Identifying and Classifying Ambiguity for Regulatory Requirements

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Legal Domain: Healthcare

- Health Insurance Portability and Accountability Act (HIPAA) passed in 1996
 - Regulates security and privacy for healthcare organizations
 - \$25,000 fines per violation per year for non-criminal violations
 - Amended by the HITECH Act in 2009 to address data breaches and increase enforcement actions
- Recent Settlement Actions:
 - Concentra Health Services \$1.7 Million (April 2014)
 - New York and Presbyterian Hosptial **\$3.3 Million** (May 2014)
 - Columbia University Hospital **\$1.5 Million** (May 2014)

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Legal Ambiguity: a Critical Challenge for Requirements

- Legal texts are often intentionally ambiguous.
 - Example:

"make *reasonable* efforts to limit protected health information to the minimum necessary to accomplish the intended purpose of the use" – HIPAA §164.502(b)

- The word "reasonable" appears 61 times in HIPAA!

- Traditional approaches, such as disambiguation or removal, do not work for legal ambiguities.
 - Legal texts cannot easily be re-written
 - Legal stakeholders cannot easily be sought out for definitive clarification.
 - Requirements engineers must interpret ambiguities in legal texts!

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What is ambiguity?

- ANSI/IEEE Standard 830-1993: a requirements specification is unambiguous only when each requirement has a single interpretation.
- Definitional Concerns:
 - Should a statement with no clear interpretation be considered ambiguous?
 - What constitutes a valid interpretation? Who decides?
- No objective standard exists.
 - There is no "correct" identification or classification of ambiguity.
 - We do have relative standards: Does a group agree as a whole on an interpretation?

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Research Overview

- Case study of 18 students identifying and classifying ambiguity
 - technologists
 - policy analysts
- Using a taxonomy based on linguistics, software engineering, and legal understandings of ambiguity.
- Legal text: §170.302 of the HITECH Act.
 - 23 paragraphs (104 lines)
 - Meaningful Use Stage 1 Criteria for a certified EHR
- Tutorial introducing the taxonomy and study procedure
- 5 Research Questions

A Taxonomy of Ambiguity

Lexical	Syntactic	Semantic	Vagueness
Incompleteness	Referential	Other	Unambiguous

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Lexical Ambiguity

- Lexical ambiguity occurs when a word or phrase has multiple valid meanings.
- Examples:
 - Conversational: Melissa walked to the bank.
 - §170.302(d): Enable a user to electronically record, modify, and retrieve a patient's active medication list as well as medication history for longitudinal care.





Syntactic Ambiguity

- Syntactic ambiguity occurs when a sequence of words has multiple valid grammatical parsings.
- Examples:
 - Conversational: I saw the man with the binoculars.
 - §170.302(f): Enable a user to electronically record, modify, and retrieve a patient's vital signs...





Semantic Ambiguity

- Semantic ambiguity occurs when a sentence has more than one interpretation based entirely on the surrounding context.
- Examples:
 - Conversational: Fred and Ethel are married.
 - §170.302(j): Enable a user to electronically compare two or more medication lists.





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Vagueness

- Vagueness occurs when a term or statement admits borderline cases or relative interpretation.
- Examples:
 - Conversational: George is tall.
 - §170.302(h)(3): Electronically attribute, associate, or link a laboratory test result to a laboratory order or patient record.





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Incompleteness

- Incompleteness occurs when a statement fails to provide enough information to have a single clear interpretation.
- Examples:
 - Conversational: Combine flour, eggs, and salt to make fresh pasta.
 - §170.302(a)(2): Provide certain users with the ability to adjust notifications provided for d





drug-allergy interaction

Referential Ambiguity

- Referential ambiguity occurs when a word or phrase in a sentence cannot be said to have a clear reference.
- Examples:
 - Conversational: The boy told his father about the damage.
 He was very upset.
 - §170.302(n): For each meaningful use objective with a percentage-based measure, electronically record the numerator and denominator...



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Per-paragraph Response Block

Please complete the table below using the procedure detailed in the tutorial.

Ambiguities Found in 45 CFR Subtitle A, § 170.302(a)(1)-(2)

Lexical	Syntactic	Semantic	Vagueness
Incompleteness	Referential	Other	None

Software engineers should be able to build software that complies with this legal text.

Circle one:

Agree

Disagree

Research Questions 1 to 3

- 1. Does the taxonomy provide **adequate coverage** of the ambiguities found in § 170.302?
- 2. **Do participants agree** on the number and types of ambiguities they identify in § 170.302?
- 3. Do participants agree on the number and types of **intentional** ambiguities they identify in § 170.302?

