Grounding Multilingual Multimodal LLMs With Cultural Knowledge









Yueqi Song



Simran Khanuja

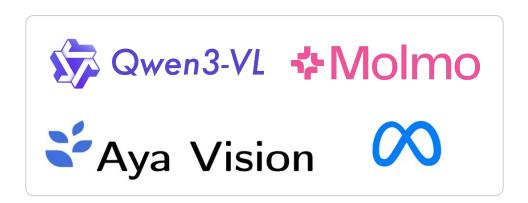


Graham Neubig





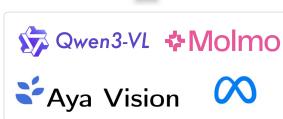
Al that serves the world should know more than Hollywood and Wikipedia – yet most open-source multimodal LLMs don't.



Issues in Multimodal LLMs

- Multimodal LLMs have deficits in understanding world cultures and long-tail entities.
- Even frontier models lags behind when it comes to cultures and open-source models are way behind.
- The cultural gap in MLLM:
 - Cultural nuances scarcity in existing large multimodal datasets(LAION, LLaVA-Ins, etc...)
 - Existing cultural datasets are small, limited cultural domains, lack of long-tail entities coverage.





CulturalGround

CulturalGround addresses major issues in existing cultural-centric datasets:

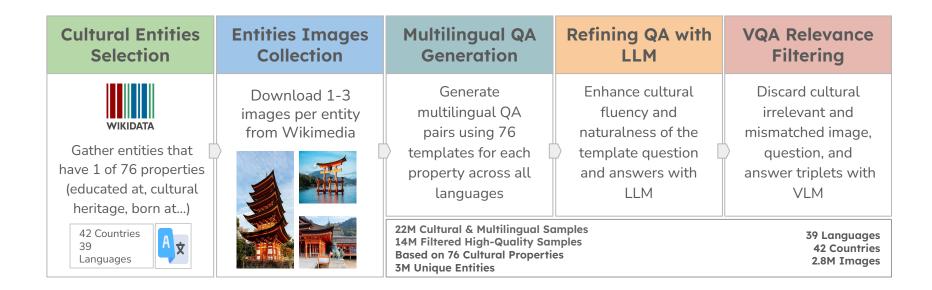
- Multilingual coverage: 39 languages
- Long-tail entities coverage
- Visual grounded: 1-3 images per entity
- A wide array of culture domains(people, places, schools, museums, heritages, etc..)
- **Size**: 30M, the largest open-source cultural VQA dataset.

Grounding Multilingual Multimodal LLMs With Cultural Knowledge

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https://neulab.github.io/CulturalGround/

CulturalGround Curation Pipeline



CulturalGround curation pipeline automatically create questions and answers about diverse set of entities from select regions.

