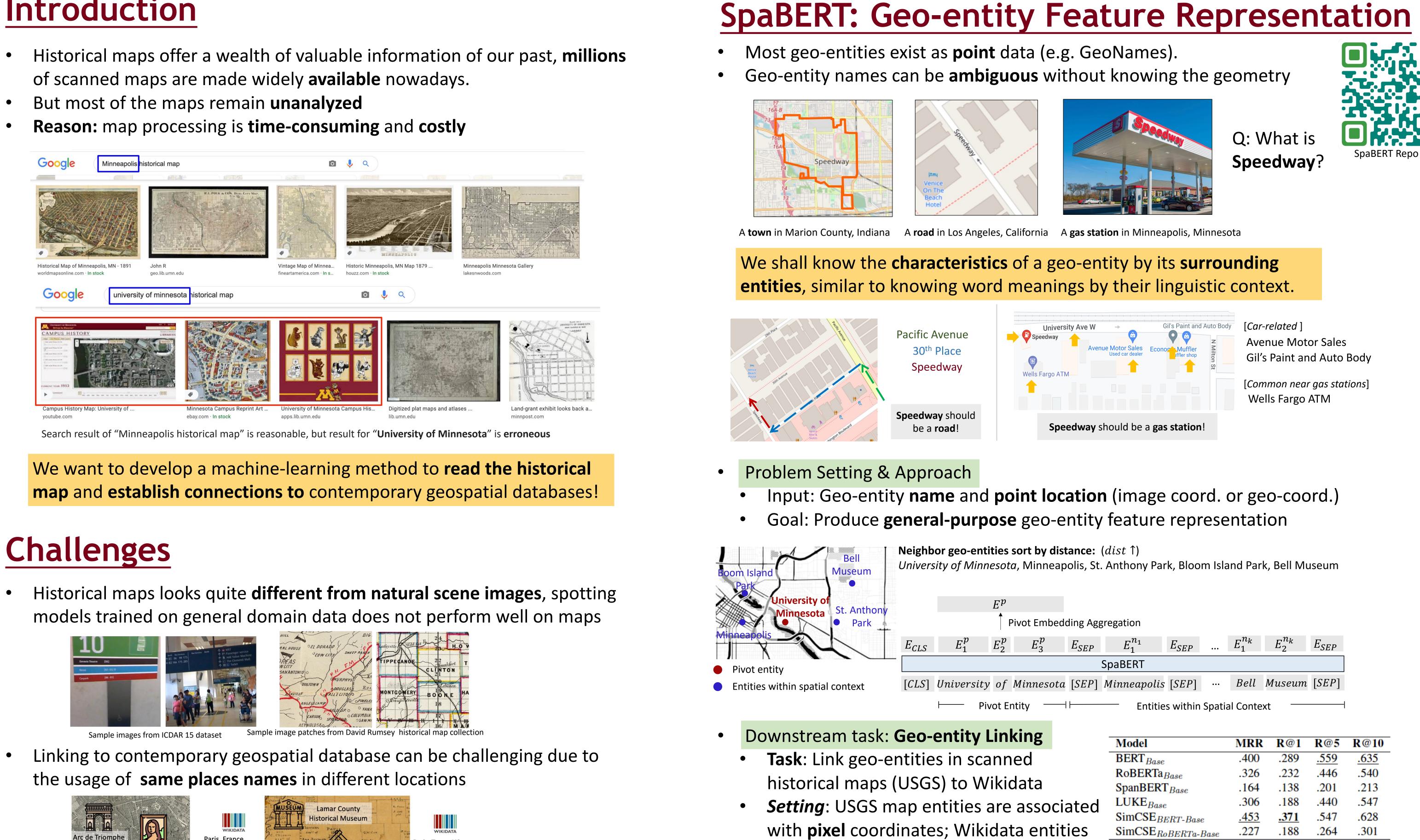
GEOSPATIAL DATA UNDERSTANDING: A Peek into Historical Maps and Contemporary Geospatial Databases **Zekun Li**, Department of Computer Science and Engineering

