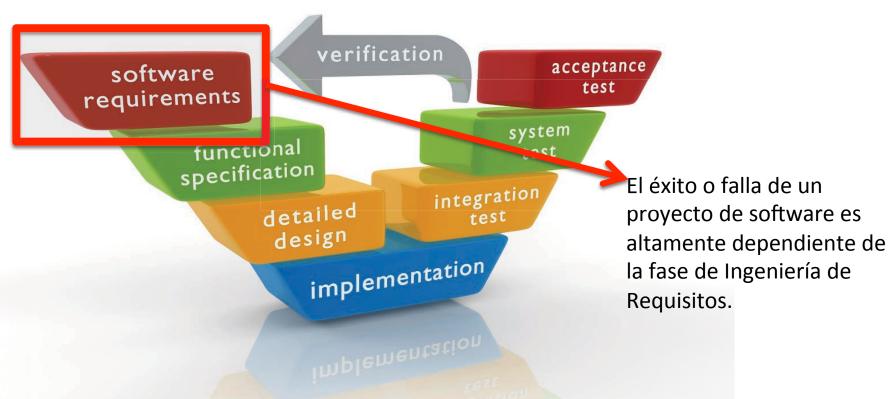
#### **Indice**

- Ingeniería de Requerimientos (RE)
- Procesamiento de Lenguaje
   Natural (NLP)
- Avances en la relación entre RE
   & NLP
- Contribuciones Potenciales
- Desafíos de Investigación

## Después de este tutorial, podrás :

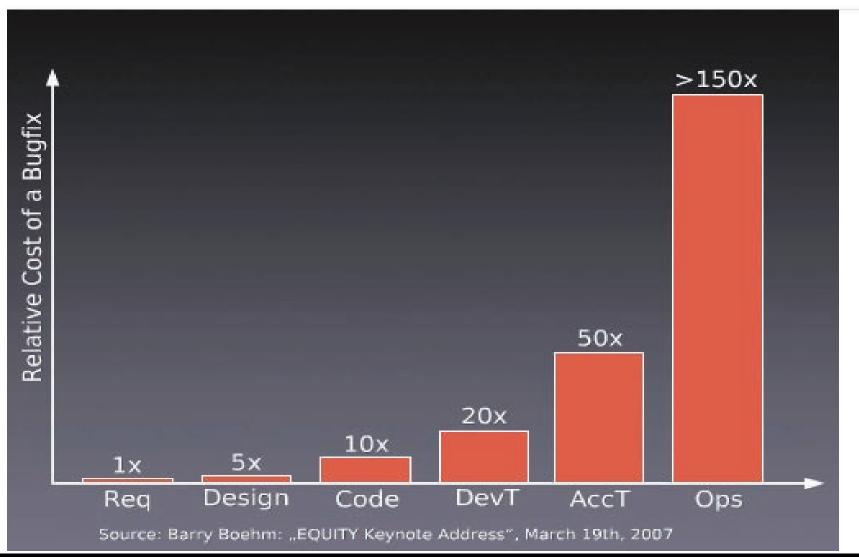
- Definir RE y su describir su importancia.
- Definir NLP y sus áreas de acción.
- Conocer el estado del arte en la relación entre NLP y RE.
- Describir las contribuciones potenciales de NLP para RE.
- Listar los principales desafíos para la investigación en RE.

#### **Software Engineering Process**

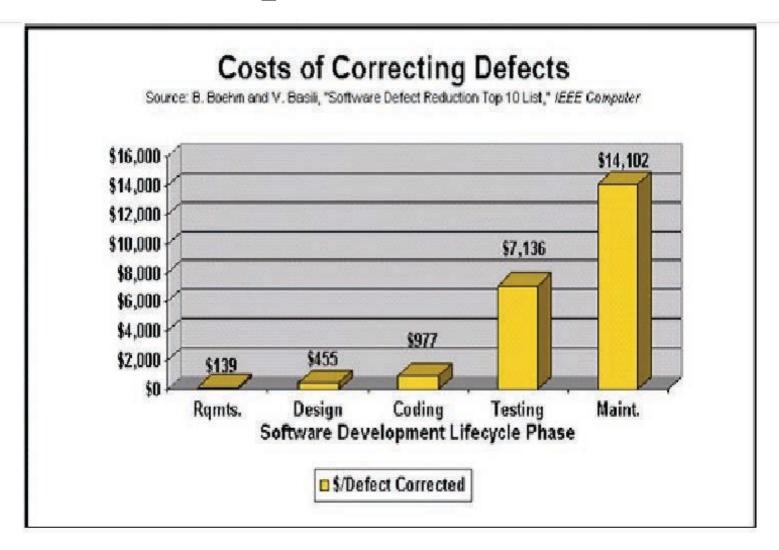


• Es una división del trabajo de desarrollo de software en distintas fases (o etapas) que tienen actividades las cuales persiguen una mejor planificación y administración.

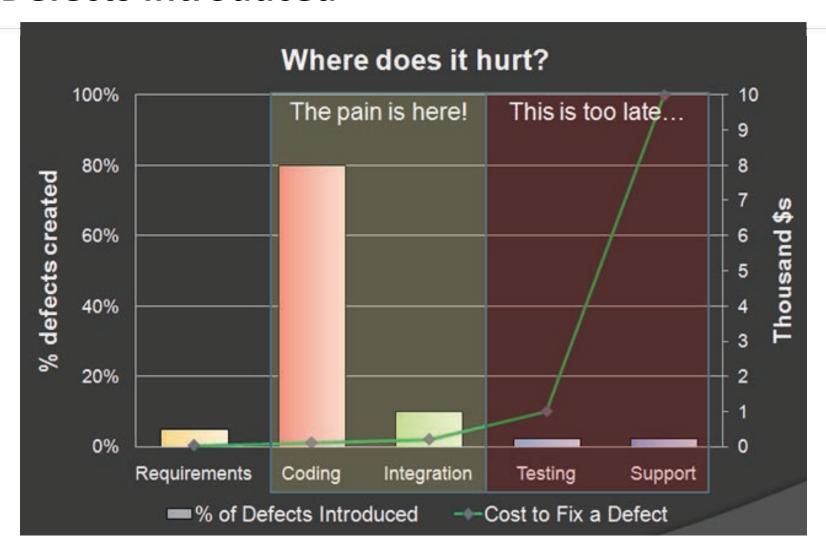
## **Increasing Cost of Defects**



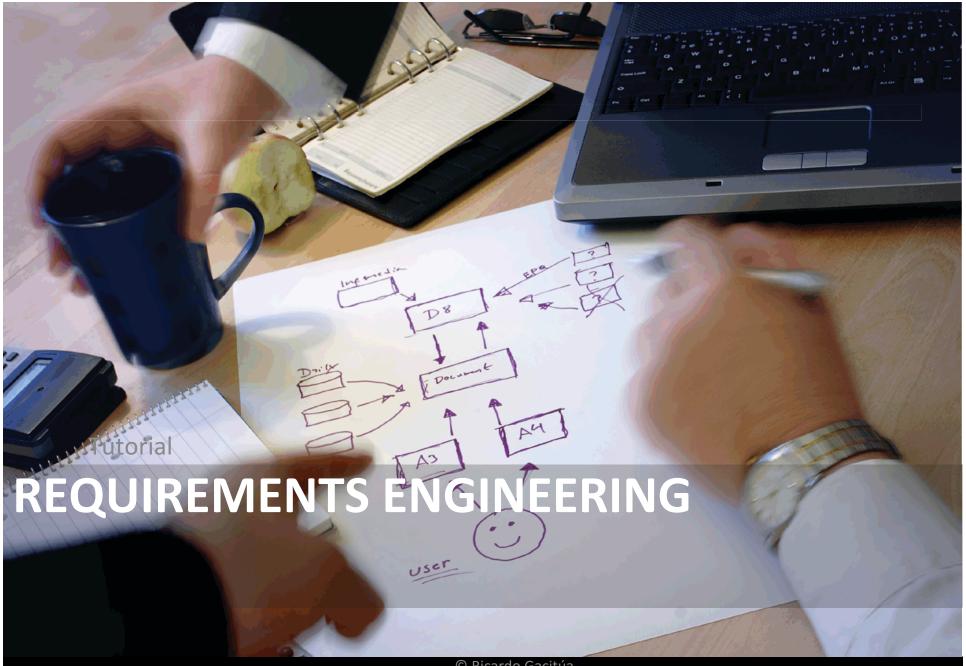
#### **Cost of Correcting Defects**



#### **Defects Introduced**



Graphs taken from: Benefit From Unit Testing in The real world.



Requirements Engineering

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### **Requirements Engineering**

**DEFINITION** Requirements engineering is the branch of software engineering concerned with the real-world goals for, functions of, and constraints on software systems. It is also concerned with the relationship of these factors to precise specifications of software behavior, and to their evolution over time and across software families.

Zave, \ (1997). Classification of Research Efforts in Requirements Engineering. ACM Computing Surveys, 29(4): 315-321.

#### "Real-world goals" represent:

- the 'why' as well as
- the 'what' of a system.

# <u>"Precise specifications"</u> provide the basis for:

- Analysing requirements,
- Validating that they are indeed what stakeholders want,
- Defining what designers have to build, and
- Verifying that they have done so correctly upon delivery.

# "Evolution over time and across software families."

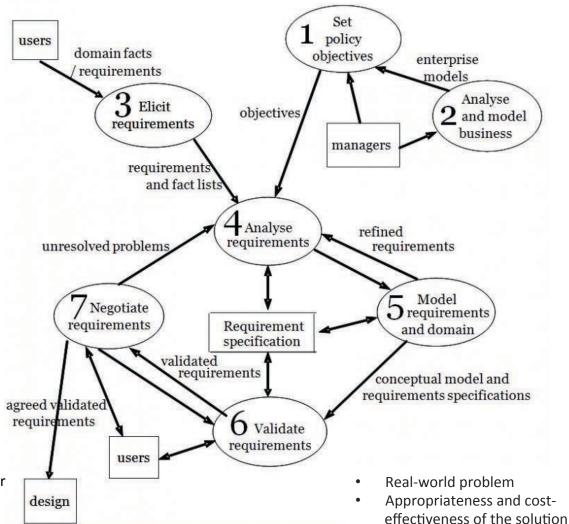
a changing world and the need to reuse partial specifications.

Requirements Engineering

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#### **Requirements Engineering Process**

 The use of the term engineering in RE serves as a reminder that RE is an important part of an engineering process.



Sutcliffe, Alistair G. (2014): Requirements Engineering. In: Soegaard, Mads and Dam, Rikke Friis (eds.). "The Encyclopedia of Human-Computer Interaction, 2nd Ed.". Aarhus, Denmark:

#### RE draws on the cognitive and social sciences

This provides an understanding of the difficulties people may have in describing their needs

Cognitive psychology

Anthropology

This provides a methodological approach to observing human activities

This provides an understanding of the political and cultural changes caused by computerisation

Sociology

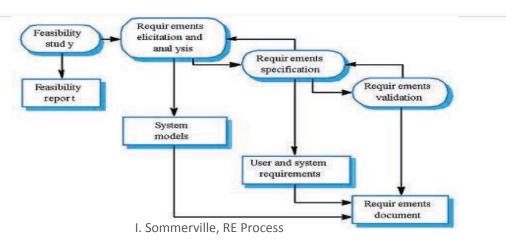
Linguistics

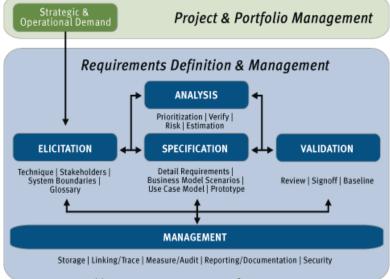
This is important because RE is largely about communication

The context in which RE takes place is usually a human activity system, and the problem owners are people.