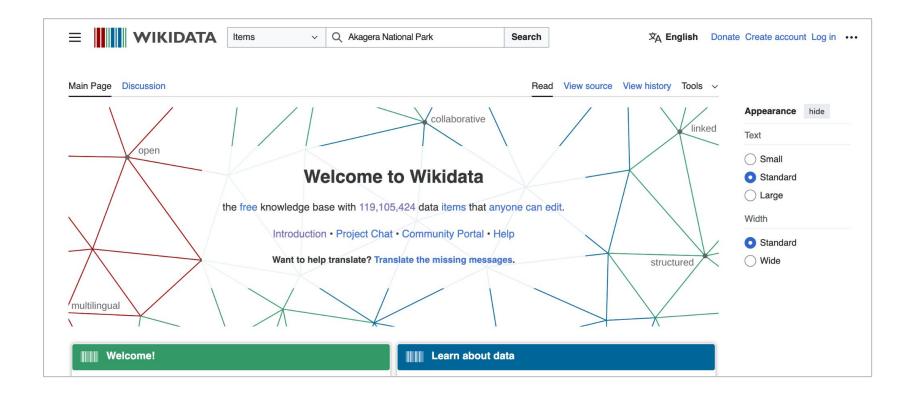
CulturalGround Curation MCQs



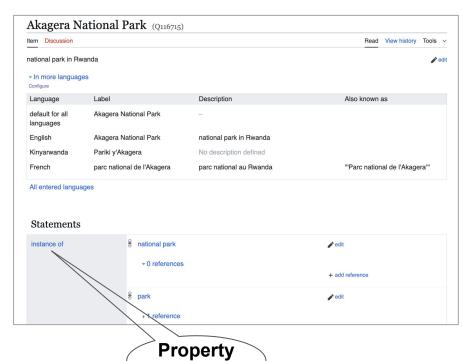
Using selected cultural entities and multilingual QAs as **seed**, we also generate multiple choice questions(MCQs) subset grounded in cultures using LLM.

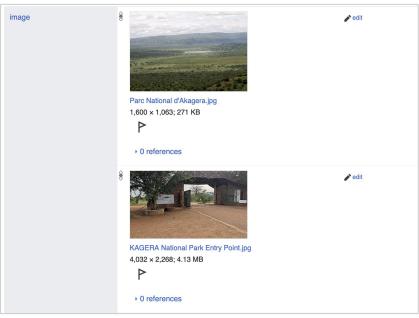
CulturalGround MCQs 8M Cultural & Multilingual Samples 6M Filtered High-Quality Samples Based on 76 Cultural Properties

A Walkthrough: Wikidata Homepage

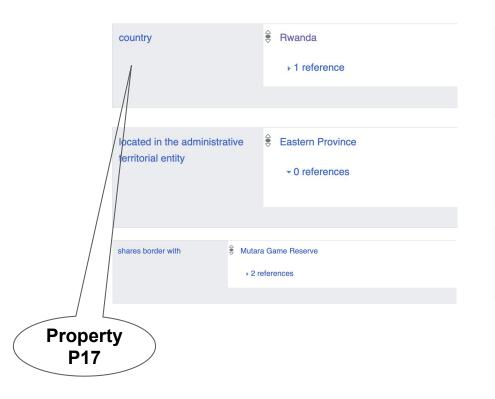


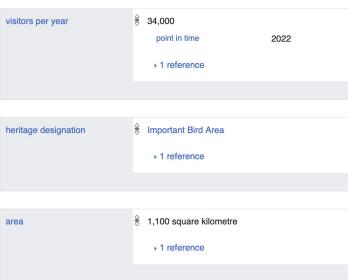
Entity in Wikidata: Akagera National Park





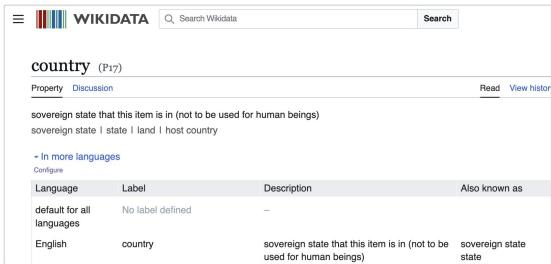
Entity in Wikidata: Akagera National Park







Properties of a given entity are also entries in Wikidata.



state of Rwanda."

Property P17

"relation": "P17",
"label": "country",
"description": "sovereign state of this
item; don't use on humans",
"template_question": "Which sovereign
state does this entity belong to?",
"answer_template": "{entity_label}
belongs to the sovereign state of
{value_label}."

Templates question and answer

"template_question": "Which sovereign state does this entity belong to?", "answer_template": "Akagera National belongs to the sovereign

Refined question and answer

"refined_question": "Where in the world would you find this national park?", "refined_answer": "Akagera National Park is located in Rwanda, a country in East Africa. It's known for its incredible biodiversity and efforts in wildlife conservation, particularly its successful reintroduction of lions and rhinos."

