







Multimodal Knowledge Graph Extraction with/for Grounded World Models

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Multimodal multilayer meaning



photo by Evan Vucci/The Associated Press

A <u>man</u> in an <u>elegant suit</u> in the center, with a <u>raised arm</u>, <u>surrounded</u> by some people gathering around his body. Some <u>blood</u> on his <u>ear</u>, an <u>American flag</u> on the background.

What we may know perceptually (physical patterns)









Multimodal multilayer meaning

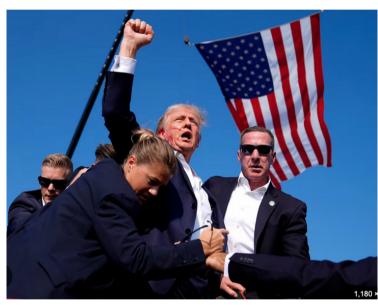


photo by Evan Vucci/The Associated Press

A <u>political figure or leader</u>, doing a <u>victory gesture</u>, surrounded by <u>guards / agents protecting him</u>. Some <u>blood spilling</u> from his ear, an <u>American flag</u> on the background.

What we may know commonsensically (social patterns)









Multimodal multilayer meaning

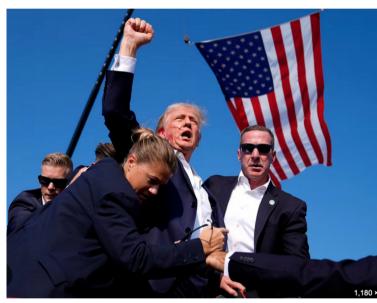


photo by Evan Vucci/The Associated Press

The assassination attempt on Donald Trump.

What we may know as a specific social fact (news-like)







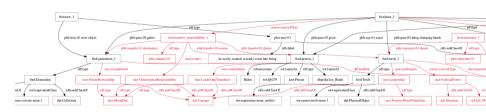


Research goals

- Content from multiple modalities, pragmatically described in natural language (NL), which acts as a hub from content to knowledge (automated extraction)
- *Knowledge** as a **supramodal hybrid representation** for joint logical and geometrical inference
- Ground World Models (passive, active) as shared reference knowledge
- Knowledge graphs representing both explicit (base KG) and implicit (eXtended KG - XKG) facts at different linguistic layers (literal, presupposed, implicated, figurative, etc.), and ontological layers (material, cognitive, social, abstract)



Extended Knowledge Graph



* Knowledge: logically represented data for automated reasoning









How to extract knowledge from arbitrary signal, and make sense of it [11]?

Direct to graph [1] Mediated by language (Multimodal generative AI) Passive vs. Active AI [2]

[1] Asprino L., Daga E., Gangemi A., & Mulholland P. Knowledge graph construction with a façade: a unified method to access heterogeneous data sources on the web. ACM Transactions on Internet Technology, 2023 (Q1)
[2] Pezzulo, G, Parr, T, Cisek, P, Clark, A, Friston, K. Generating meaning: active inference and the scope and limits of passive AI, Trends in Cognitive Sciences, 2024 (Q1)

[11] Ciroku, F, et al. Automated multimodal sensemaking: Ontology-based integration of linguistic frames and visual data. Computers in Human Behavior, 2024 (Q1)