### **Requirements Engineering Process**

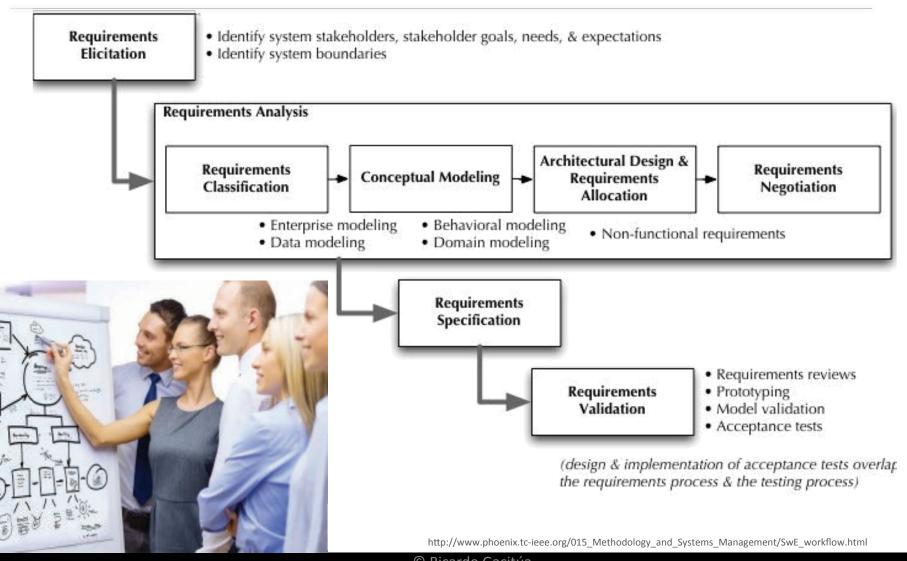


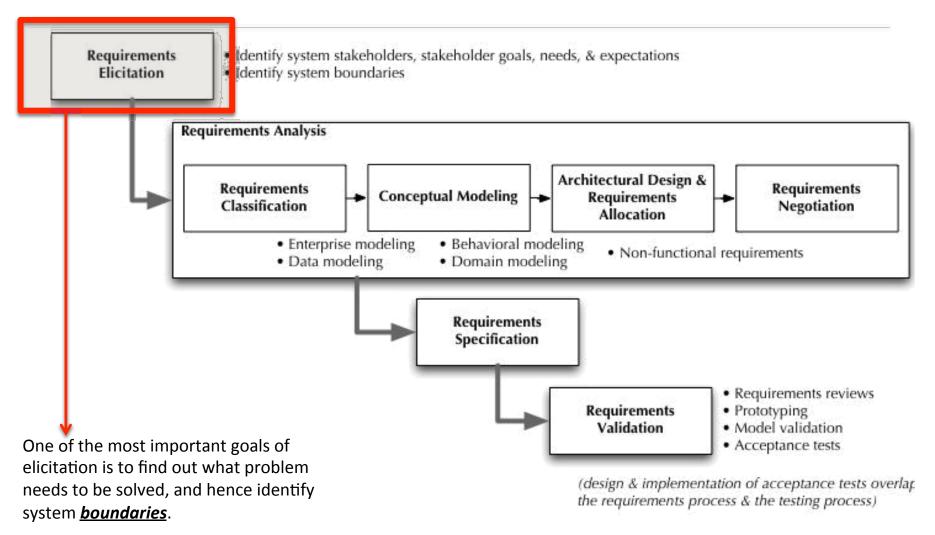




Micro Focus CaliberRM™ - Enterprise Software Requirements Management System

### **Requirements Engineering Process**





### **Requirements Elicitation Problems**

#### Problems of scope

- the boundary of the system is ill-defined
- unnecessary design information may be given

#### Problems of understanding

- users have incomplete understanding of their needs
- users have poor understanding of computer capabilities and limitations
- analysts have poor knowledge of problem domain
- user and analyst speak different languages
- ease of omitting "obvious" information
- conflicting views of different users
- requirements are often vague and untestable, e.g., "user friendly" and "robust"

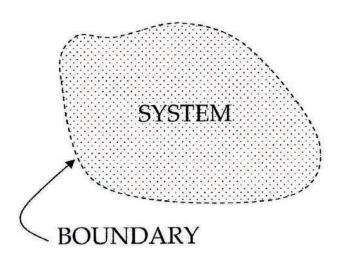
#### Problems of volatility

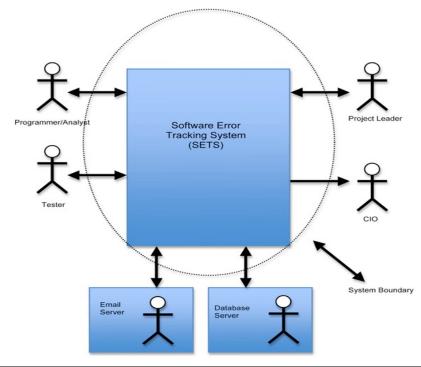
requirements evolve over time

• A) BOUNDARIES - Identifying and agreeing a system's boundaries affects all subsequent elicitation efforts.

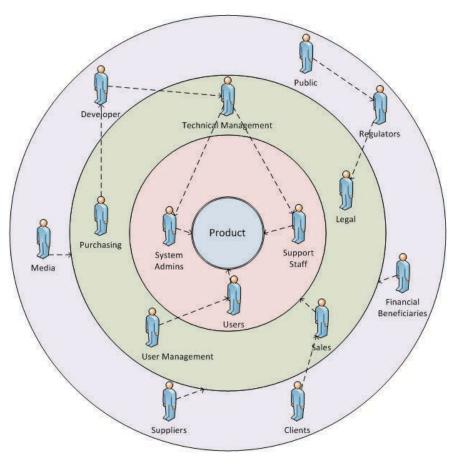
The identification of stakeholders and user classes, of goals and tasks,
 and of scenarios and use cases all depend on how the boundaries are

chosen.



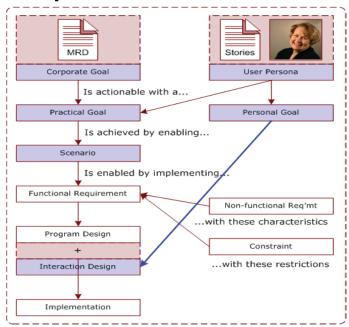


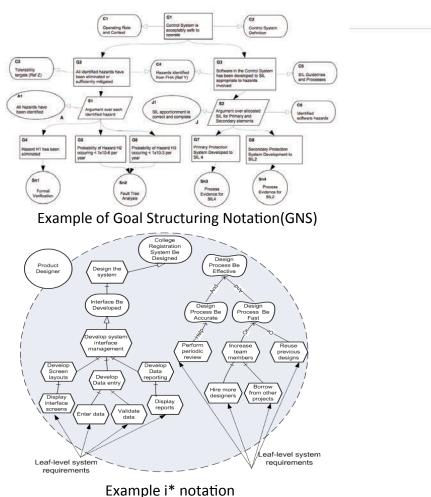
- B) IDENTIFYING STAKEHOLDERSindividuals or organisations
  who stand to gain or lose from
  the success or failure of a
  system is also critical.
  - Stakeholders include:
    - customers or clients (who pay for the system),
    - developers (who design, construct and maintain the system), and
    - users (who interact with the system to get their work done).



Alexander, I., "Stakeholders: who is your system for?," *Computing & Control Engineering Journal*, vol.14, no.2, pp.22,26, April/May 2003

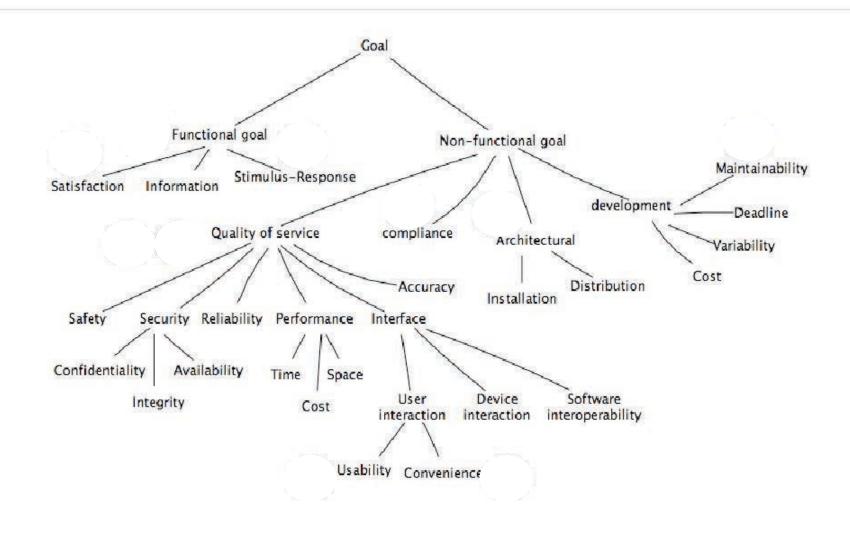
 C) GOALS - denote the objectives a system must meet.





Eliciting goals focuses the requirements engineer on the problem domain and the needs of the stakeholders, rather than on possible solutions to those problems.

## **RE Process -> Elicitation- Goal categorization**

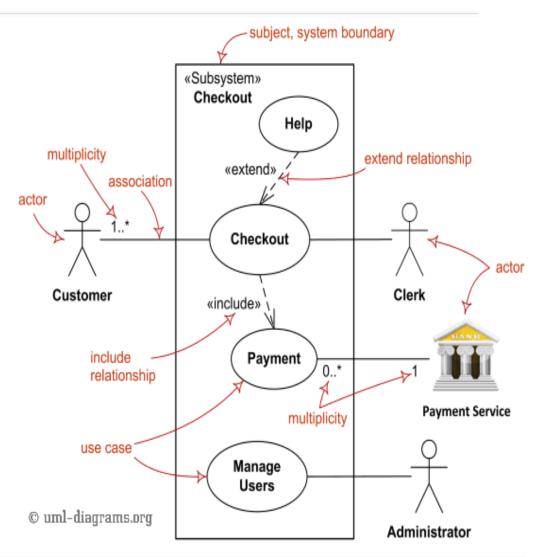


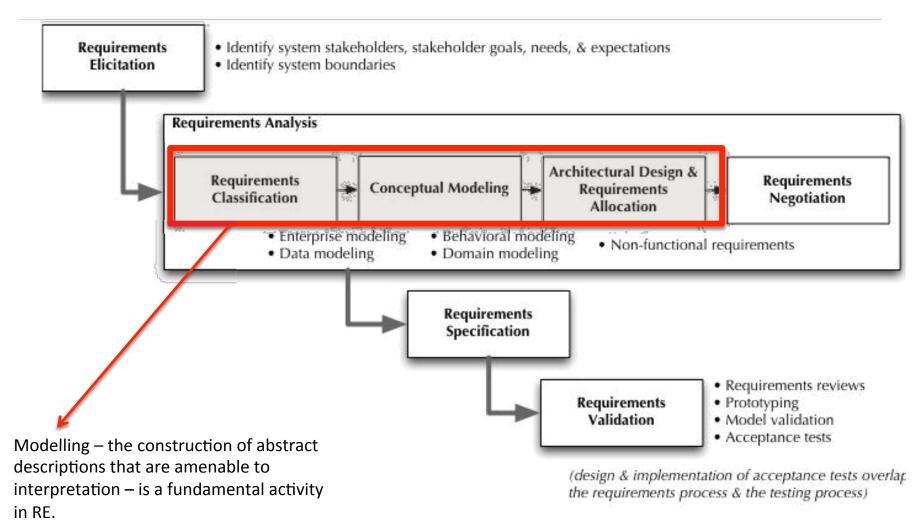
# RE Process -> Elicitation - Use Cases to represent user tasks

 It is often the case that users find it difficult to articulate their requirements.

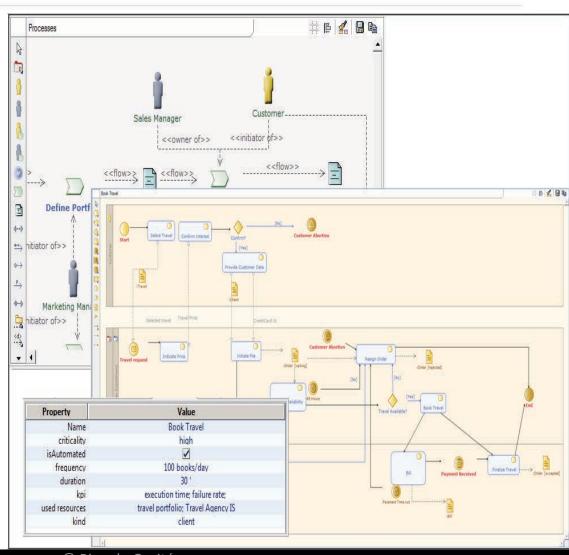
RE

 A requirements engineer can resort to eliciting information about the tasks users currently perform and those that they might want to perform.

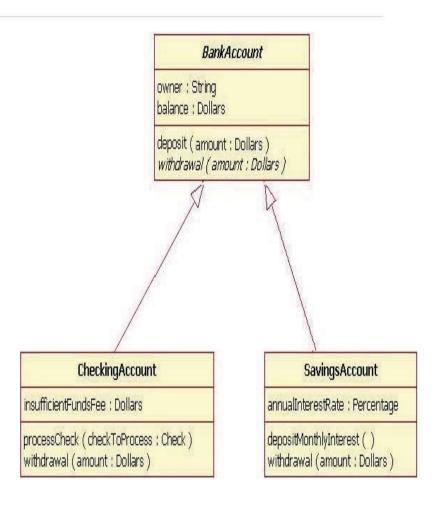




- A) Enterprise
   Modelling Enterprise modelling
   and analysis deals
   with:
  - understanding an <u>organisation's</u> <u>structure</u>;
  - the <u>business rules</u> that affect its operation;
  - the <u>goals, tasks and</u>
     <u>responsibilities</u> of its
     constituent members;
     and
  - the <u>data</u> that it needs, generates and manipulates.

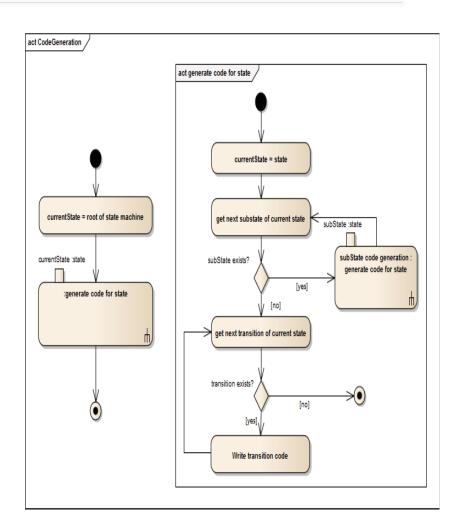


- B) Data Modelling -
  - Decisions need to be made about what information the system will need to represent, and how the information held by the system corresponds to the real world phenomena being represented.
    - Data modelling provides the opportunity to address these issues in RE.