Property P17

```
"relation": "P17",
"label": "country",
"description": "sovereign state of this
item; don't use on humans",
"template_question": ""此实体属于哪个主权
国家?",
"answer_template": "{entity_label}属于主
```

权国家{value_label}。"

Templates question and answer

"template_question": "此实体属于哪个主权国家**?"**,

"answer_template": "阿卡蓋拉國家公園属于主权国家卢旺达, 非洲东部的国家。"

Refined question and answer

"refined_question": "这片国家公园位于哪个国家?",

"refined_answer": "阿卡蓋拉國家公園位于卢旺达, 这是一个位于非洲东部的国家。它以其多样化的野生动物和美丽的自然风光而闻名, 是卢旺达重要的生态保护区。"

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The same questions and answers formulation are done in 39 languages!!

Each QA with multiple images







Refined question and answer

"refined_question": "这片国家公园位于哪个国家?",

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CulturalPangea

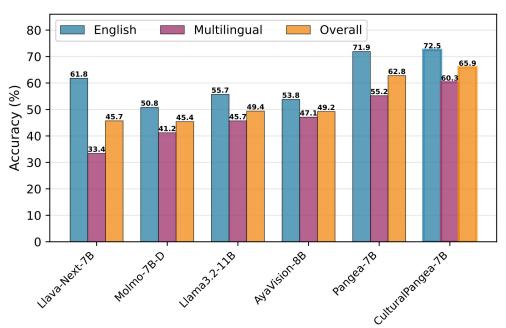
	Cultura	l Understa	inding	Entity Recognition	Multilin	igual VQA	Captioning	Average
Models	CVQA	MARVL	ALM	MERLIN	MAXM	M3EXAM	XM100	ALL
Llava-Next-7B	40.9	50.9	42.4	34.1	21.4	28.4	15.5	33.4
Molmo-7B-D	58.7	54.9	49.1	42.9	37.5	39.1	6.0	41.2
Llama3.2-11B	69.6	58.1	56.6	49.1	43.9	36.6	5.8	45.7
PaliGemma-3B	42.5	52.2	35.7	13.1	19.9	25.6	0.6	27.1
mBLIP-mT0-XL	37.5	66.7	36.9	15.8	36.8	25.0	6.8	32.2
AyaVision-8B	50.8	64.5	55.1	55.3	52.1	41.7	10.0	47.1
Pangea-7B	56.9	78.7	59.9	66.0	53.3	42.0	29.7	<u>55.3</u>
CulturalPangea	59.1	80.3	63.5	81.1	53.9	46.7	36.9	60.3
Δ over Pangea	+2.2	+1.6	+3.6	+15.1	+0.6	+4.7	+7.2	+5.0

By training Pangea on subset of CulturalGround while interleaving subset of Pangea training data, a resulting model **CulturalPangea** achieves SOTA on multiple cultural benchmarks while maintaining good performance on general multilingual benchmarks.

CulturalPangea

While most performance gains are observed on multilingual split, CulturalPangea surprisingly maintains good English performance on all benchmarks compared to Pangea.

Performance Comparison: English vs Multilingual

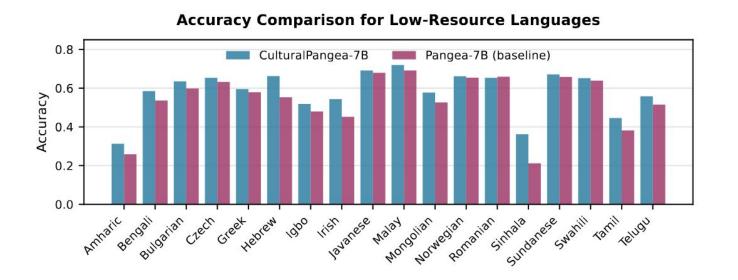


Does CulturalPangea Transfer to OOD Cultures and Languages?

41.7 45.0
42.1 48.3
+0.4 +3.3

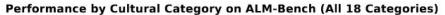
CulturalPangea perform better on languages that were neither in base model training data or CulturalGround. This strong generalization across cultures and languages is attributed to extensive data curation pipeline.