AMR to FRED

Symbolic coercions Your text Metonymic coercions Moral value-driven coercions Biden said it was time to pass the torch to a new generation. Full graph - All extensions This is the result of converting the AMR graph to a FRED-like knowledge graph using AMR2FRED. Graphical Full graph This is full graph built by merging the base graph with all the extensions provided by the LLM. Zoom: double-click to zoom-in and Shift + Double-click to zoom-out, or use Shift + Scroll ireu:new_r a purs:new-wr ; Pan: drag left mouse button pblr:new-01.new-object fred:generation_1 . fred:pass_1 a pbrs:pass-05 ; coerce:coercedType fred:transfer_responsibility_1, fred:say_1 fred:new 1 fred:pass 1 mor:TransferOfPower ; pblr:pass-05.thing-changing-hands pblr:pass-05.giver vn.role:Time pblr:say-01.sayer vn.role:Time fred:time_1 ; pblr:say-01.utterance rdf:type pblr:pass-05.getter rdf:type pblr:new-01.new-object rdf:type pblr:pass-05.getter fred:generation_1 ; fred:torch_1 fred:person_1 fred:time_1 pbrs:pass-05 fred:generation_1 pbrs:say-01 pbrs:new-01 pblr:pass-05.giver fred:person 1 : pblr:pass-05.thing-changing-hands fred:responsibility_1, rdf:type rdf:type owl:sameAs rdf:type rdfs:label rdf:type rdfs:subClassOf rdfs:label rdfs:subClassOf rdfs:subClassOf rdfs:label fred:torch_1 . schema:name owl:sameAs fred:Generation fred:Torch amr:Person dbpedia:Joe_Biden Biden wd:Q6279 fred:Time give say dul:Event be newly created; recently come into being fred:represent 1 a pbrs:represent-01 ; pblr:represent-01.arg1 fred:responsibility_1 ; owl:equivalentClass rdfs:subClassO rdfs:subClassOf rdfs:subClassOf owl:equivalentClass rdfs:subClassOf rdfs:subClassOf owl:equivalentClass rdfs:subClassOf pblr:represent-01.theme fred:torch_1 . dul:PhysicalObject wn30:supersense-noun_artifact dul:Amount wn30:supersense-noun_time dul:Collection wn30:synset-coevals-noun-1 wn30:synset-torch-noun-1 wn30:synset-time-noun-4 wn30:supersense-noun_group fred:say_1 a pbrs:say-01 ; pblr:say=01.sayer fred:person_1 ; pblr:say-01.utterance fred:time_1 . mor:commitment a owl:ObjectProperty . mor:evokes a owl:ObjectProperty . fred:new_1 fred:say_ fred:pass 1 coerce:coercedType rdf:type coerce:coercedType vn.role:Time p rdf:type pblr:new-01.new-object pblr:pass-05.getter fred:transfer responsibility pbrs:new-01 pblr:pass-05.giver pblr:say-01.sayer pblr:pass-05.thing-changing-hands fred:represent_1 pbrs:pass-05 mor:TransferOfPower fred:time rdf:type rdfs:label rdfs;subClassOf pblr:transfer-01.destination rdf:type pblr:transfer-01.source rdfs:label rdfs:subClassOf rdfs:subClassOf rdfs:subClassOf owl:equivalentClass pblr:pass-05.thing-changingpblr:transfer-01.theme pblr:represent-01.theme fred:generation_1 pbrs:transfer-01 mor:evokes be newly created; recently come into being fred:person_1 fred:torch_1 pbrs:represent-01 pblr:represent-01.arg1 dul:Event wd:Q1150070 give rdf:type schema:name owl:sameAs / rdf:type owl:sameAs dul:hasRole rdf:type oerce:coercedType dul:represents owl:eq mor:commitment mor:commitment mor:symbolize mor:LeadershipTransition Biden wd:O6279 amr:Person dbpedia:Joe_Biden fred:Torch mor:PoliticalPower fred:responsibility_1 Generation mor:FutureStewardship mor:GenerationalResponsibility mor:Leadership wn:synset-time-noun-4 rdfs:subClassOf rdf:type rdfs:subClassOf rdfs:subClassOf rdfs:subClassOf equivalentClass rdf:type rdfs:subClassOf owl:equivalentClass rdfs:subClassOf rdf:type rdfs:subClassOf owl:equivalentClass owl:sameA rdf:type rdfs:subClassOf owl:sameAs rdf:type un-1 dul:Collection mor:MoralDue dul:Concept wn:supersense-noun_artifact wn:synset-torch-noun-1 dul:PhysicalObject mor:PositiveMoralValueGain dul:Situation wd:Q1128437 wn:synset-responsibility-noun-1 wd:Q10470 fred:Responsibility