Research Questions 4 and 5

- 4. **Do participants agree** on whether software engineers should be able to build software that complies with each paragraph of § 170.302?
- 5. Does an **identified ambiguity affect** whether participants believe that software engineers should be able to build software that complies with each paragraph of § 170.302?

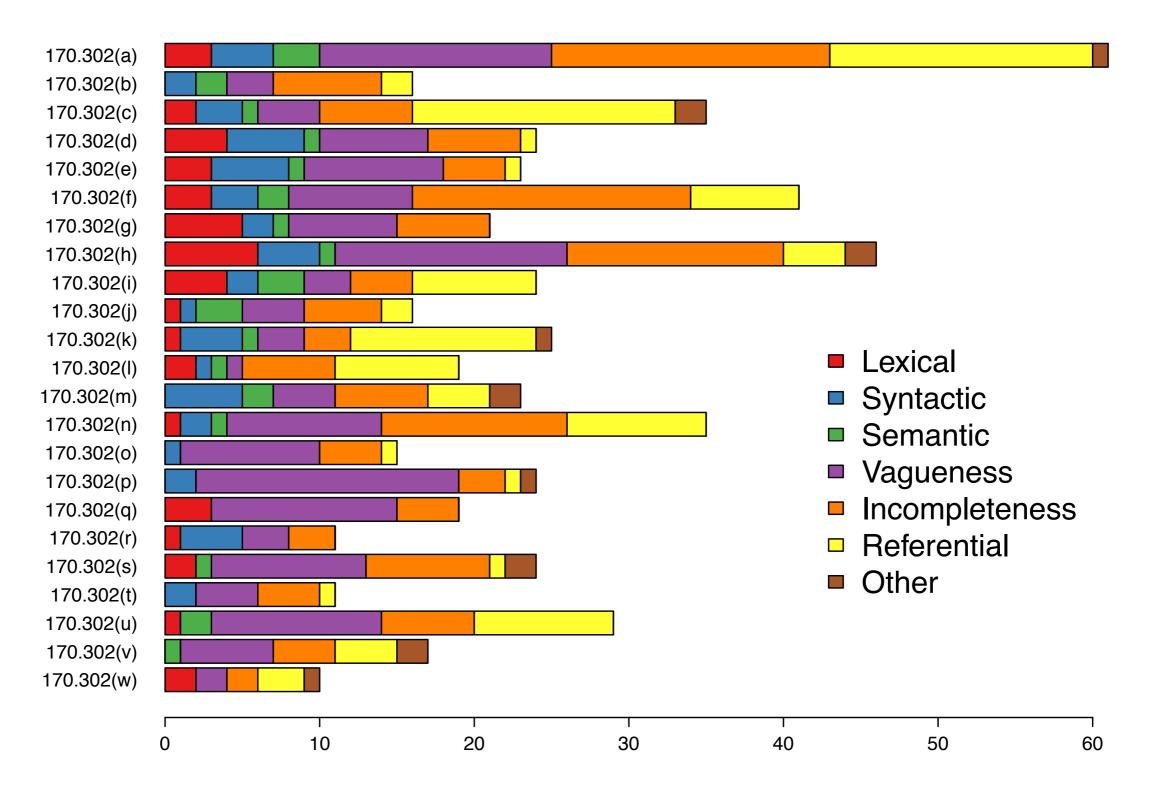
Research Question Measures

- Q1 Measures: (1) Use of each of the first six ambiguity types and (2) minimal use of the "Other" type.
- Q2 Measures: ICC for both number and type of ambiguities identified
- Q3 Measures: ICC for both number and type of intentional ambiguities identified
- Q4 Measures: Fleiss Kappa agreement on implementability of the paragraph.
- Q5 Measures: The percentage of paragraphs deemed unimplementable that contain identified ambiguities

Q1: Taxonomy Coverage

- Participants identified on average 33.47 ambiguities for the 23 paragraphs examined.
 - 50 minutes provided for the study
 - All participants finished before time was up
- Every ambiguity type was used.
 - Least frequent: Semantic (1.59 on average)
 - Most frequent: Vagueness (9.82 on average)
- The "Other" type was less common than the least common ambiguity classification we defined (0.82 on average).
- Result: Yes, the taxonomy provides adequate coverage.

