

# INT D 461: Artificial Intelligence Everywhere Capstone

## Literature Review

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# IdiomSense: Sense-Augmented LLMs for Idioms, Proverbs, and Metaphors

Inference-time “sense cards” to boost detection & translation

We test whether LLMs truly understand figurative language and show that adding a tiny retrieved sense card (from WordNet/BabelNet/IdiomKB) before inference improves idiom detection and idiomatic translation—without fine-tuning.

*Sentence → Sense Lookup → Sense Card → LLM → Output*

# Problem & Motivation

## Why figurative language breaks models:

- Idioms/proverbs/metaphors are non-compositional (surface words ≠ meaning).
- LLMs and MT often choose the literal sense, producing confident but wrong outputs.
- Downstream impact: mistranslation, misunderstanding, cultural bias, and user trust issues.

## Concrete example (minimal pair):

- **Idiomatic:** “After years of service, he kicked the bucket.” → means died.
- **Literal:** “The child kicked the bucket in the yard.” → means struck a container.

## Our goal:

- Build a small, hard evaluation and a simple inference-time sense hint (“sense card”) that helps models pick the right meaning and translate accordingly.

# BERT Idiom Detection (MWE'24)

- Fine-tuned BERT to decide if a highlighted span is **idiomatic or literal**, augmenting training with two signals:
- **word cohesion** (idioms “stick together”) and **translation drift** (idioms behave oddly under MT).
- Consistently outperforms older baselines on multiple English datasets → small, targeted signals lift performance.
- Our project **keeps the same idea** of adding *light, simple signals*, but instead of training a model with them (like BERT did), you apply them **at inference time** (when the model is already trained).
- We will do this using **sense cards** - short pieces of extra information (like a gloss or definition) that you prepend to the model’s input prompt.

“BERT-Based Idiom Detection” — ACL MWE Workshop 2024. <https://aclanthology.org/2024.mwe-1.26.pdf>

## Theme - Data Collection

**IdioTS: Hard Idiom Minimal Pairs for LLMs**

**CoAM: Corpus of all type multiword expressions**

- An **expert-curated test suite** built from **minimal pairs**, nearly identical sentences where one usage is idiomatic and the other literal.
- Even modern instruction-tuned LLMs **stumble** on these contrasts → fluency ≠ correct sense selection.
- We **adopt this evaluation style** (EN + one L2) and then test if a brief sense hint flips errors without any fine-tuning.

“IdioTS: A Hard Idiom Test Suite for LLMs.” arXiv preprint. <https://arxiv.org/pdf/2409.01053>

“CoAM: Corpus of all type multiword expressions” [\[2412.18151\] CoAM: Corpus of All-Type Multiword Expressions](#)

# Theme - Idiomatic translation & knowledge-base support for idioms

## IdiomKB: Retrieval Helps Idiomatic Translation

- **Multilingual idiom knowledge base** (glosses, examples) used at inference to **remind** models of figurative meanings before translating.
- Yields **fewer literal mistranslations**, especially for smaller models, when a short description of the intended sense is included.
- We generalize this into a compact sense card (gloss + micro-example) from **WordNet/BabelNet/IdiomKB** and test it not only for **translation** but also **detection** (and optional paraphrase).

  

- “IdiomKB: A Knowledge Base for Idioms.” arXiv preprint (v2). <https://arxiv.org/html/2308.13961v2>
- [Improving LLM Abilities in Idiomatic Translation - ACL Anthology](#)

# Theme - Multilingual figurative language understanding (idioms, similes) in LLMs

## Idiom↔Idiom Alignment Approaches

- Rather than explaining a sense, these methods **find an equivalent idiom** in the target language (via embeddings/LLMs) to preserve **style and tone**.
- Works well when a good target idiom **exists**, but offers less help when it **doesn't**, and doesn't solve **idiom vs literal** detection.
- Our approach is **complementary**: a **sense card** still guides the model when no clean target idiom exists, and it directly supports **the first decision**, is the usage idiomatic?

“Idiomatic Translation via Idiom-to-Idiom Alignment.” arXiv preprint. <https://arxiv.org/html/2405.10579v1>  
[Comparative Study of Multilingual Idioms and Similes in Large Language Models - ACL Anthology](#)

# Method & Evaluation

## Baselines:

- BERT idiom detector (EN only).
- LLM zero-shot prompts (detect / translate).

## Our variant:

- LLM + sense card.

## Metrics:

- Detection: accuracy/F1 on held-out pairs.
- Translation (if included): count literal errors; optionally COMETKiwi.
- Ablations: with/without card; small k (1-3); KB source.

# Plan & Deliverables

## Phases (sequence only)

- **Setup:** choose L2; finalize minimal-pair template; add WordNet lookup.
- **Baselines:** run BERT (EN) and LLM zero-shot; save outputs.
- **Sense Card:** add retrieval → ≤30-token card; run with/without; log coverage.
- **Wrap-up:** tables + a few examples; note limits; finalize slides/report.

## Deliverables

- **Mini dataset** (EN + L2 minimal pairs)
- **Scripts** (baselines + sense-card inference)
- **Brief report & slides** (method + observations)

# Commonalities

All the related works highlight the same pain point: LLMs are fluent but **semantically shallow** with idioms.

They show that **adding meaning cues**, whether through fine-tuning, retrieval, or data augmentation helps models disambiguate idioms.

Our project continues that trend but explores a **smaller, more practical intervention**: inference-time sense hints that don't require retraining or new infrastructure.

# The gap we are trying to fill

Most past research either focuses on **building new datasets** or **changing the model itself** through fine-tuning.

But there hasn't been much work on testing whether a **simple meaning hint at inference time**, like our sense card can actually help models understand idioms better.

That's where *IdiomSense* sits, between detection and translation, combining both into a minimal, model-agnostic framework that could make LLMs more semantically aware using just retrieved sense cards.

# Bibliography

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