Extended graphs

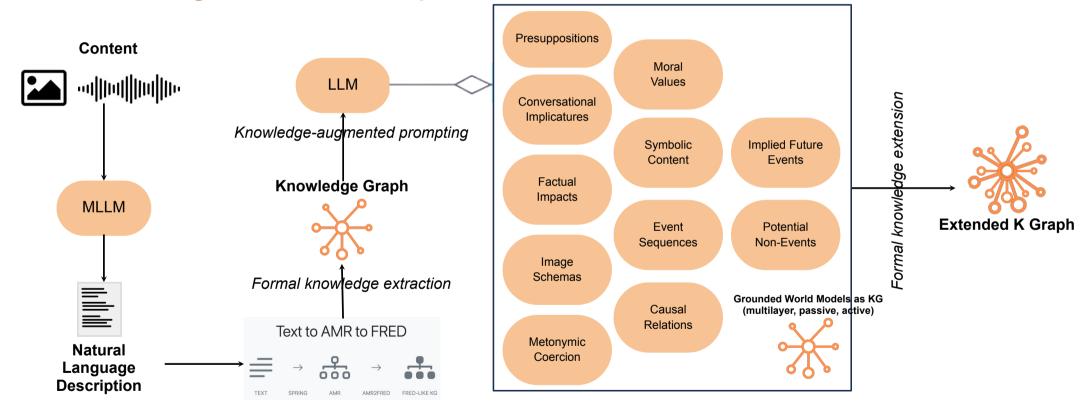








Knowledge Extension Pipeline 1/6



Off-the-shelf formal graph extensions from automated LLM prompting (11 implicit motifs)



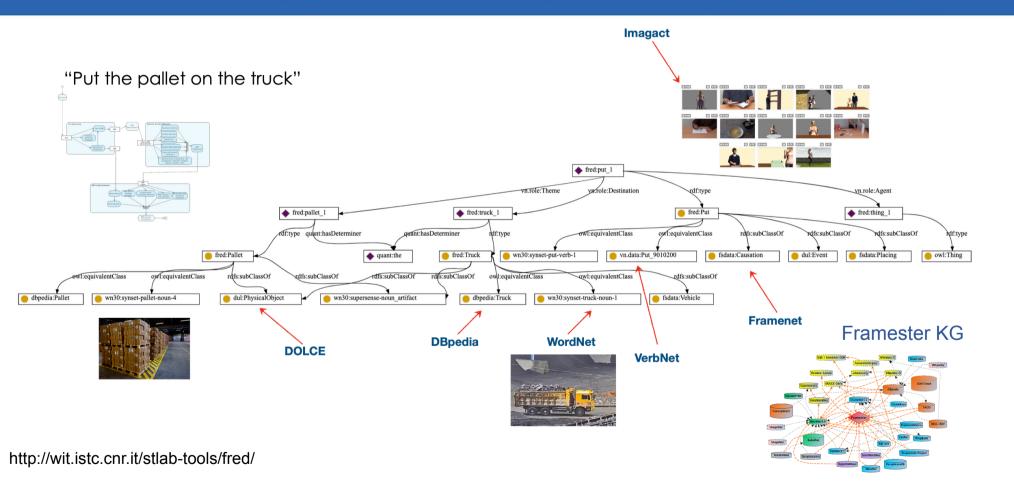






Alignment to GWM [3]

KG Extraction and



[3] Gangemi et al. Semantic Web Machine Reading, *Semantic Web Journal*, 2017 (Q1)
[4] Gangemi et al. Identifying motifs for evaluating open knowledge extraction on the Web. *Knowledge-Based Systems*, 2016 (Q1)

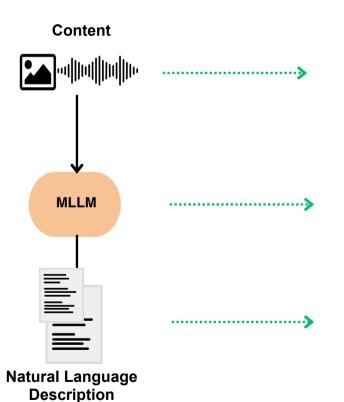


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Knowledge Extension Pipeline – 2/6







Clever Prompts:

- 1. Describe the overt, implicit and symbolic content of the image. Do not use full stops, Do not use "this image depict"-like expressions, describe directly the content of the image
- 2. Describe this photo in detail, including gestures, participants, public roles, situations, bodily position, and general implied meaning.
- 3. ...