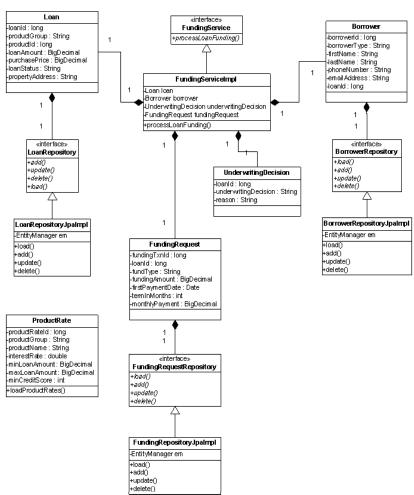
#### C) Behavioural Modelling

 Modelling requirements often involves modelling the dynamic or functional behaviour of stakeholders and systems, both existing and required

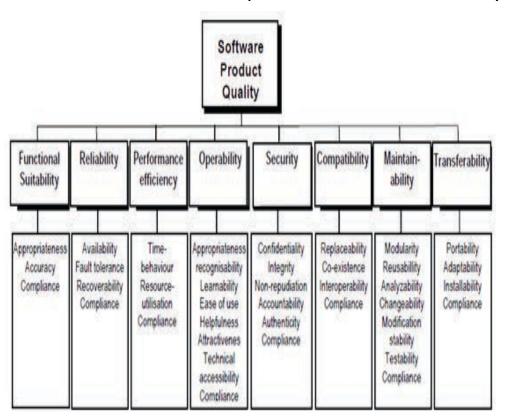


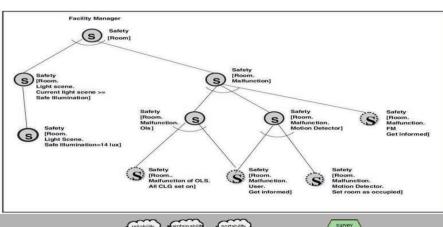
**Loan Processing Application Domain Model** 

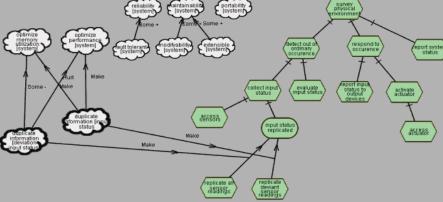
• D) **Domain Modelling** - A significant proportion of the RE process is about developing domain descriptions. A model of the domain provides an abstract description of the world in which an envisioned system will operate.



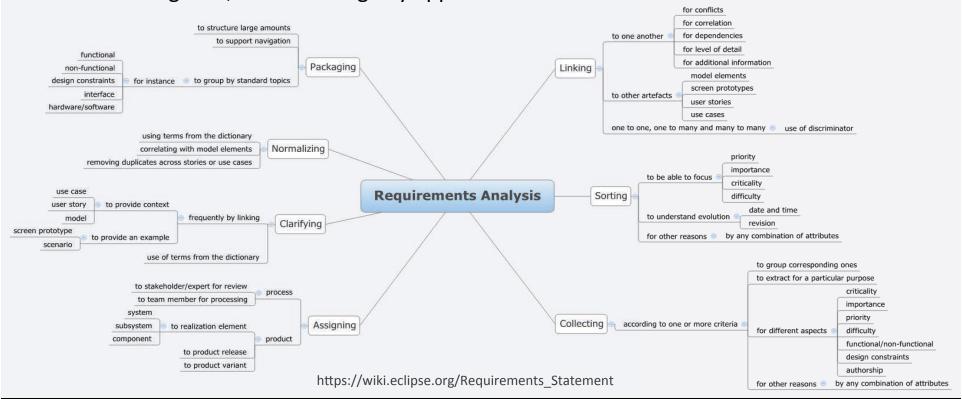
- E) Modelling Non-Functional Requirements (NFRs)
- Non-functional requirements (also known as quality requirements) are generally more difficult to express in a measurable way, making them more difficult to analyse



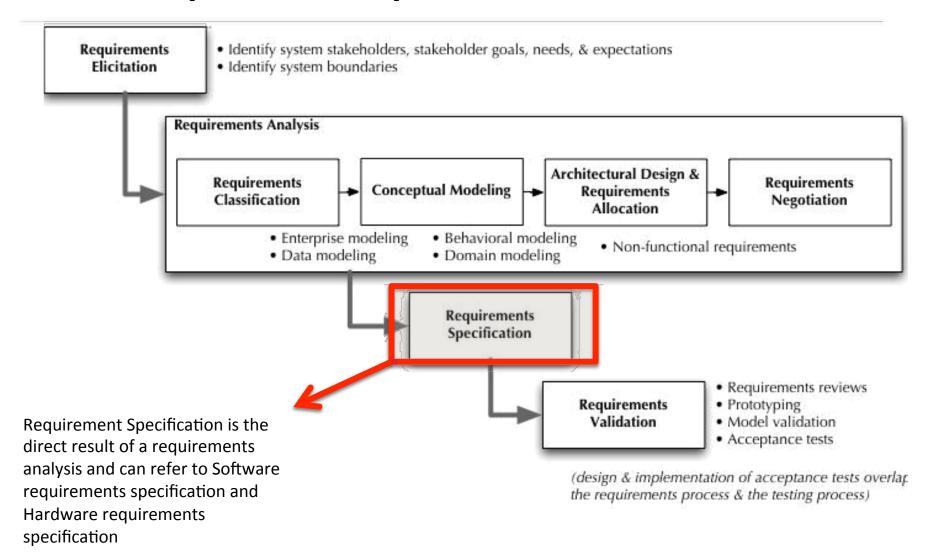




• F) **Analysing Requirements Models** - A primary benefit of modelling requirements is the opportunity this provides for analysing them determining whether the stated requirements are clear, complete, consistent and unambiguous, and resolving any apparent conflicts.

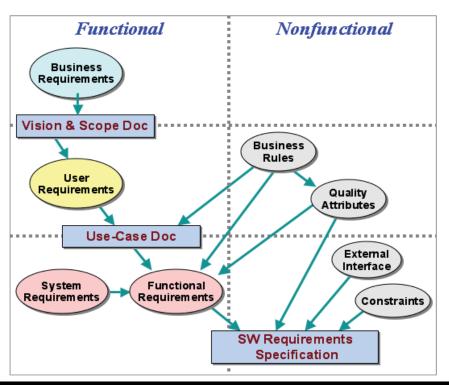


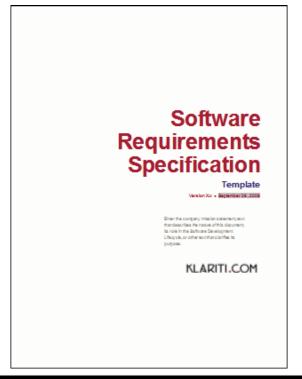
#### **RE -> Requirement Specification**



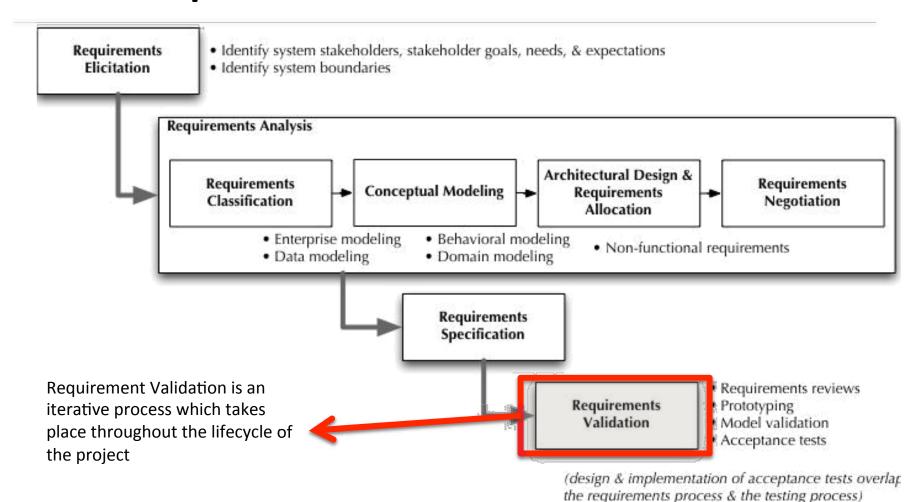
## RE - Software requirements specification (SRS)

- A software requirements specification (SRS) is a comprehensive description of the intended purpose and environment for software under development. The SRS fully describes what the software will do and how it will be expected to perform.
- The process of writing down the user and system requirements in a requirements document.





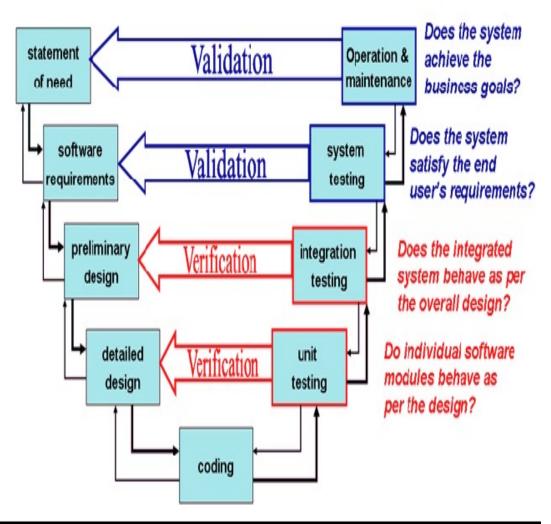
#### **RE -> Requirements Validation**



#### **RE -> Requirements Validation**

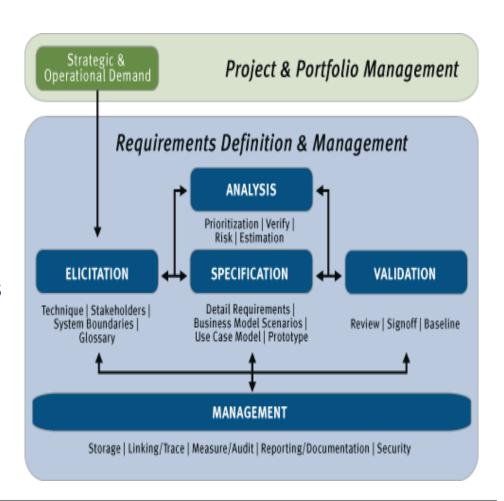
#### Boehm's definition:

- Validation: Am I building the right system?
- Verification: Am I building the system right?
- Requirements validation is difficult for two reasons:
  - Philosophical in nature, and concerns the question of truth and what is knowable.
  - Social, and concerns the difficulty of reaching agreement among different stakeholders with conflicting goals.



## **RE - > Additional - Evolving Requirements**

- Managing change is a fundamental activity in RE.
  - Changes to requirements documentation
    - Traceability links to monitor and control the impact of changes in different parts of the documentation.
  - Managing inconsistency in requirements specifications as they evolve is a major challenge.



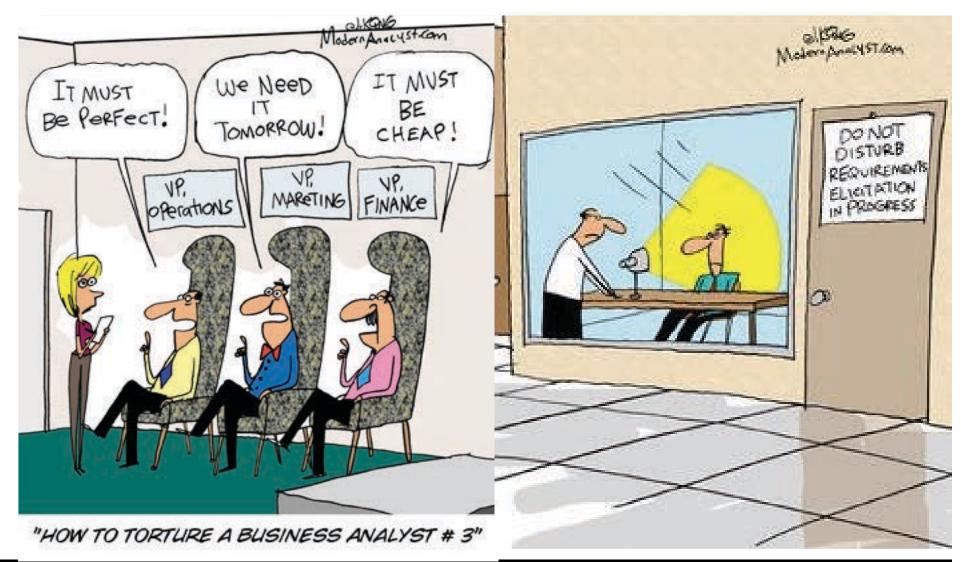
## **RE** - requirements validation session



#### **RE Summary**

- **1. SOCIAL SCIENCES:** RE draws on the cognitive and social sciences:
  - Cognitive Psychology, Anthropology, Sociology, Linguistics
- **2. BOUNDARIES:** The identification of stakeholders and user classes, of goals and tasks, and of scenarios and use cases all depend on how the boundaries are chosen.
- **3. MODELS:** Models can be used to represent a whole range of products of the RE process.
- **4. COMMUNICATION:** RE is not only a process of discovering and specifying requirements, it is also a process of facilitating effective communication of these requirements among different stakeholders.
- **5. DOCUMENTATION:** The way in which requirements are documented plays an important role in ensuring that they can be read, analysed, (re-)written, and validated.
- **6. VALIDATION:** Requirements validation is difficult for two reasons.
  - The first reason is philosophical in nature, and concerns the question of truth and what is knowable.
  - The second reason is social, and concerns the difficulty of reaching agreement among different stakeholders with conflicting goals.
- 7. CHANGES: Managing changing requirements is not only a process of managing documentation, it is also a process of recognising change through continued requirements elicitation, re- evaluation of risk, and evaluation of systems in their operational environment.

