

INT D 461: Artificial Intelligence Everywhere Capstone

Information Interview

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Problem & Motivation

- Our project explores how **large language models (LLMs)** handle **figurative language** — idioms, proverbs, metaphors across different languages.
- Idioms/proverbs/metaphors are non-compositional (surface words ≠ meaning).
- LLMs and MT often choose the literal sense, producing confident but wrong outputs.
- LLMs often **hallucinate** or “agree” too easily — like a *yes-man*.
- They sometimes **explain nonsense confidently**, showing a lack of real understanding.
- Downstream impact: mistranslation, misunderstanding, cultural bias, and user trust issues.
- We want to test **how well they actually reason** about figurative meaning, not just output correct-looking translations.

Potential Candidates

We brainstormed researchers in figurative language + NLP

Dr. Greg Kondrak

Bradley Hauer

Ning Shi

Dr. Greg Kondrak

Professor, Natural Language Processing & Computational Linguistics

Field: Sub-word NLP and Lexical Semantics

Focus Areas:

- Letter-phoneme conversion, transliteration, morphology, and word similarity
- Cognate identification and applications in translation and diachronic linguistics
- Lexical semantics: synonymy, polysemy, and word sense disambiguation

In essence: He develops algorithms that capture relationships between words and their forms across languages — foundational for translation, linguistic analysis, and decoding unknown scripts.

Ning Shi

Ph.D. Candidate, NLP & Semantics, supervised by Dr. Kondrak

Field: Computational Lexical Semantics (CLS)

Focus Areas:

- Multilingual semantic modeling and lexical disambiguation
- Applying algorithmic approaches to improve cross-lingual understanding
- Bridging theoretical semantics with practical language applications

In essence: He builds systems that understand word meaning across contexts and languages — crucial for tasks like translation and question answering.

Bradley Hauer

Postdoctoral Researcher, Natural Language Processing (NLP)

Field: Computational Lexical Semantics (CLS)

Focus Areas:

- Word sense disambiguation and multilingual semantic relations
- Building and applying semantic knowledge bases (e.g., WordNet, BabelNet)
- Using translation and multilingual data for semantic understanding

In essence: He studies how computers can distinguish and represent word meanings across contexts and languages — advancing translation, semantic search, and question answering.

Our Goal

- Goal: understand **where LLMs still struggle and why they struggle** along with to understand what would make our project **meaningful in 2025**.

Questions we brainstormed:

- Where do LLMs still fall short in idiom or figurative language understanding?
- If we built a lightweight system around LLMs, what would add real value?
- What would make a project on idioms stand out today? (e.g., evaluation, multilingual focus, interpretability)

Key Takeaways from Bradley

- LLMs **still hallucinate** idiom meaning and can't always **justify** why.
- **No true explanation generator** — they can “explain” wrong answers too.
- **Sense retrieval** or external reference systems could add value.
- **Multilingual focus** is promising — current models still **fall short for non-European languages**.
- Average-case performance is good, but **hard cases reveal real weaknesses**.

Other interesting topics in the Interview

- Retrieval + dictionary definitions → improves understanding.
- BubbleNet-style models still **struggle with non-European idioms**.
- Word sense disambiguation remains **important**.
- Multilingual idiom examples (e.g., Chinese) show **missing metaphors**.
- Small, open-weight models → good for **interpretability** and **testing logic**.

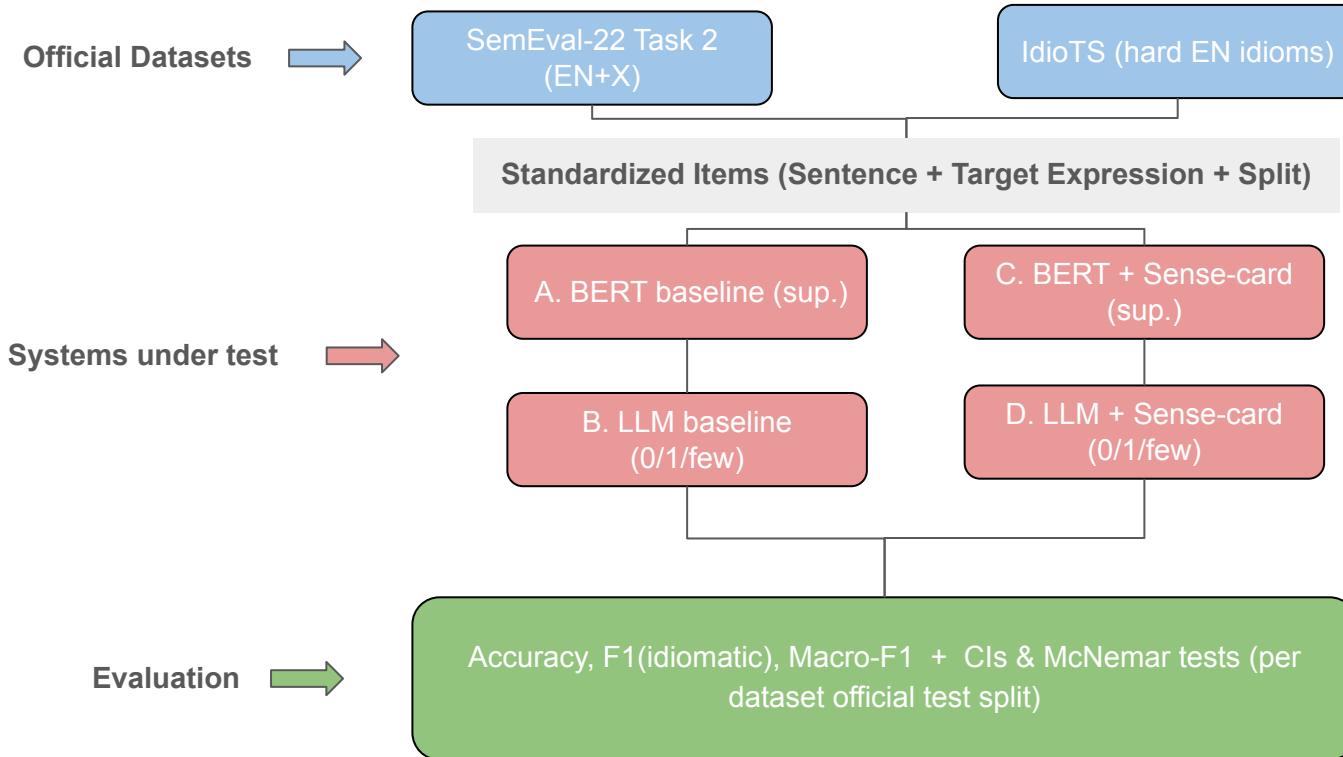
What We Learned

- Strong **evaluation design** is key — not just accuracy.
- Need to test **reasoning and explanations** behind outputs.
- Multilingual and **word-sense-aware evaluation** could stand out as novel research.
- “Open-weight” models let us inspect the model’s logic and are better for experimentation.

How we are using this information in our project

- Add **prompt engineering** (sense card) to disambiguate the sentence.
- Test **multilingual idiom understanding**.
- Explore **open-weight models** for transparency.

A flow chart for better understanding



**A big Thanks to Bradley Hauer for taking
the time out of his day to talk with us**

Feedback / Questions?