A high-profile political figure raises his fist in defiance, surrounded by vigilant security personnel during a public rally, with a large American flag prominently displayed in the background, emphasizing themes of patriotism and protection.



Test it! [5]

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🖅 FROM TEXT







FROM IMAGE

Given some text as input, this tool will parse it into an <u>AMR</u> (Abstract Meaning Representation) graph, using <u>SPRING</u>. The AMR graph is then converted into an RDF/OWL knowledge graph that follows <u>FRED</u>'s knowledge representation patterns, using <u>AMR2FRED</u>. The KG can then be extended with implicit contextual knowledge with the support of a Large Language Model (LLM).

A high-profile political figure raises his fist in defiance, surrounded by vigilant security personnel during a public rally, with a large American flag prominently displayed in the background, <u>emphasizing</u> themes of patriotism and protection.

https://arco.istc.cnr.it/itaf/

Settings



[5] Gangemi et al., Text2AMR2FRED, a Tool for Transforming Text into RDF/OWL Knowledge Graphs via Abstract Meaning Representation, International Semantic Web Conference, 2023 (A)







🗉 FROM TEXT 🖪 FROM IMAGE

Image to Text to AMR to FRED

Given an image, this tool will get a textual description from an LLM and parse it into an AMR (Abstract Meaning Representation) graph, using <u>SPRING</u>. The AMR graph is then converted into an RDF/OWL knowledge graph that follows <u>FRED</u>'s knowledge representation patterns, using AMR2FRED.

Provide your image

Choose file trump_ap.jpg

AMP2ERED ERED-LIKE KG

Knowledge Extension Pipeline – 3/6

A high-profile political figure raises his fist in defiance, surrounded by vigilant security personnel during a public rally, with a large American flag prominently displayed in the background, emphasizing themes of patriotism and protection.



[6] Bevilacqua, M et al., One SPRING to rule them both: Symmetric AMR semantic parsing and generation without a complex pipeline, AAAI, 2021 (A*)









Knowledge Extension Pipeline – 4/6

graph excerpt due to its size pblr:emphasize-01.emphasizer fred:raise 1 pblr:raise-01.logical-subject vn.role:Time fred:raiseBy fred:fist 1 pblr:raise-01.agent fred:defy_1 fred:surround 1 fred:high_1 fred:vigilant_1 fred:ra rdf:type rdf:type amr:part-of pblr:defy-01.defiant-one rdf:type pblr:surround-01.thing-surrounding pblr:surround-01.thing-surrounded pblr:vigilant-01.watcher fred:figure_2 rdf:type pblr:high-02.thing-located-up-high pbrs:defy-01 fred:personnel_1 pbrs:surround-01 pbrs:high-02 pbrs:rally-01 rdf:type rdf:type rdf:type rdfs:label rdfs:label rdfs:subClassOf rdfs:subClassOf fred:PoliticsFigure fred:ProfileFigure to oppose, resist, or refuse fred:SecurityPersonnel be on all sides of, encompassing in the upper ranges (a rdfs:subClassOf rdfs:subClassOf alentClass dul:associatedWith rdfs:subClassOf dul:associatedWith dul:associatedWith rdfs:subClassOf fred:Politics dul:PhysicalObject fred:Security fred:Figure fred:Profile fred:Personnel rdfs:subClassOf owl:equivalentClass rdfs:subClassOf owl:equivalentClass rdfs:subClassOf rdfs:subClassOf rdfs:subClassOf wn30:supersense-noun_person wn30:synset-name-noun-4 dul:Person wn30:synset-security-noun-9 wn30:supersense-noun_act d0:Activity wn30:supersense-noun_group dul

Formal frame representation of AMR graphs using FRED's motifs, and alignment to (factually, linguistically) grounded knowledge graphs from the Framester hub (PropBank, FrameNet, WordNet, Wikidata, etc.)