#### **RE Humour**



#### **Tutorial:**

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# **Indice**

- Ingeniería de Requerimientos (RE)
- Procesamiento de Lenguaje
   Natural (NLP)
- Avances en la relación entre RE
   & NLP

Contribuciones Potenciales Desafíos de Investigación

# Después de este tutorial, podrás :

- Definir RE y su describir su importancia.
- Definir NLP y sus áreas de acción.
- Conocer el estado del arte en la relación entre NLP y RE.
  - Describir las contribuciones potenciales de NLP para RE.

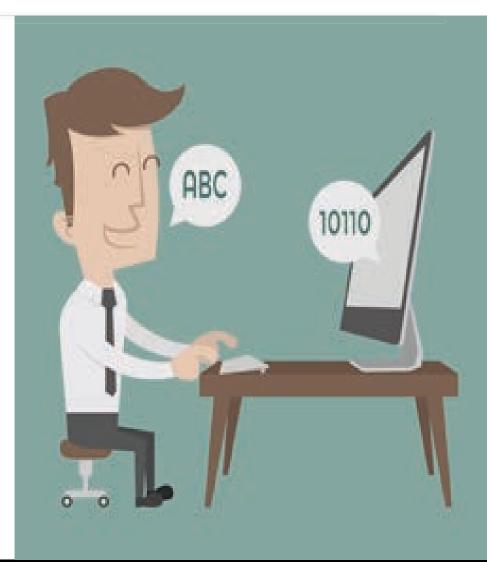
# Can computational methods aid the understanding of human language?





# **Motivation - Natural Language Processing (NLP)**

- NLP is the branch of computer science focused on developing systems that allow computers to communicate with people using everyday language.
- Also called Computational Linguistics
  - Also concerns how computational methods can aid the understanding of human language.

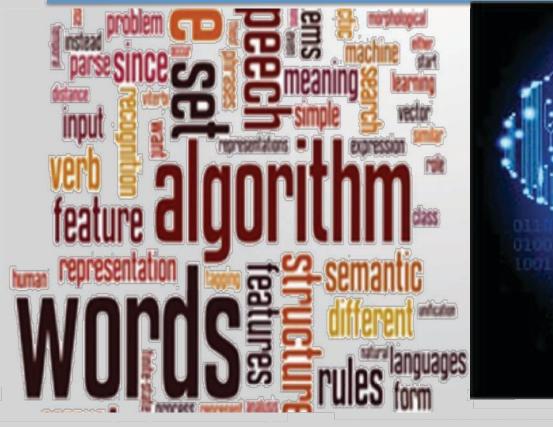


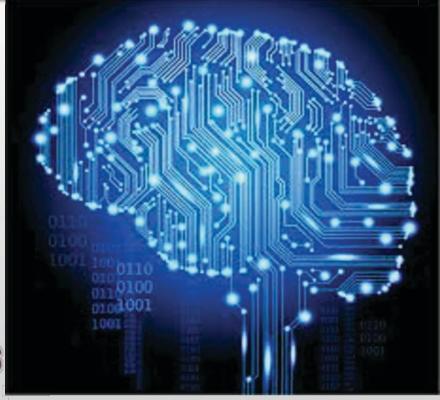
RE Survey Challenges

# **Natural Language Processing**



**DEFINITION** Natural language processing (NLP) is a field of computer science, artificial intelligence, and computational linguistics concerned with the interactions between computers and human (natural) languages.



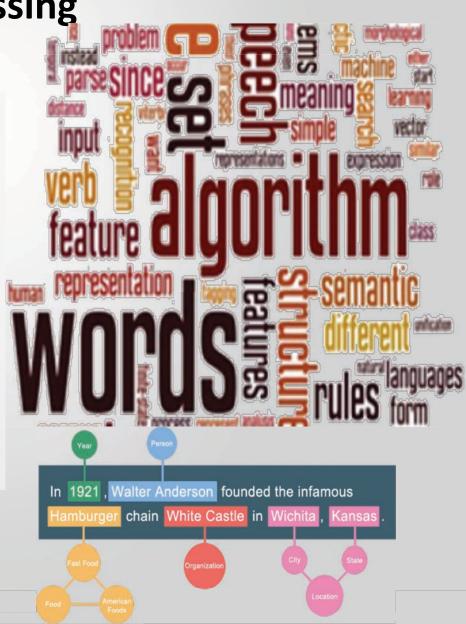


**Natural Language Processing** 

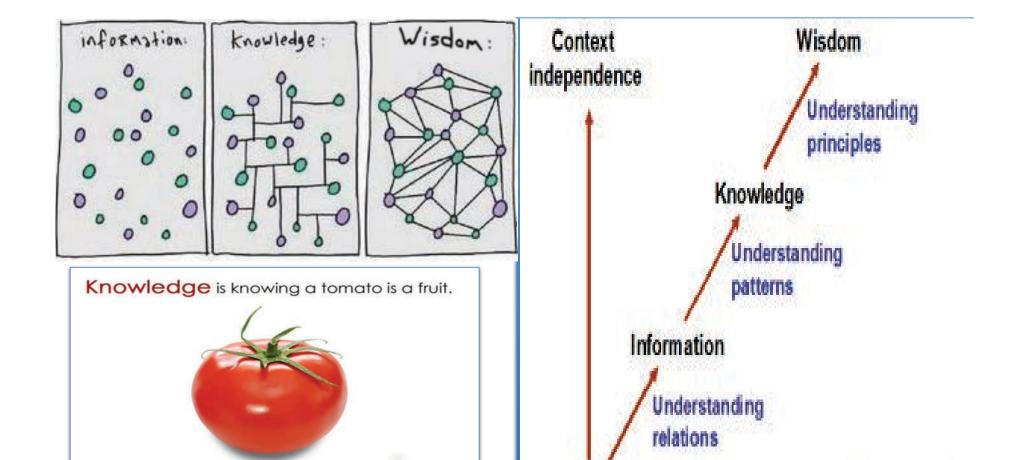
**NLP** is a field of computer science, artificial intelligence, and linguistics concerned with the interactions between computers and human (natural) languages.

- It is the analysis of human language so that computers can understand natural languages as humans do.
- Many challenges in NLP involve natural language understanding, that is, enabling computers to derive meaning from human or natural language input, and others involve natural language generation.

Christopher D. Manning and Hinrich Schütze (1999). Foundations of Statistical Natural Language Processing. The MIT Press.

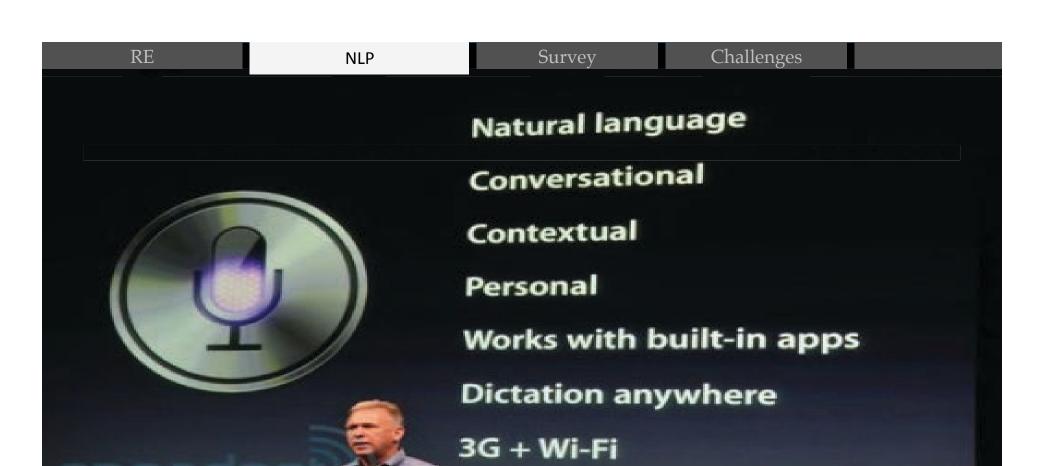


# Information, Knowledge and Wisdom

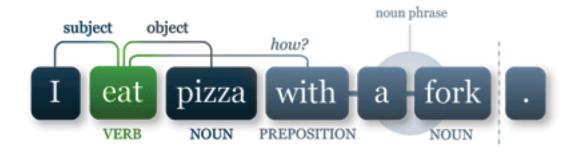


Wisdom is not putting it in a fruit salad.

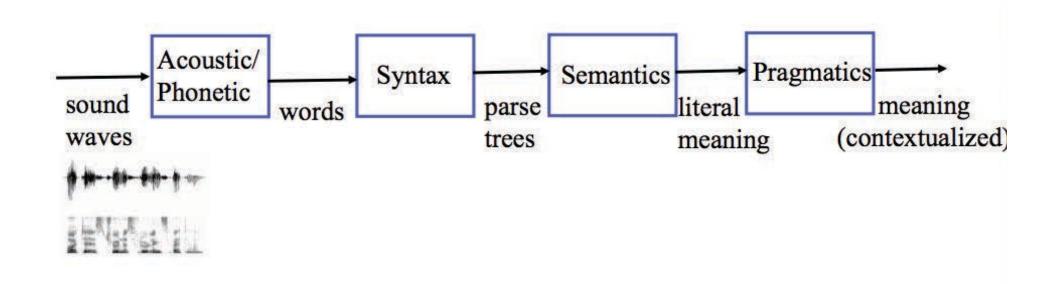
Understanding



# **Natural Language Processing**

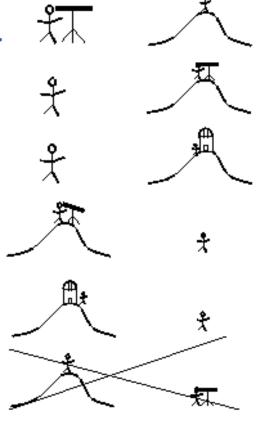


# **Modular Comprehension**



# **Ambiguity**

- Natural language is highly ambiguous and must be *disambiguated*.
  - I saw the man on the hill with a telescope.



NLP

Survey

Challenges

### Automatic Summarization

Produce a readable summary of a chunk of text

### Co-reference Resolution

Given a sentence, determine which words ("mentions") refer to the same objects ("entities")

### Discourse **Analysis**

To identifying the discourse structure of connected text.(i.e. elaboration. explanation, contrast)."

### Machine Translation

Automatically translate text from one human language to another.

Part-ofspeech tagging

Given a sentence, determine the part of speech for each word

### **Parsing**

Determine the parse tree (grammatical analysis) of a given sentence

### Morphological Segmentation

Separate words into individual morphemes and identify the class of the morphemes.

Major tasks in NLP

### **Named Entity** Recognition (NER)

Given a stream of text, determine which items in the text map to proper names

### Natural Language **Understanding**

Convert chunks of text into more formal representations such as first-order logic

### Optical character recognition

Given an image representing printed text, determine the corresponding text.

### Question **Answering**

Given a humanlanguage question, determine its answer

### Relationship Extraction

Given a chunk of text, identify the relationships among named entities

### Sentence breaking

Given a chunk of text, find the sentence boundaries.

### Sentiment **Analysis**

Extract subjective information to determine "polarity" have to select the

### Word sense disambiguation

Many words have more than one meaning; we about specific objectsmeaning which makes the most sense in context

### Speech Segmentation

Given a sound clip of a person or people speaking, separate it into words

### Topic Segmentation and recognition

Given a chunk of text, separate it into segments each of which is devoted to a topic.

### Word Segmentation

Separate a chunk of continuous text into separate words.

### Speech Recognition

Given a sound clip of a person speaking, determine the textual representation of the speech

Retrieval

This is concerned with storing, searching and retrieving information

# Extraction

This is concerned in general with the extraction of semantic information from text

### Automatic Summarization

Produce a readable summary of a chunk of text

Given a sentence, determine which words ("mentions") refer to the same objects ("entities"

### Discourse **Analysis**

To identifying the discourse structure of connected text, i.e. elaboration. explanation, contrast)."

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Separate words into individual morphemes and identify the class of the morphemes.