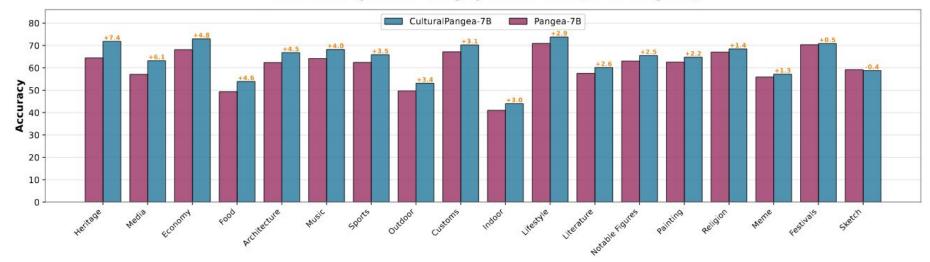
Does CulturalGround Helps Low-Resource Languages?



We also find that CulturalPangea does perform better on low-resourceful languages compared to Pangea

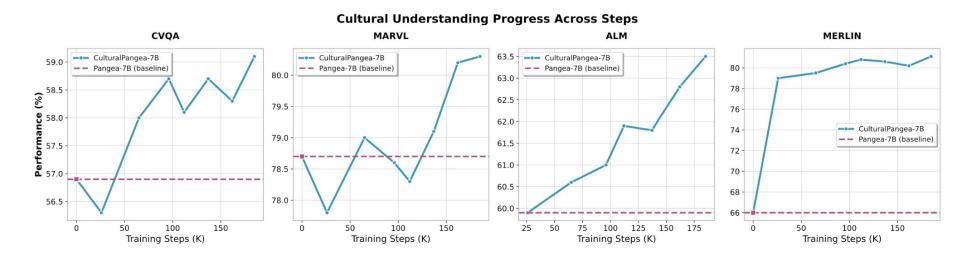
Which Cultural Domains Benefit Most?





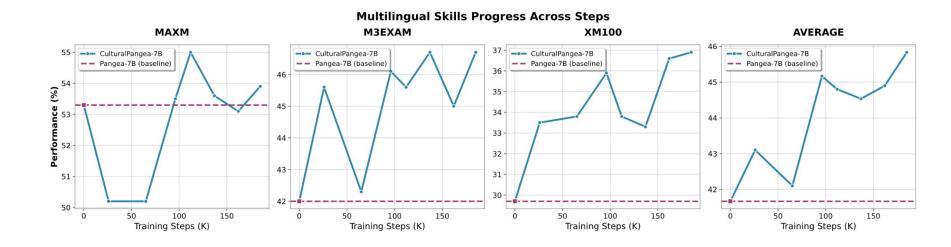
We find that CulturalPangea performs well on nearly all cultural domains and top gains are observed on cultural-rich domains(heritage, media, music, food, architecture, etc..) while minimal or negative gains are on general domains(memes, sketch)

Cultural Data Scaling and General Skill Preservation



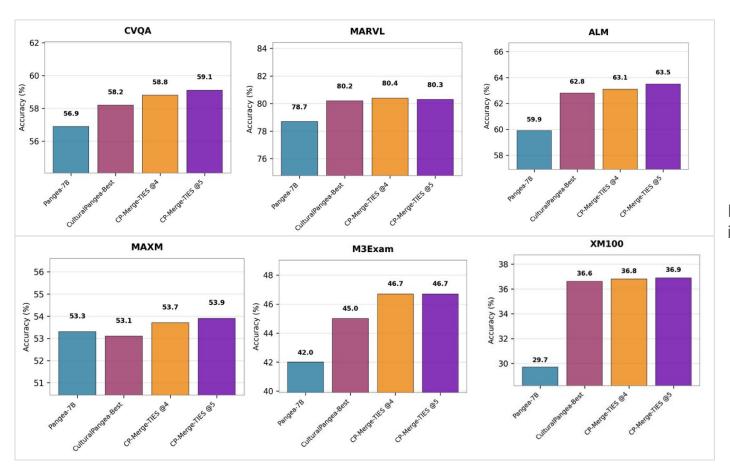
Performance on culture-centric benchmarks consistently increase with more cultural data and general multilingual vision–language proficiency is concurrently preserved and even enhanced. Our training methodology(interleaving CulturalGround and Pangealns) effectively alleviate catastrophic forgetting.