FRED's OWL graph



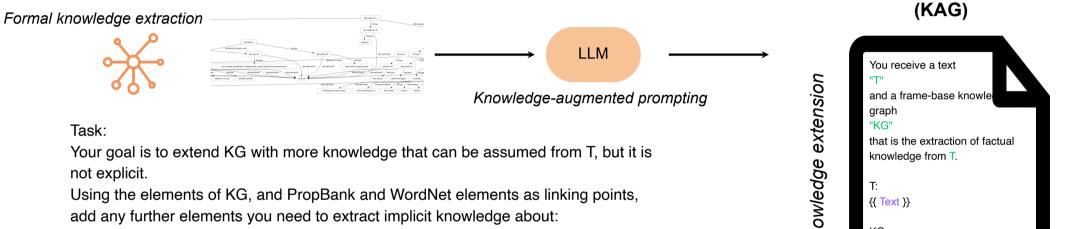




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Knowledge Extension Pipeline – 5/6



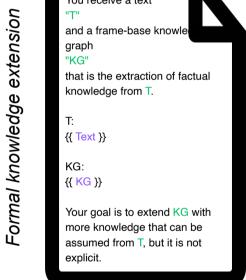
Conversational implicatures.

Conversational implicatures, in the sense of Grice's pragmatics.

Here are natural language inference examples:

1) She won't necessarily get the job -> She will possibly get the job

. . .



Knowledge-Augmented Generation

conversational implicatures.prompt



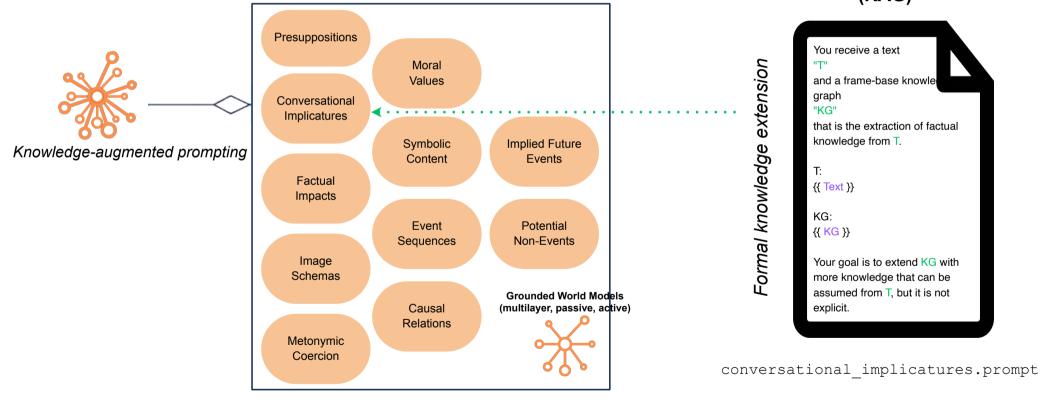






Knowledge Extension Pipeline – 6/6

Knowledge-Augmented Generation (KAG)



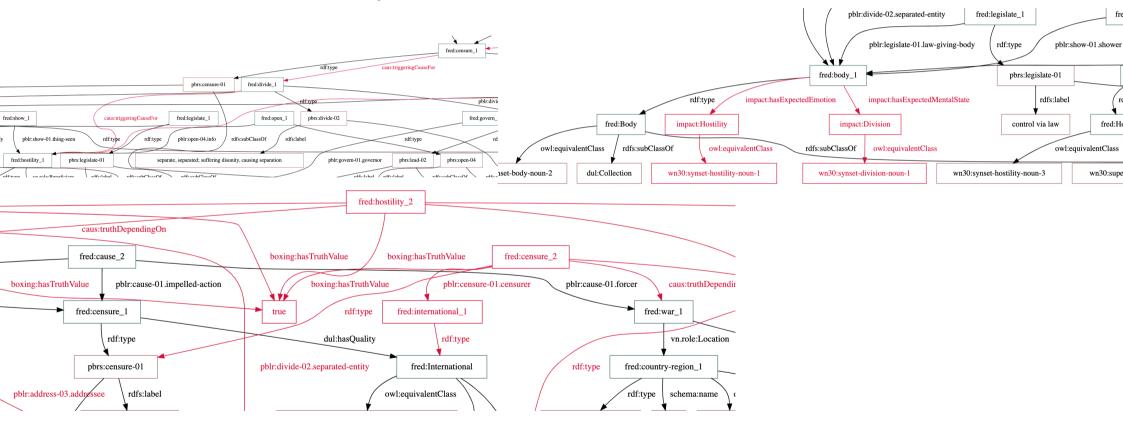








Visualisation of Graph Extensions









fred:fist 1 coerce:coercedType

fred:flag_1 coerce:coercedType

mor:Resistance, mor:Strength.