Named Entity Recognition (NER)

Given a stream of text, determine which items in the text map to proper names

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Convert chunks of text into more formal representations such as first-order logic

Given an image representing printed text, determine the corresponding text.

Question **Answering** 

Given a humanlanguage question, determine its answer Relationship Extraction

Given a chunk of text, identify the relationships among named entities

Sentiment **Analysis** 

Extract subjective information to determine "polarity" about specific objects

Word sense disambiguation

Many words have more than one meaning; we have to select the meaning which makes the most into words sense in context

Given a sound clip of

Speech Recognition Given a sound clip of a person speaking, determine the textual representation of the speech

Information Extraction

This is concerned in general with the extraction of semantic information from text

# **NLP-> Automatic Discourse Analysis**

- "The study of well-established meanings or ideas around a topic which shape how we can [think and] talk about it." Stephanie Taylor's (2010)
- For example: Segmentation by thematic

### BACKGROUND KNOWLEDGE:

Recent studies indicate ...

- ... the previously proposed ...
- ... is universally accepted ...

### CONRASTING IDEAS:

... unorthodox view resolves ... paradoxes ...

In contrast with previous hypotheses ...

... inconsistent with past findings ...

### GENERALIZING:

importance ...

... emerging as a promising approach
Our understanding ... has grown
exponentially ...
... growing recognition of the

### NOVELTY:

... new insights provide direct evidence ..... we suggest a new ... approach ...

... results define a novel role ...

### SIGNIFICANCE:

studies ... have provided important advances

Knowledge ... is crucial for ...

understanding

valuable information ... from studies

### SURPRISE:

We have recently observed ... surprisingly We have identified ... unusual The recent discovery ... suggests intriguing roles

### OPEN QUESTION:

... little is known ...

... role ... has been elusive

Current data is insufficient ...

### SUMMARIZING:

The goal of this study ...

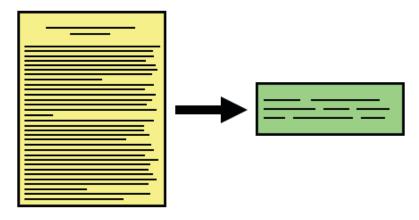
Here, we show ...

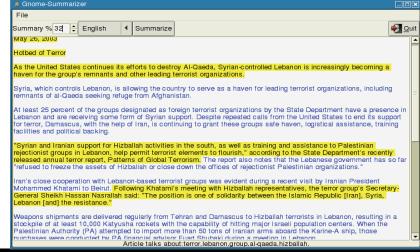
Altogether, our results ... indicate

# **NLP -> Text Summarization**

- Text summarisation
  - a reductive transformation of source text to summary text through content reduction by selection and/or generalisation on what is important in the source. (Sparck Jones, 1999)
  - the process of distilling the most important information from a source (or sources) to produce an abridged version for a particular user (or users) and task (or tasks). (Mani and Maybury, 1999)

Automatic text summarisation = The process of producing summaries automatically.





# **NLP - > Machine Translate**

- Automatically translate text from one human language to another.
  - On a basic level, MT performs simple substitution of words in one natural language for words in another, but that alone usually cannot produce a good translation of a text because recognition of whole phrases and their closest counterparts in the target language is needed.
  - Solving this problem with corpus and statistical techniques is a rapidly growing field that is leading to better translations, handling differences in linguistic typology, translation of idioms, and the isolation of anomalies.

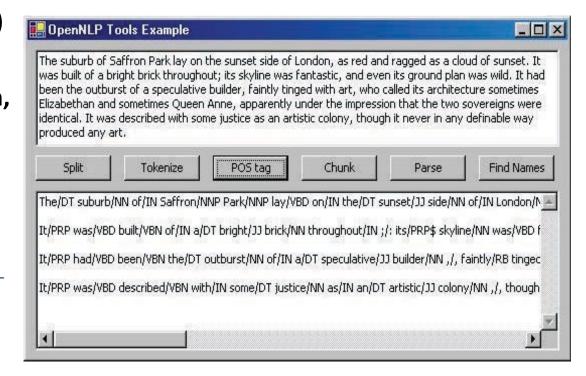


# **NLP - > Part-of-Speech Tagging**

### POS Tagging

- It is the process of marking up a word in a text (corpus) as corresponding to a particular part of speech, based on both its definition, as well as its context—i.e. relationship with adjacent and related words in a phrase, sentence, or paragraph.
  - A simplified form of this is commonly taught to schoolage children, in the identification of words as nouns, verbs, adjectives, adverbs, etc.





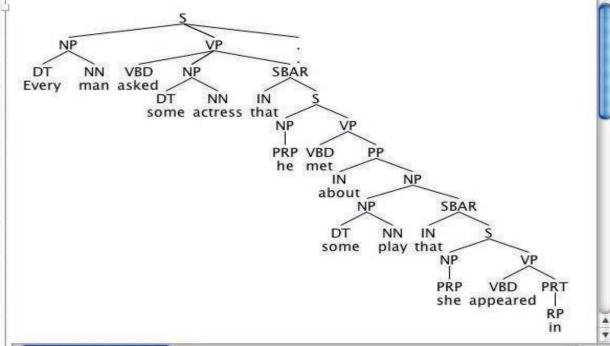
# **NLP - > Parsing**

# Parsing

Parsing or syntactic analysis is the process of analysing a string of symbols, either in natural language or in computer languages, according to the rules of a formal grammar.

Sentence Count: 317 Displayed Tree (Sentence): 287

((I (PRP)) (am (VBP)) (proud (II)) (that (IN)) (John (NNP) ((John (NNP)) (is (VBZ)) (tough (JJ)) (to (TO)) (please (VI ((It (PRP)) (is (VBZ)) (tough (JJ)) (to (TO)) (please (VB)) ((John (NNP)) (is (VBZ)) (likely (JJ)) (to (TO)) (win (VB)) ((Who (PRP)) (does (VBZ)) (his (PRP\$)) (mother (NN)) (I ((His (PRPS)) (mother (NN)) (loves (VBZ)) (everyone (NI ((Every (RB)) (man (NN)) (likes (NNS)) (some (DT)) (syn ((Every (RB)) (man (NN)) (asked (VBD)) (some (DT)) (ar ((Someone (NN)) (gave (VBD)) (every (JJ)) (actress (NN) ((Which (NNP)) (man (NN)) (liked (VBD)) (which (WDT)) ((John (NNP)) (seems (VBZ)) (that (IN)) (he (PRP)) (likes ((John (NNP)) (asked (VBD)) (Mary (NNP)) (about (IN)) ((We (PRP)) (like (IN)) (myself (PRP))) ((John (NNP)) (seems (VBZ)) (is (VBZ)) (crazy (NN))) ((John (NNP)) (tried (VBD)) (Bill (NNP)) (to (TO)) (seem ((John (NNP)) (tried (VBD)) (to (TO)) (be (VB)) (arrestec ((Mary (NNP)) (is (VBZ)) (proud (VBN)) (of (IN)) (Bill (NN ((John (NNP)) (seems (VBZ)) (that (IN)) (he (PRP)) (is (V ((The (DT)) (destruction (NN)) (Rome (NNP)) (worked ((The (DT)) (destruction (NN)) (of (IN)) (Rome (NNP)) ( ((The (DT)) (belief (NN)) (John (NNP)) (to (TO)) (be (VB ((The (DT)) (belief (NN)) (of (IN)) (John (NNP)) (to (TO)) ((The (DT)) (belief (NN)) (that (IN)) (John (NNP)) (is (VB ((John (NNP)) ('s (POS)) (belief (NN)) (to (TO)) (be (VB)) ((John (NNP)) (seems (VBZ)) (that (IN)) (his (PRP\$)) (be



# **NLP -> Named Entity Recognition (NER)**

# Automatically find names of people, places, products, and organizations in text

In 1917, Einstein applied the general theory of relativity to model the large-scale structure of the universe. He was visiting the United States when Adolf Hitler came to power in 1933 and did not go back to Germany, where he had been a professor at the Berlin Academy of Sciences. He settled in the U.S., becoming an American citizen in 1940. On the eve of World War II, he endorsed a letter to President Franklin D. Roosevelt alerting him to the potential development of "extremely powerful bombs of a new type" and recommending that the U.S. begin similar research. This eventually led to what would become the Manhattan Project. Einstein supported defending the Allied forces, but largely denounced using the new discovery of nuclear fission as a weapon. Later, with the British philosopher Bertrand Russell, Einstein signed the Russell-Einstein Manifesto, which highlighted the danger of nuclear weapons. Einstein was affiliated with the Institute for Advanced Study in Princeton, New Jersey, until his death in 1955.

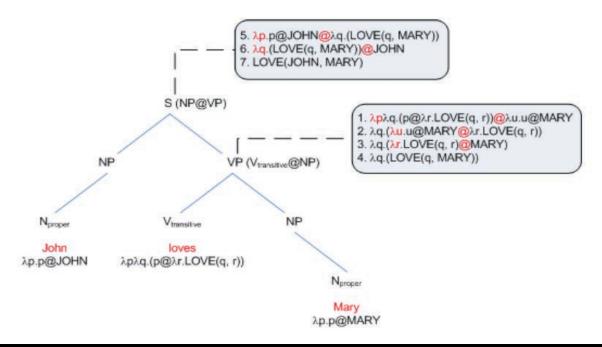
Tag colours:

LOCATION TIME PERSON ORGANIZATION MONEY PERCENT DATE

# **NLP -> Natural Language Understanding**

 Convert chunks of text into more formal representations such as first-order logic.

# (1) John loves Mary



# **NLP -> Natural Language understanding**

After two years of online Sating, John finally found the courage to Formalize his relationship... Ix I +x e D, x=myself, 3 Y Y I Y E B. Teppenson ZAAL

# NLP -> Question Answering

 Question Answering (QA) is a computer science discipline within the fields of information retrieval and natural language processing (NLP), which is concerned with building systems that automatically answer questions posed by humans in a natural language. 'William Wilkinson's 'An account of the Bram Stoke principalities of Wallachia and Moldavia' inspired this author's most famous novel



Watson showed itself to be imperfect, but researchers at I.B.M. and other companies are already developing uses for Watson's technologies that could have a significant impact on the way doctors practice and consumers buy products.

# **NLP** -> Question Answering









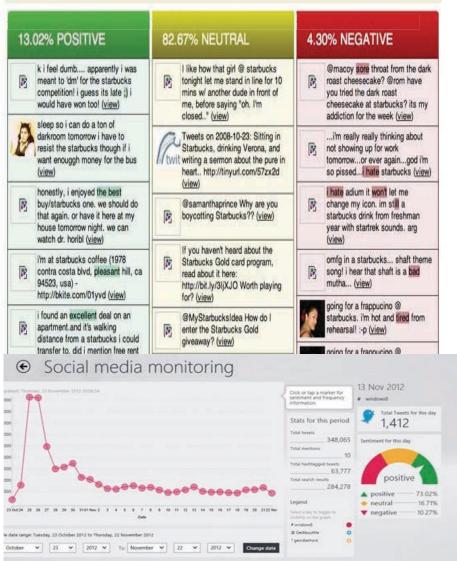
funding. The Emeryville, Calif. company, which was founded in Paris, plans to use the investment for growth in both the United States and Europe.

The company's virtual agents, or chatbots, are programmed to automate sales support for large and mid-sized businesses. By helping online shoppers resolve issues or make purchase decisions, VirtuOz helps web sites cut labor costs. The company claims that its agents, equipped with natural language processing (NLP) capabilities, can provide customer service to clients for one-tenth the cost of a traditional human support team.



# NLP - Sentiment Analysis starbucks

- It aims to determine the attitude of a speaker or a writer with respect to some topic or the overall contextual polarity of a document.
- The attitude may be his or her judgment or evaluation (see appraisal theory), affective state (that is to say, the emotional state of the author when writing), or the intended emotional communication (that is to say, the emotional effect the author wishes to have on the reader)



NEUTRAL TWEETS NEGATIVE TWEETS

234

5437

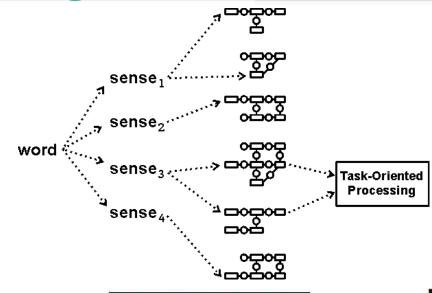
4495

708

# Word Sense Disambiguation

### WSD

- It is the process of identifying which sense of a word (i.e. meaning) is used in a sentence, when the word has multiple meanings.
- Word-sense disambiguation (WSD) is an open problem of natural language processing and ontology.







Bass: fish

???

Bass: instrument

RE Survey Challenges

# information extraction

One stop solution to extract structured information from unstructured data.



Lester Bowles 'Mike' Pearson, PC, Ami, oxford (23 April 1897 - 21 December 1972) was a Canadian professor, historian, c'wi servant, statesman, diplomat, and politician, who won the professor interpretation of the control of the con

**Information extraction** (IE) is the task of automatically extracting structured information from unstructured and/or semi-structured machine-readable documents.

Subject : curriculum meeting

Date : January 21, 2014

To: Ricardo Gacitua

Event: Curriculum mtg.