Cultural Data Scaling and General Skill Preservation



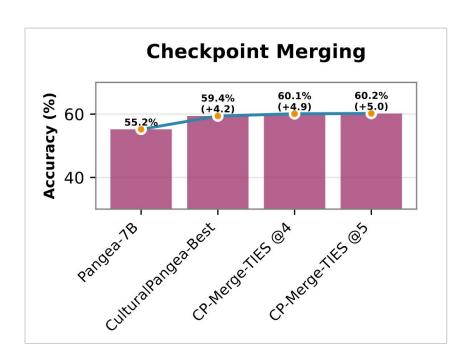
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Different Checkpoints have Individual Strengths



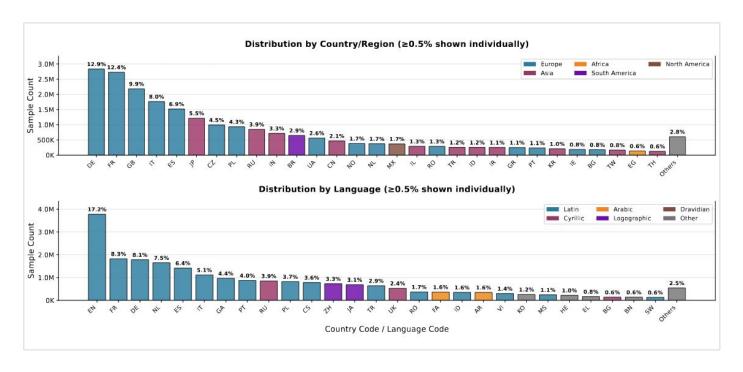
Different checkpoints have individual strengths across different languages and benchmarks.

Performance Gains from Merging Models



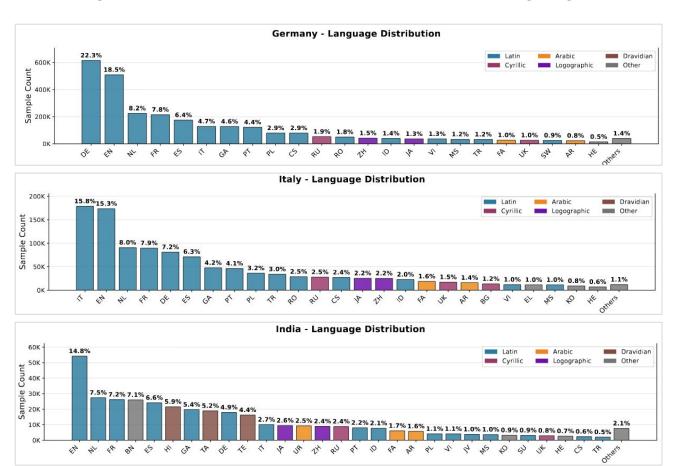
Merging checkpoints at different stages improves performance and recovers complementary model strengths often lost during continual training.

Taking a Look at CulturalGround: Distribution by Country/Region



CulturalGround covers 42 regions from different continents, 39 languages of different scripts