mor:NationalIdentity.mor:Patriotism.

fred:raise 1 a pbrs:raise-01;

pblr:raise-01.agent fred:figure 2; pblr:raise-01.logical-subject fred:fist 1.

fred:surround 1 a pbrs:surround-01; pblr:surround-01.thing-surrounded fred:personnel 1; pblr:surround-01.thing-surrounding fred:figure 2.

The figure has a raised fist. The personnel surrounds the figure.

FACTUAL IMPACT

MORAL

GRAPH

BASE

fred:figure 2 impact:hasExpectedEmotion impact:Defiance ; impact:hasExpectedSocialImpact impact:IncreasedPublicSupport .

fred:personnel 1 impact:hasExpectedEmotion impact:Tension ; impact:hasExpectedMentalState impact:Alertness.

The figure has expected social impact an increased public support. The personnel has expected mental state alertness.

fred:assert 1 a pbrs:assert-02; VALUES mor:evokes mor:PowerDemonstration ; pblr:assert-02.agent fred:figure_2; pblr:assert-02.topic fred:power_1.

The figure assertion is a demonstration of power.

fred:support_1 a owl:ObjectProperty, pbrs:support-01; pblr:support-01.supported fred:figure_2; pblr:support-01.supporter fred:audience 1.

We can expect a raise in the support from the audience.



The flag is a symbol of national identity and patriotism.

The fist is a symbol of Resistance and strength.

Formal knowledge extension

> **IMPL** FUTURE EVENTS Π









Evaluation methods and downstream tasks

- A new benchmark: experts create sample extensions, manually based on the heuristics, and we automatically estimate the distance between human and AI using a large KG embedding space and measuring the similarity between the basic KG and XKG
- Silver standard for precision@k of generated and inferred triples: experts¹ review the extended triples for correctness, relevance, and meaningfulness for each heuristics
- Stability and Reproducibility: replicating the XKG extensions for the different heuristics to establish how deterministic a LLM functionality is, and the difference between multiple LLMs



Extended K Graph

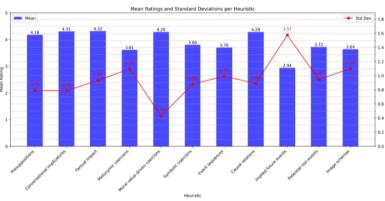


Figure 6: Mean Ratings and Standard Deviation per Heuristics.









Evaluation methods and downstream tasks

- Semantic similarity: using an independent LM, measure the similarity between the original text and a text generated back from XKG: high distance → improper extension; low distance → minimal extension
- Logical validity: OWL reasoners check the consistency/coherence of the extended KG
- Foundational ontology-grounded validity: deeper reasoning via full alignment of predicates to foundational ontologies like DOLCE Zero (as made for DBpedia in [7])
- **Downstream tasks**: deeper question answering (e.g., FigurativeQA), nuanced content recommendation, exploratory search, automated hate speech detection, fact checking, legal document analysis, medical diagnosis support, risk assessment, VR interaction, synchronic and diachronic analysis of domain knowledge [8], etc.



Extended K Graph

[7] Paulheim H., & Gangemi A., Serving DBpedia with DOLCE–more than just adding a cherry on top, ISWC 2015 (A)
[8] Gangemi A., Graciotti A., Meloni A., Nuzzolese A., Presutti V., Reforgiato D., MusicBO, an application of Text2AMR2FRED to the Musical Heritage domain, ESWC 2024 (B)