Date : Jan-22-2014

End: 11:30

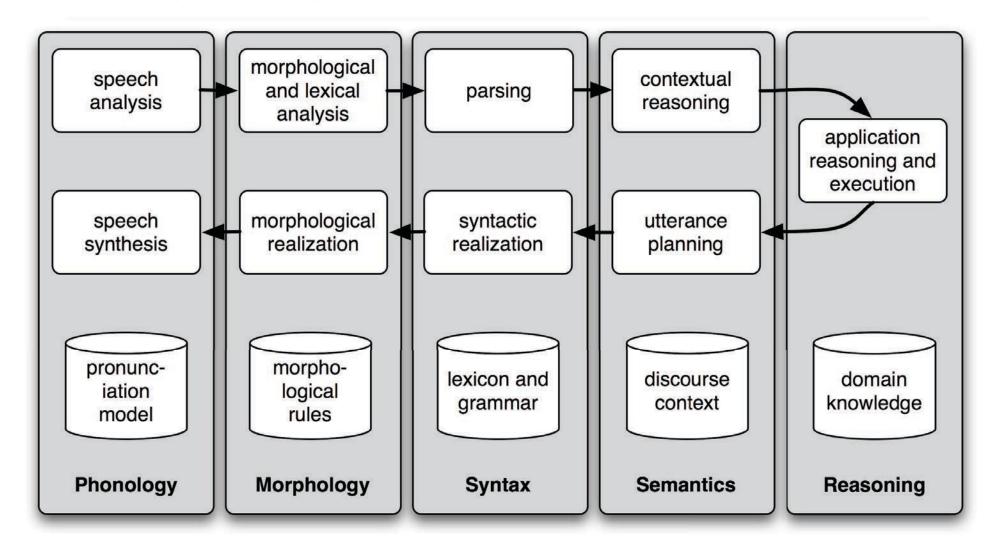
Start: 10:00 am

Where: gates 171

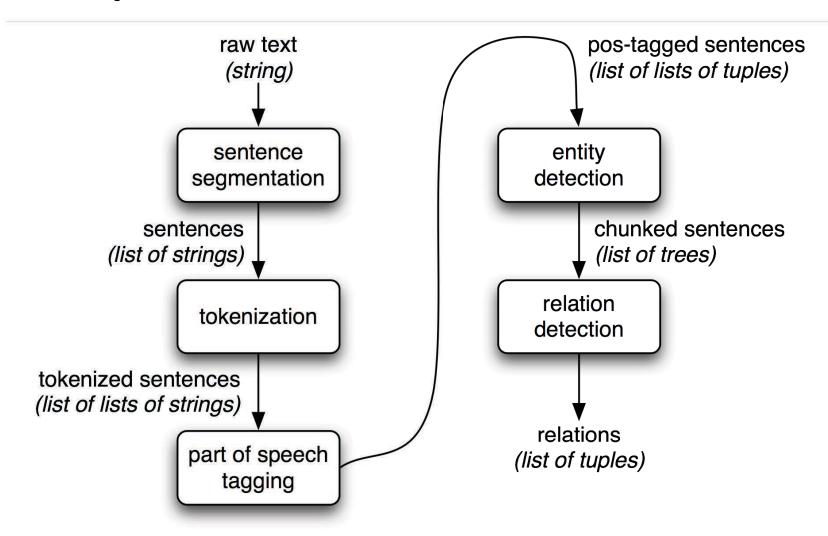
Hi Ricardo, we've now scheduled the curriculum meeting. It will be in Gates 171 tomorrow from 10:00 – 11:30. Carlos.

Create new Calendar Entry

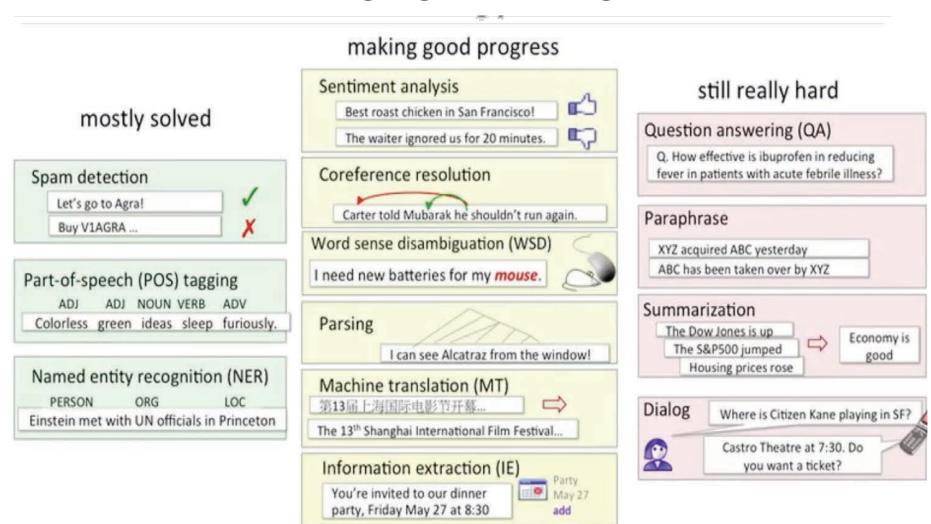
# **Examples - Pipeline**



# **Example – Relation Detection**

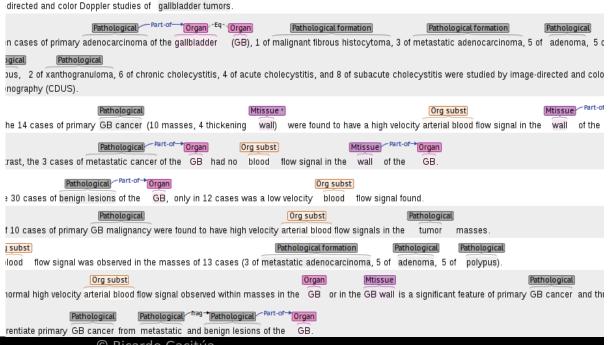


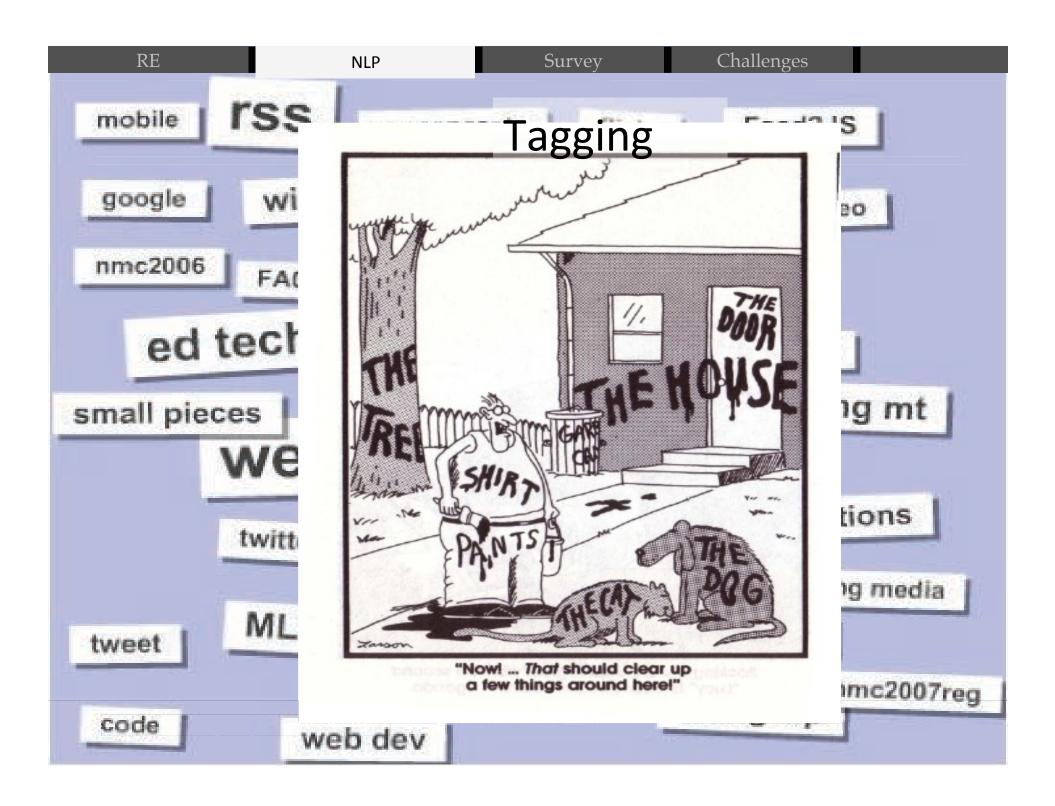
# State of the art in language technologies





- A tag is a non-hierarchical keyword or term assigned to a piece of information (such as an Internet bookmark, digital image, or computer file).
- This kind of metadata helps describe an item and allows it to be found again by browsing or searching.





# Statistical NLP

 Statistical natural-language processing uses stochastic, probabilistic and statistical methods to resolve some of the difficulties discussed above, especially those which arise because longer sentences are highly ambiguous when processed with realistic grammars, yielding thousands or millions of possible analyses

# **NLP Summary**

- Many challenges in NLP involve natural language understanding, that is, enabling computers to derive meaning from human or natural language input, and others involve natural language generation.
- NLP research is gradually shifting from lexical semantics to compositional semantics and, further on, narrative understanding.
  - Human-level natural language processing, however, is an Al-complete problem. That is, it is equivalent to solving the central artificial intelligence problem—making computers as intelligent as people, or strong Al.
  - NLP's future is therefore tied closely to the development of AI in general.

### **Tutorial:**

Procesamiento de Lenguaje Natural en Ingeniería de Requisitos: Contribuciones Potenciales y Desafíos de Investigación

CIBSE 2015 – Lima, Peru

Ricardo Gacitúa

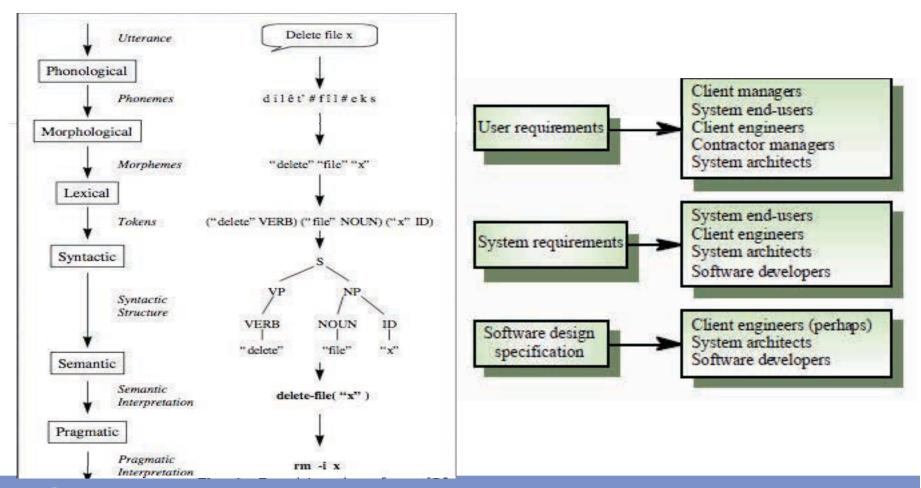
Centro de Estudios en Ingeniería de Software Depto. Cs. Computación e Informática Universidad de La Frontera Temuco, Chile

# **Indice**

- Ingeniería de Requerimientos (RE)
- Procesamiento de Lenguaje
   Natural (NLP)
- Avances en la relación entre RE
   & NLP
- Contribuciones Potenciales
- Desafíos de Investigación

# Después de este tutorial, podrás :

- Definir RE y su describir su importancia.
- Definir NLP y sus áreas de acción.
- Conocer el estado del arte en la relación entre NLP y RE.
- Describir las contribuciones potenciales de NLP para RE.
- Listar los principales desafíos para la investigación en RE.