Taking a Look at CulturalGround: Diverse Languages Within Regions

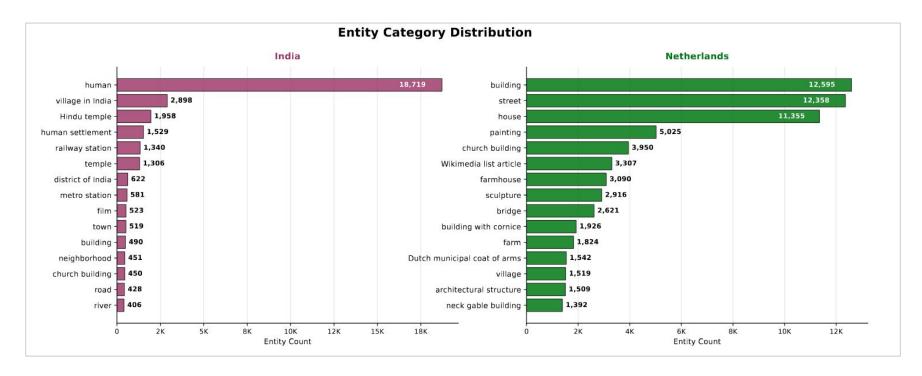


CulturalGround entities are represented in diverse set of languages and most regional entities are represented by regional-rich languages.

German: DE, EN, FR, ES, IT Italy: IT, EN, NL, FR, DE India: EN, FR, BN, HI, TA, TE

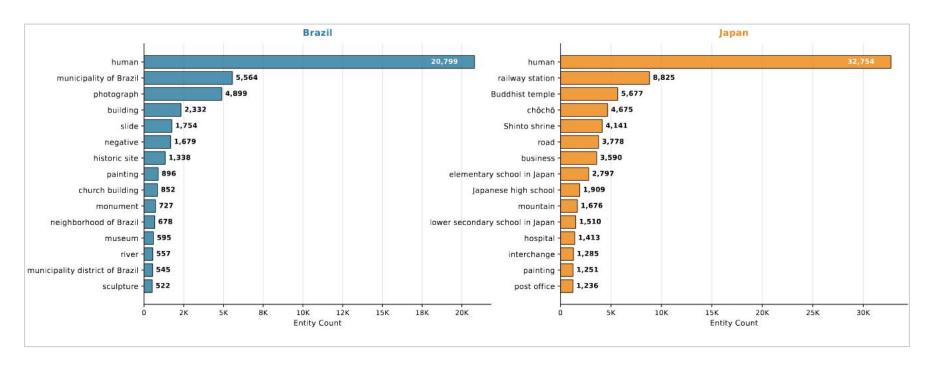


Taking a Look at CulturalGround: Diverse Culture Categories



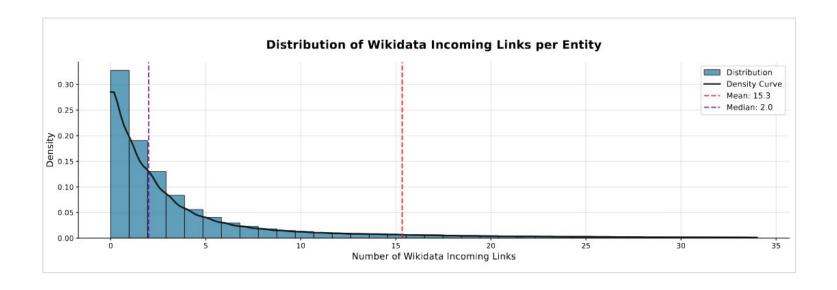
Entities in CulturalGround are of wide range of cultural domains and regional specific domains sometimes: people, church, farm house, district of, temple, etc...

Taking a Look at CulturalGround: Diverse Culture Categories



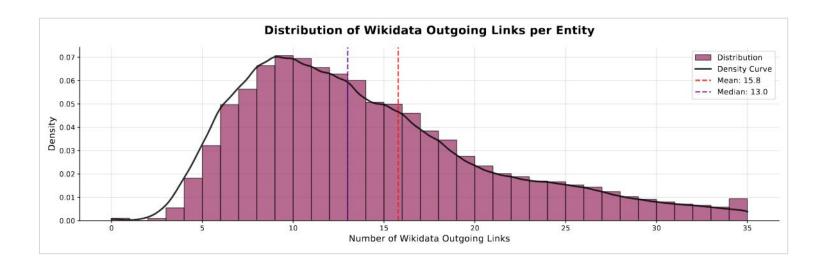
Entities in CulturalGround are of wide range of cultural domains and regional specific domains sometimes: people, church, farm house, district of, temple, etc...