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Critical issues

- Shared GWM KGs like Wikidata are difficult for LLM-based alignment (opaque ID): *more traditional approaches* work (e.g., BLINK for the base KG of iTAF)
- Active GWMs need to be adapted to produce a KG output associated with shared and local KGs: *ongoing work* for spatial and causal GWM and Bayesian networks
- KAG requires a good amount of data, and prompts are heavy (and eventually costly under intense usage): moving to *fine-tuning open source models* is a natural move to reduce data at prompting time
- KAG with our heuristics runs well with Claude's Sonnet 3.5, but other large models (incl. GPT4o) do not: *moving to open source, mixture of experts, and small model stitching* in the medium term
- Eventually, *Agentic MLLMs* might get the full-monty, and KAG gets a built-in feature that helps making them robust, flexible, interpretable, and sustainable









Ongoing work: Scene Recognition, Hybrid Reasoning, Active GWMs

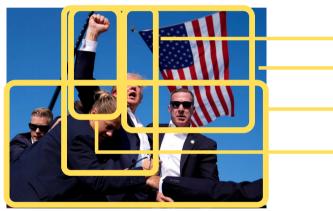


photo by Evan Vucci/The Associated Press

:AttackVictim , :PoliticalFigure :AttackContext , :AmericanFlag :symbolOf :Patriotism . :VictimDefender , :SecretServiceAgent :RaisedFist :symbolOf :Triumph . :RaisedFist :partOf :PoliticalFigure .

Cf. WP10.3 on EASG, e.g. within industrial cascade call

- Native hybrid, frame-based reasoning can be provided by novel neuro-symbolic reasoners such as Sandra [9]
- Open-ended knowledge extensions, e.g., norms, trust (WP8), affordances, etc.
- Further extensions for **dynamic in-context representation**, **anticipation and inference** can be assisted by dedicated active GWMs [2][10]

[9] Lazzari, N., De Giorgis, S., Gangemi, A., & Presutti, V. (2024). Sandra--A Neuro-Symbolic Reasoner Based On Descriptions And Situations. arXiv preprint arXiv:2402.00591.
Collaboration with SPOKE 8
[10] Petri, F., Asprino, L., & Gangemi, A. (2024). Transformers and Slot Encoding for Sample Efficient Physical World Modelling. arXiv preprint arXiv:2405.20180.
[2] Pezzulo, G, Parr, T, Cisek, P, Clark, A, Friston, K. Generating meaning: active inference and the scope and limits of passive Al, Trends in Cognitive Sciences, 2024.



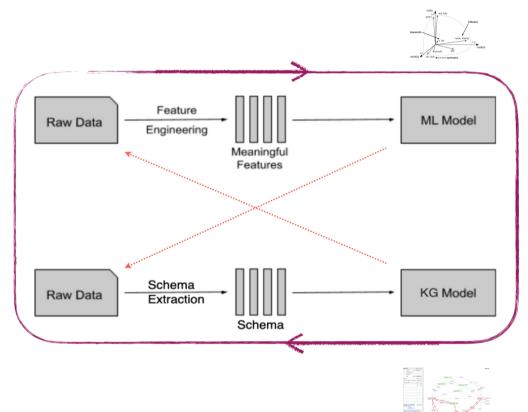
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Foundational issue: Hybridising GWMs

- Generative Als are signal processing machines: input from any modality is used to learn activation patterns, and at inference time they *output symbols that we may interpret as world-referring*
- Knowledge-based Als are logical machines: they extract/design symbolic representation patterns of the world, with a model-theoretical interpretation to ensure correct inference
- A **feedback loop** for multimodal, multilayered Grounded World Models (GWM)
- **GWM are multi-varied**: physical, neurocognitive, social, cultural, ...



Ongoing work: including cognitive, normative, causal, physical, affordance GWM









Foundational issue: a good KG for starting, then sensible patterns

- LLMs are not (yet) good at extracting the qualitative knowledge we need, but are **sensible to local, circumstantial patterns**
- Automated machine reading extracts the basic KG expressed by NL and grounds it on public knowledge
- KG are **firstly extracted neuro-symbolically**, then they are fed to a LLM jointly with adequate prompting for **in-context learning**
- Starting from a qualitative KG, ICL extends it (XKG) with contextual knowledge and GWMs, using heuristic pragmatic patterns (presuppositions, implicatures, impact, coercion, etc.)



Extended K Graph









Thanks for your attention