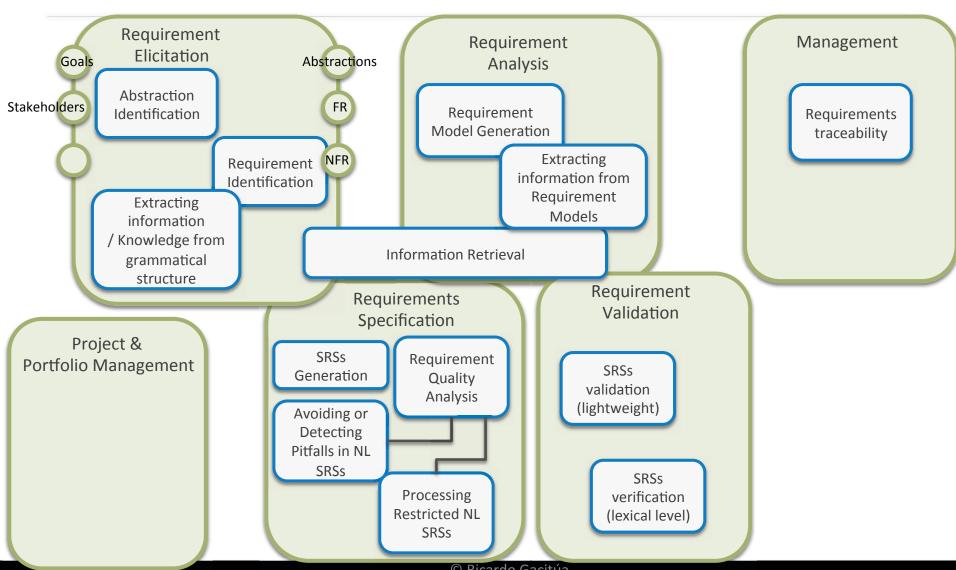


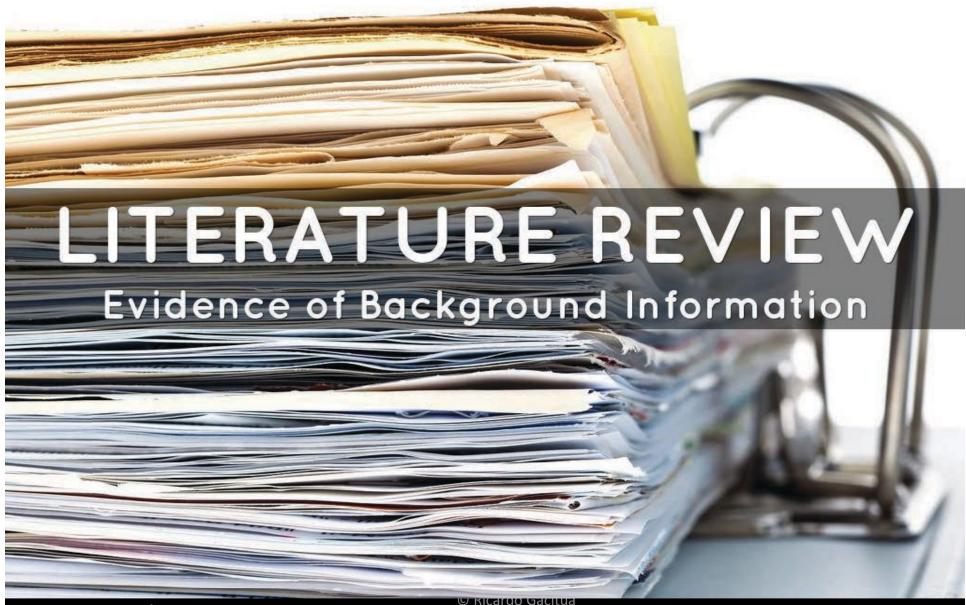
**RE & NLP** 

# **RE** main challenges

Sommerville, I.	Bashar Nuseibeh and Steve Easterbrook (2000)
<ol> <li>Changing requirements</li> <li>Lack of standardization</li> <li>Differing perspectives</li> <li>People and politics</li> </ol>	<ol> <li>Development of new techniques for formally modelling and analysing properties of the environment</li> <li>Bridging the gap between requirements elicitation approaches based on contextual enquiry and more formal specification and analysis techniques.</li> <li>Richer models for capturing and analysing nonfunctional requirements</li> <li>Better understanding of the impact of software architectural choices on the prioritisation and evolution of requirements.</li> <li>Reuse of requirements models.</li> <li>Multidisciplinary training for requirements practitioners.</li> </ol>
Ian Sommerville. 2015. Software Engineering: 10th Edition)-International Computer Science. Addison-Wesley Longman Publishing Co., Inc., Boston, MA, USA.	Bashar Nuseibeh and Steve Easterbrook. 2000. Requirements engineering: a roadmap. In Proceedings of the Conference on The Future of Software Engineering (ICSE '00). ACM, New York, NY, USA, 35-46

#### **NLP in RE – Current Potential Contributions**





#### NLP in RE -> Abstraction Identification

Requirement Quality Analysis

SRSs Generation Avoiding or Detecting Pitfalls in NL SRSs

Processing Restricted NL SRSs SRSs validation (lightweight) SRSs verification (lexical level)

Requirements traceability

Abstraction Identification

Requirement Identification

Extracting information / Knowledge from grammatical structure

Requirement Model Generation

Extracting information from Requirement Models

- R. Gacitua, P. Sawyer, V. Gervasi, Relevance-based abstraction identification: technique and evaluation, Requirements Engineering 16 (2011) 251–265.
- J.J.G. Flores, Linguistic processing of natural language requirements: the contextual exploration approach, in: Proceedings of the 10th International Workshop on Requirements Engineering: Foundation for Software, Quality (REFSQ'04), 2004, pp. 99–112
- P. Sawyer, P. Rayson, R. Garside, REVERE: support for requirements synthesis from documents, Information Systems Frontiers 4 (2002) 343–353.

## **NLP in RE -> Requirement Identification**

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- R. Ankori, Automatic requirements elicitation in agile processes, in: Proceedings of the IEEE International Conference on Software - Science, Technology and, Engineering, 2005, pp. 101–109.
- A. Casamayor, D. Godoy, M. Campo, Identification of non-functional requirements in textual specifications: a semi-supervised learning approach, Information and Software Technology 52 (2010) 436–445.
- C. Castro-Herrera, C. Duan, J. Cleland-Huang, B. Mobasher, A recommender system for requirements elicitation in large-scale software projects, in: Proceedings of the 2009 ACM Symposium on Applied Computing SAC '09, 2009, pp. 1419–1426.
- Casamayor, A., Godoy, D., & Campo, M. (2010). Identification of non-functional requirements in textual specifications: A semi-supervised learning approach. *Information and Software Technology*, *52*(4), 436-445.

### **NLP in RE -> Extracting information**

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- Ashish Sharma and Dharmender Singh Kushwaha. 2011. Natural language based component extraction from requirement engineering document and its complexity analysis. SIGSOFT Softw. Eng. Notes 36, 1 (January 2011).
- Leonid Kof, Requirements Analysis: Concept Extraction and Translation of Textual Specifications to Executable Models, Application of Natural Language to Information Systems, LNCS, 2010.
- Martin Ringsquandl1 and Mathias Schraps, Taxonomy Extraction From Automotive Natural Language Requirements Using Unsupervised Learning, International Journal on Natural Language Computing (IJNLC) Vol. 3, No.4, August 2014

## NLP in RE -> Requirement Model Generation

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- V. Ambriola, On the systematic analysis of natural language requirements with CIRCE, Automated Software Engineering 13 (2006) 107–167.
- H.M. Harmain, R.J. Gaizauskas, CM-builder: a natural language-based CASE tool for object-oriented analysis, Automated Software Engineering 10 (2003) 157–181.
- Deeptimahanti, D. K., & Sanyal, R. (2011, February). Semi-automatic generation of UML models from natural language requirements. In *Proceedings of the 4th India Software Engineering Conference* (pp. 165-174). ACM

#### NLP in RE ->Extracting information from Req. Models

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- Mathias Landhäuβer, Sven J. Körner, and Walter F. Tichy. 2014. From requirements to UML models and back: how automatic processing of text can support requirements engineering. Software Quality Control 22, 1 (March 2014)
- Hasegawa, R., Kitamura, M., Kaiya, H., & Saeki, M. (2009, January). Extracting conceptual graphs from Japanese documents for software requirements modeling. In *Proceedings of the Sixth Asia-Pacific Conference on Conceptual Modeling-Volume 96* (pp. 87-96). Australian Computer Society, Inc.
- Overmyer, S.P., Lavoie, B., Rambow, O. (2001) Conceptual Modeling through Linguistic Analysis using LIDA. Proceedings of the 23rd International Conference on Software Engineering. IEEE, Los Alamitos, pp. 401-410.

#### **NLP in RE -> Information Retrieval**

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- Andrea De Lucia, Fausto Fasano, Rocco Oliveto, and Genoveffa Tortora. 2007. Recovering traceability links in software artifact management systems using information retrieval methods. ACM Trans. Softw. Eng. Methodol. 16, 4, Article 13 (September 2007).
- Sannier, N., Baudry, B.: Toward multilevel textual requirements traceability using modeldriven engineering and information retrieval. In: MoDRE 2012 (2012).
- Rayson, P., Garside, R., and Sawyer, P. (2000). Assisting requirements engineering with semantic document analysis. In Proceedings of RIAO 2000 (Recherche d'Informations Assistie par Ordinateur, Computer-Assisted Information Retrieval) International Conference, Collège de France, Paris, France, April 12--14, 2000. C.I.D., Paris, pp. 1363--1371.

## **NLP in RE -> Requirements Quality Analisis**

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- Fantechi, A., Gnesi, S., Lami, G., & Maccari, A. (2003). Applications of linguistic techniques for use case analysis. Requirements Engineering, 8(3), 161-170.
- Fabbrini, F., Fusani, M., Gnesi, S., & Lami, G. (2001). The linguistic approach to the natural language requirements quality: benefit of the use of an automatic tool. In Software Engineering Workshop, 2001. Proceedings. 26th Annual NASA Goddard (pp. 97-105). IEEE..
- Lami, G., Gnesi, S., Fabbrini, F., Fusani, M., & Trentanni, G. (2004). An automatic tool for the analysis of natural language requirements. Informe técnico, CNR Information Science and Technology Institute, Pisa, Italia, Setiembre.

#### **NLP in RE -> SRSs Generation**

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- Ghosh, S., Elenius, D., Li, W., Lincoln, P., Shankar, N., & Steiner, W. (2014). Automatically Extracting Requirements Specifications from Natural Language. *arXiv preprint arXiv:* 1403.3142.
- Ilieva, M. G., & Ormandjieva, O. (2005). Automatic transition of natural language software requirements specification into formal presentation. In Natural Language Processing and Information Systems (pp. 392-397). Springer Berlin Heidelberg..
- Osborne, M., & MacNish, C. K. (1996, April). Processing natural language software requirement specifications. In Requirements Engineering, 1996., Proceedings of the Second International Conference on (pp. 229-236). IEEE..
- Bajwa, I. S., Lee, M. G., & Bordbar, B. (2011, March). SBVR Business Rules Generation from Natural Language Specification. In AAAI Spring Symposium: AI for Business Agility (pp. 2-8).

#### **NLP in RE -> Avoiding or Detecting Pitfalls in SRSs**

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- Berry, D. M. (2008). Ambiguity in natural language requirements documents. In Innovations for Requirement Analysis. From Stakeholders' Needs to Formal Designs (pp. 1-7). Springer Berlin Heidelberg.
- Erik Kamsties, Daniel M. Berry, Barbara Paech, Detecting Ambiguities in Requirements Documents Using Inspections, in Proceedings of the First Workshop on Inspection in Software Engineering (WISE'01), 2001.
- Hui Yang, Anne N. De Roeck, Vincenzo Gervasi, Alistair Willis, Bashar Nuseibeh: Speculative requirements: Automatic detection of uncertainty in natural language requirements. RE 2012: 11-20.
- Hui Yang, Anne N. De Roeck, Vincenzo Gervasi, Alistair Willis, Bashar Nuseibeh: Analysing anaphoric ambiguity in natural language requirements. Requir. Eng. 16(3): 163-189 (2011)

## **NLP in RE -> Processing Restricted NL SRSs**

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- Cobleigh, R. L., Avrunin, G. S., & Clarke, L. A. (2006, November). User guidance for creating precise and accessible property specifications. In *Proceedings of the 14th ACM SIGSOFT international symposium on Foundations of software engineering* (pp. 208-218). ACM.
- Tjong, S. F., Hartley, M., & Berry, D. M. (2007, May). Extended Disambiguation Rules for Requirements Specifications. In *WER* (pp. 97-106).
- Giganto, R. (2008). Generating class models through controlled requirements. In NZCSRSC-08, New Zealand Computer Science Research Student Conference Christchurch, New Zealand.

#### **NLP in RE -> SRSs Validation**

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- Gervasi, V., & Nuseibeh, B. (2002). Lightweight validation of natural language requirements. Software: Practice and Experience, 32(2), 113-133.
- Falessi, D., Cantone, G., & Canfora, G. (2010, September). A comprehensive characterization of NLP techniques for identifying equivalent requirements. In *Proceedings* of the 2010 ACM-IEEE international symposium on empirical software engineering and measurement (p. 18). ACM.
- Kaiya, H., & Saeki, M. (2005, September). Ontology based requirements analysis: lightweight semantic processing approach. In *Quality Software*, 2005.(QSIC 2005). Fifth International Conference on (pp. 223-230). IEEE.

#### **NLP in RE -> SRSs Verification**

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- Durán, A., Ruiz-Cortés, A., Corchuelo, R., & Toro, M. (2002). Supporting requirements verification using XSLT. In *Requirements Engineering*, 2002. Proceedings. IEEE Joint International Conference on (pp. 165-172). IEEE.
- Fanmuy, G., Fraga, A., & Llorens, J. (2012). Requirements verification in the industry. In *Complex Systems Design & Management* (pp. 145-160). Springer Berlin Heidelberg.
- MacDonell, S. G., Min, K., & Connor, A. M. (2014). Autonomous requirements specification processing using natural language processing. *arXiv preprint arXiv:1407.6099*.