Ambiguities per Paragraph



Q2: Number and Type agreement

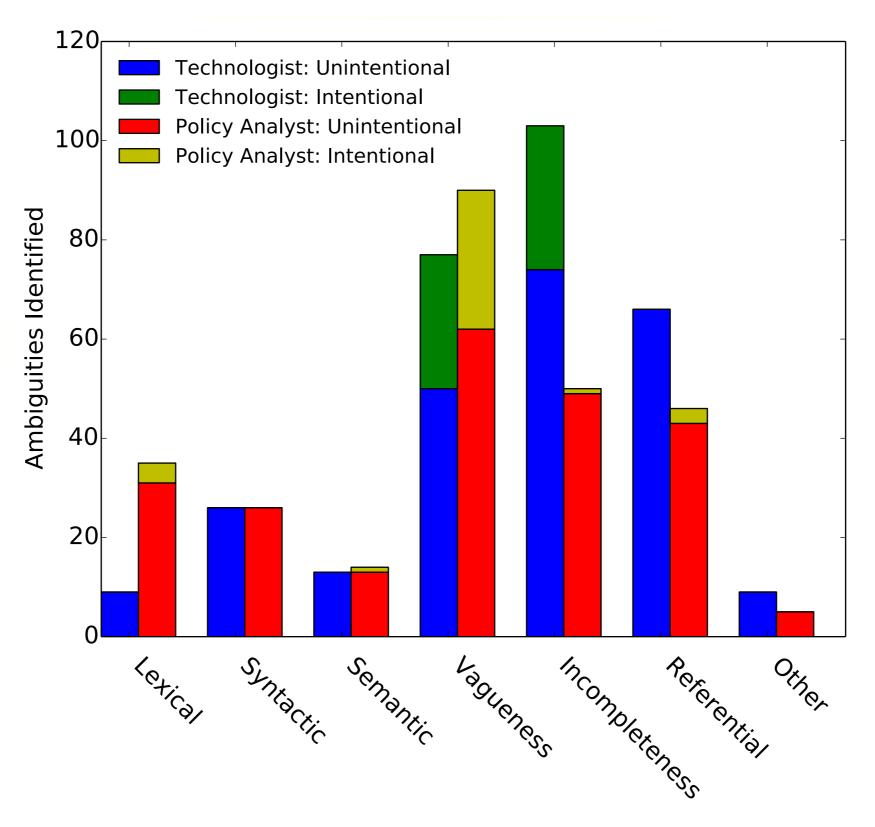
- Number agreement: ICC: 0.316, indicating fair agreement on number (p < 0.001)
- Type agreement:
 - For 2 of the 23 paragraphs, the participants demonstrated near-universal agreement.
 - For the remaining 21, the participants demonstrated only slight agreement.
 - Overall Fleiss Kappa for type agreement: 0.0446, indicating slight agreement on type (p < 0.0029)

Q3: Intentional Number and Type

- Number: ICC 0.141, (p < 0.0001)
- **Type:** ICC 0.201 (p < 0.001)
- The Incompleteness category was a primary driver of type disagreement.
 - Technologists identified significantly more ambiguities of this type.
 - Removing Incompleteness, Type agreement ICC becomes 0.39, indicating fair agreement (p < 0.0001)
- Result: Participants agreed less on intentional ambiguities than on total ambiguities.

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Ambiguities by Type and Intent



Q4: Implementability

- All participants: Fleiss Kappa value of 0.0052, p < 0.788 not statistically significant.
- Technologists: Fleiss Kappa value of 0.0455, p < 0.116 not statistically significant.
- Result: Participants agreement on implementability was not statistically significant.

Q5: Ambiguity and Implementability

- 89% of unimplementable paragraphs contained an ambiguity
- 48% of implementable paragraphs contained an ambiguity
- Result: Yes, ambiguity is more commonly identified in paragraphs deemed unimplementable.

Summary

- In 50 minutes over 104 lines of legal text our participants identified 33.47 ambiguities on average
- The taxonomy provided reasonable coverage: 97.5% of all ambiguities identified were classified as one of the six defined types
- Participants accepted paragraphs with unintentional ambiguity as implementable!

Future Work

- Participants did not exhibit strong agreement on the number and type of ambiguity.
 - 50 minute limit?
 - Better guidelines for the taxonomy?
- Additional case studies
 - More partipants
 - Different legal domains
 - Does identifying and classifying ambiguity prior to other legal requirements activities improve performance?

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Thank You! Questions?

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