Taking a Look at CulturalGround: Long-tail Entities Coverage



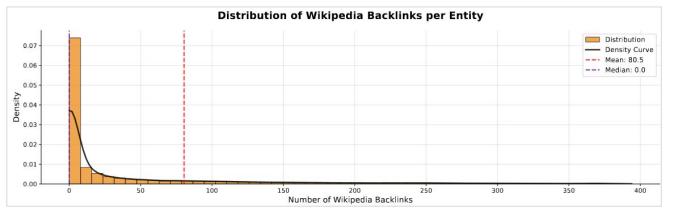
The majority of entities have very few incoming links in Wikidata. Incoming links are other entities that link to a given entity.

Taking a Look at CulturalGround: Long-tail Entities Coverage

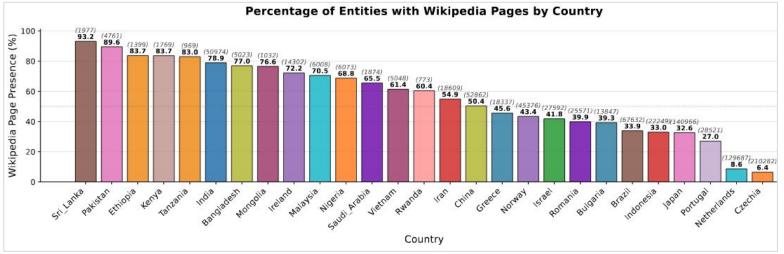


Entities have a good number of outgoing links(entity properties). This is great since these are essentially what we use in data curation.

Taking a Look at CulturalGround: Long-tail Entities Coverage



The majority of entities doesn't have Wikipedia pages and few Wikipedia backlinks



Taking a Look at CulturalGround

CulturalGround(Spanish)



castro

Entity Info

Language: Spanish Entity name: castro Country: Spain

Wikidata ID: Q5050823

Question type: Property-level question

Property: P2596 (culture)

Template question: ¿Con qué cultura está asoci-

ada esta entidad?

(Which culture is this entity associated with?) Template answer: El *castro* está asociado a la cultura castreña, una cultura arqueológica ibérica.

(The *castro* is associated with the Castro cul-

ture, an Iberian archaeological culture.)

Question

¿A qué cultura o tradición arqueológica se asocia normalmente este tipo de asentamiento? (What culture built these kinds of settlements?)

Answer

El *castro* está asociado a la cultura castreña, una cultura arqueológica propia del noroeste de España. Estos asentamientos fortificados son característicos de la época prerromana y reflejan las formas de vida de las comunidades locales antes de la llegada del Imperio Romano.

(This *castro* is associated with the Castro culture, an archaeological culture primarily found in northern Spain. These fortified settlements were often constructed before the Roman period and represent a significant part of the region's pre-Roman history.)

Taking a Look at CulturalGround

CulturalGround(French)



Marie Déa

Entity Info

Language: French Entity name: Marie Déa Country: France

Wikidata ID: Q3292505

Question type: Property-level question Property: P1412 (languages spoken / written) Template question: Quelle(s) langue(s) cette entité parle-t-elle ou écrit-elle ?(Which language(s)

does this entity speak or write)

Template answer: Marie Déa parle ou écrit français, langue romane.(Marie Déa speaks or

writes French, a Romance language)

Question

Quelle langue parle ou écrit l'actrice française que vous voyez sur cette photo ? (Which language does the French actress you see in this photo speak or write?)

Answer

Marie Déa parle et écrit en français, la langue romane de France. (Marie Déa speaks and writes French, the Romance language of France.)

Summary

- Multimodal LLMs lacks understanding of cultural nuances and this is factual cultural data issue
- We introduce CulturalGround, large-scale cultural/multilingual VQA dataset covering 42 regions, 39 languages
- CulturalPangea, a model trained on subset of CulturalGround achieves SOTA results on many cultural benchmarks
- We open-source all project artifacts: dataset at all stages of the curation and the model.