## **NLP** in RE -> Requirements traceability

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- Sawyer, P., Gacitua, R. and Stone, A. Profiling and Tracing Stakeholder Needs. In INNOVATIONS FOR REQUIREMENT ANALYSIS. FROM STAKEHOLDERS' NEEDS TO FORMAL DESIGNS, pages 196-213, Springer-Verlag, LNCS, 2008
- Deng, M., Stirewalt, R.E.K., Cheng, B.H.C.: Retrieval by construction: a traceability technique to support verification and validation of UML formalizations. Int. J. Softw. Eng. Knowl. Eng. 15(5), 837-872 (2005).
- Zisman, A., Spanoudakis, G., Pérez-Miñana, E., & Krause, P. (2003, June). Tracing Software Requirements Artifacts. In *Software Engineering Research and Practice* (pp. 448-455).
- McMillan, C., Poshyvanyk, D., & Revelle, M. (2009, May). Combining textual and structural analysis of software artifacts for traceability link recovery. In *Traceability in Emerging Forms of Software Engineering*, 2009. TEFSE'09. ICSE Workshop on (pp. 41-48). IEEE.

#### **RE & NLP Considerations**

- TEXTUAL FORM: Requirements normally does not exist in textual form.
  - Some information occurs naturally as text, typically process descriptions or predefined procedures, but much more is to be found in diagrams or in the physical reality surrounding the client.
  - Assuming however that the requirements definition task is being performed by an
    intelligent human and that a substantial body of machine readable text is available,
    there is no doubt that tools to scan, search, browse and tag that text could assist in
    developing a full and accurate statement of needs.

#### COMPLEXITY:

- Understanding in NLP seems to be a fractal-like problem.
- Narrow domain understanding of natural language may be achieved in the medium term.

#### COGNITIVE AND SOCIAL:

 It is not possible to depend on NLP to solve the RE bottle-neck, since systems are, and will increasingly be recognised to be, social organisms, embodying everything from the deterministic microchip to the emotional and personal needs of the people involved.



Tutorial

## **RESEARCH CHALLENGES**

### **RE Summary**

- **1. SOCIAL SCIENCES:** RE draws on the cognitive and social sciences:
  - Cognitive Psychology, Anthropology, Sociology, Linguistics
- **2. BOUNDARIES:** The identification of stakeholders and user classes, of goals and tasks, and of scenarios and use cases all depend on how the boundaries are chosen.
- **3. MODELS:** Models can be used to represent a whole range of products of the RE process.
- **4. COMMUNICATION:** RE is not only a process of discovering and specifying requirements, it is also a process of facilitating effective communication of these requirements among different stakeholders.
- **5. DOCUMENTATION:** The way in which requirements are documented plays an important role in ensuring that they can be read, analysed, (re-)written, and validated.
- **6. VALIDATION:** Requirements validation is difficult for two reasons.
  - The first reason is philosophical in nature, and concerns the question of truth and what is knowable.
  - The second reason is social, and concerns the difficulty of reaching agreement among different stakeholders with conflicting goals.
- 7. CHANGES: Managing changing requirements is not only a process of managing documentation, it is also a process of recognising change through continued requirements elicitation, re- evaluation of risk, and evaluation of systems in their operational

#### **RE Challenges**

#### Sommerville, I.

- Changing requirements
- 2. Lack of standardization
- Differing perspectives
- 4. People and politics

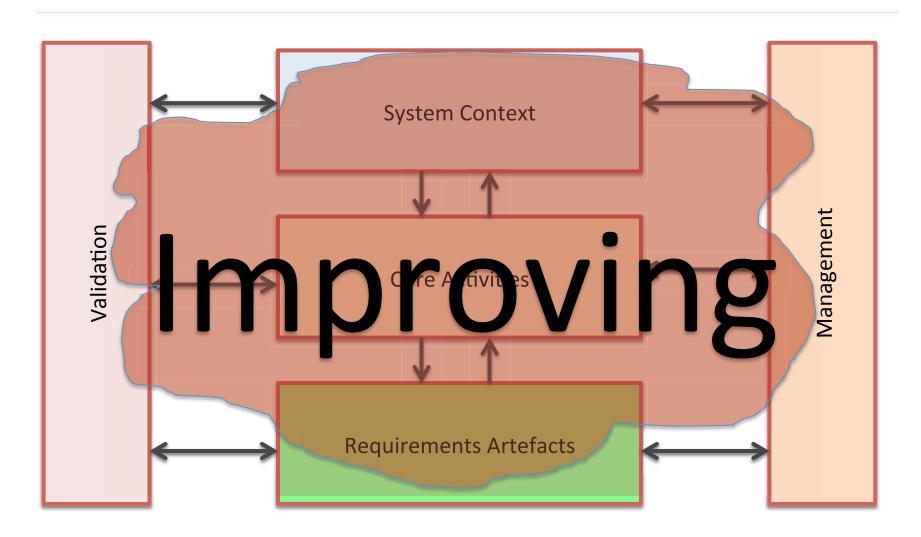
#### **Bashar Nuseibeh and Steve Easterbrook (2000)**

- Development of new techniques for formally modelling and analysing properties of the environment
- 2. Bridging the gap between requirements elicitation approaches based on contextual enquiry and more formal specification and analysis techniques.
- 3. Richer models for capturing and analysing nonfunctional requirements
- 4. Better understanding of the impact of software architectural choices on the prioritisation and evolution of requirements.
- Reuse of requirements models.
- Multidisciplinary training for requirements practitioners.

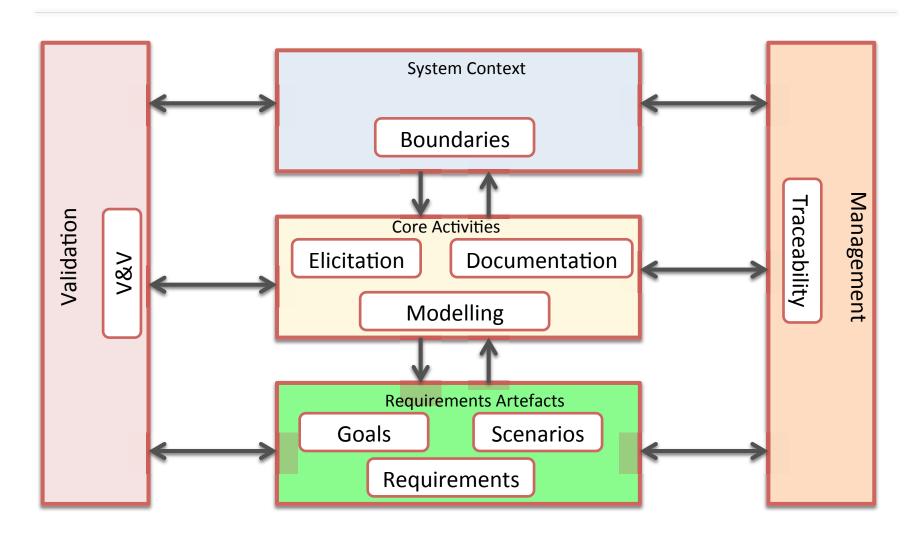
Ian Sommerville. 2015. Software Engineering: 10th Edition)-International Computer Science. Addison-Wesley Longman Publishing Co., Inc., Boston, MA, USA.

Bashar Nuseibeh and Steve Easterbrook. 2000. Requirements engineering: a roadmap. In Proceedings of the Conference on The Future of Software Engineering (ICSE '00). ACM, New York, NY, USA, 35-46

## **Research Challenges - Framework**



## Some Research Challenges for NLP in RE



## **RE & NLP -> Research Challenges**

Elicitation		Analyse	Specify	Validate	Ma	anage	
Extract	Identify	Generate				Trace	Version

# System Context

Core Activities

#### Requirements Artefacts

1. From speech to text: Automatic interview transcription

- 2. Automatic extraction and selection of pertinent domain documents.
  - Big Data, Semantic Web
- 3. Automatic extraction of scenarios
- 4. Automatic adequacy to standards and guidelines
- 5. Automatic identification of domain artefacts
  - Domain terminology and Semantic models (e.g. Ontologies)

### **RE & NLP -> Research Challenges**

Elicitation		Analyse	Specify	Validate	Ma	anage	
Extract	Identify	Generate				Trace	Version

# System Context

Core Activities

Requirements Artefacts

- 1. Automatic identification of requirements artefacts
  - FR, NFR, Goals, Stakeholders, Business rules, constraints etc.
- 2. Automatic synthesize and validation of domain models
- 3. Automatic validation of Requirements
  - Consistency, coherence, duplicity, etc.
- 4. Automatic improvement of requirements
  - Detection of ambiguity,
- 5. Automatic validation of SRSs
- 6. Automatic building SRSs from domain models and NL text

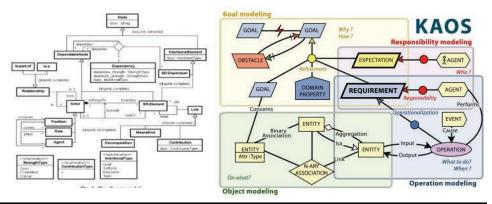
Elicitation		Analyse	Specify	Validate	Ma	nage	
Extract	Identify	Generate				Trace	Version

# System Context

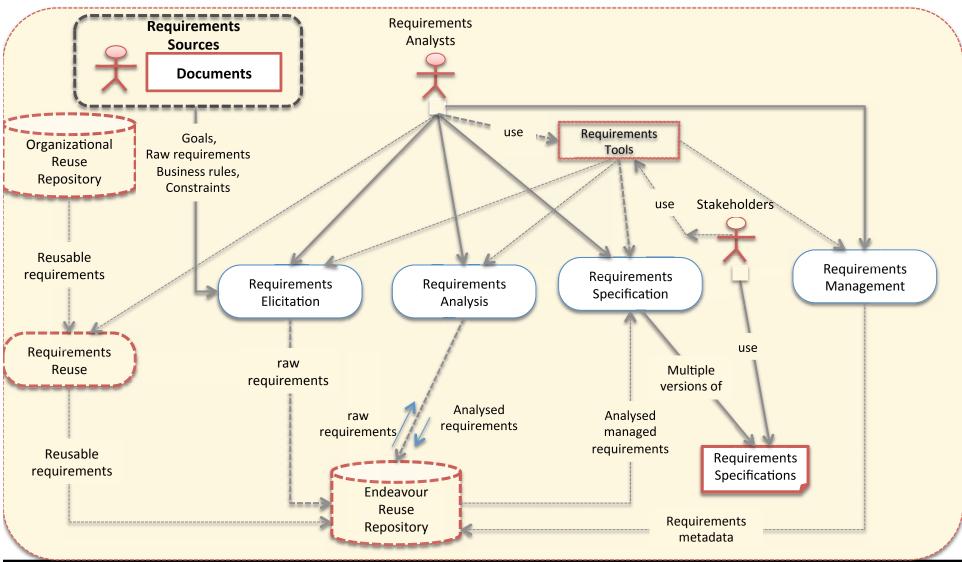
**Core Activities** 

Requirements
Artefacts

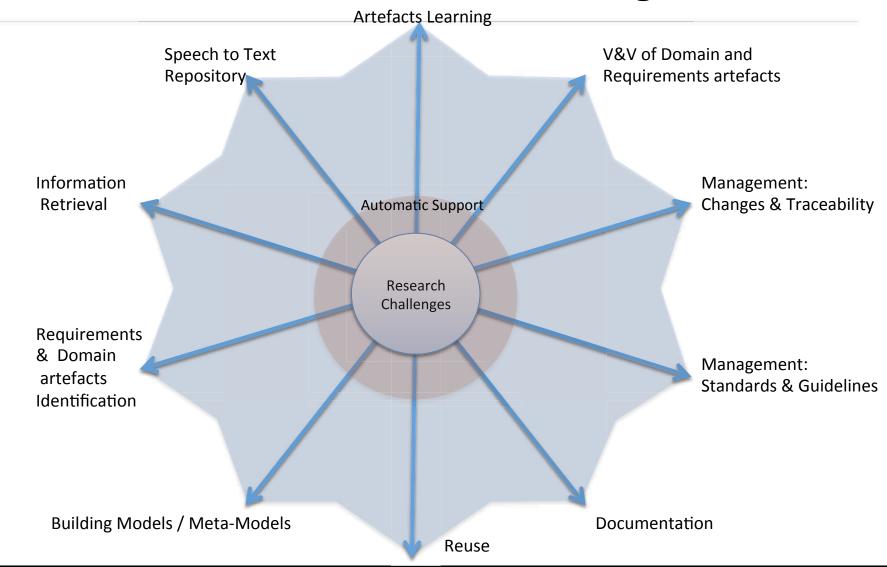
- 1. Reuse of domain and requirements artefacts
- 2. Scenarios Learning
- 3. Automatic building models of requirements artefacts from Meta-Models
  - Example: i\* Metamodel and Kaos Metamodel



## **RE & NLP -> Research Challenges -> Scope**



## **RE & NLP -> Some Research Challenges**



#### **Conclusion**

- Requirements engineering is a non—trivial task and NLP is not able to solve all the requirements engineering problems.
  - However, it tackles an extremely important step, namely providing preliminary requirements and domain artefacts (e.g. preliminary models, preliminary lists of requirements and document summaries) for the stakeholders and requirements engineers.
- It is in supporting the social process, and not in supplanting it, that natural language processing will have its proper role.



RE & NLP|

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