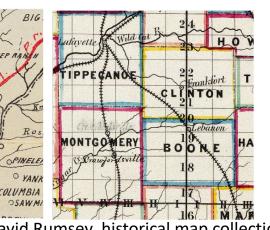
Introduction

- of scanned maps are made widely **available** nowadays.
- But most of the maps remain **unanalyzed**
- **Reason:** map processing is **time-consuming** and **costly**

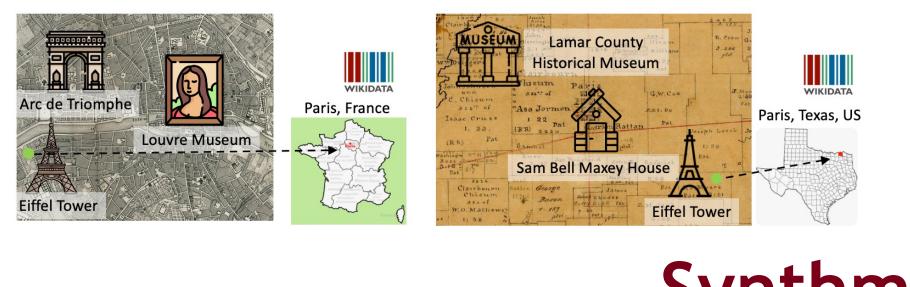


Challenges

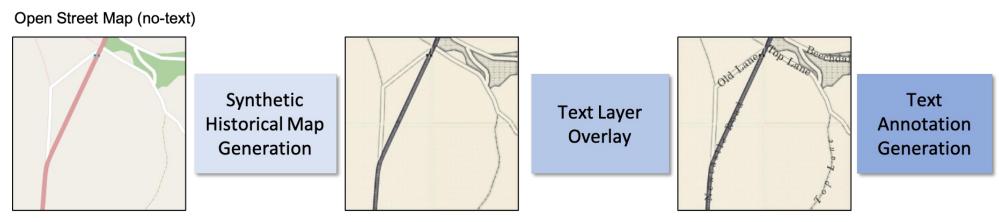




the usage of same places names in different locations

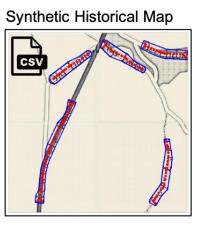


- Gathering training data for historical maps is important, while manual annotation takes a lot of time and effort
- We propose to generate **synthetic historical maps** to aid the training of text detection models
- General Idea:
 - Create synthetic map **background without** any **text labels**
 - Automatically **place text labels** and compute ground-truth annotation (of text bounding polygon)



Synthmap: Generate Synthetic Historical Maps





Source map: Clean (no text) OpenStreetMap tiles to provide background **Target** map style: **Ordnance Survey** 6-inch map during year 1888-1913

are associated with **geo-coordinates**

Model: CycleGAN to efficiently perform style transfer

Historical OSM *G*(*s*) • F(t) Ad 1997

B.H. 204-1 Lake'y Church Old Quarry Buring Progen Mark End



Cycle Consistency Loss $\mathcal{L}_{\text{cyc}}(G, F) = \mathbb{E}_{x \sim p_{\text{data}}(x)}[\|F(G(x)) - x\|_1]$ $+ \mathbb{E}_{y \sim p_{\text{data}}(y)}[\|G(F(y)) - y\|_1].$

 $\mathcal{L}_{ ext{GAN}}(G, D_Y, X, Y) = \mathbb{E}_{y \sim p_{ ext{data}}(y)}[\log D_Y(y)]$

 $+ \mathbb{E}_{x \sim p_{\text{data}}(x)}[\log(1 - D_Y(G(x)))]$

Adversarial Loss:

Q: What is **Speedway**?

[Car-related] Avenue Motor Sales Gil's Paint and Auto Body

[Common near gas stations] Wells Fargo ATM

ding Aggregation										
E _{SEP}	$E_{1}^{n_{1}}$	E _{SEP}		$E_1^{n_k}$	$E_2^{n_k}$	E _{SEP}				
SpaBERT										
[SEP]	Minneapolis	[SEP]		Bell	Museum	[SEP]				
	Entitie	s within	Spatia	al Conte	xt —					

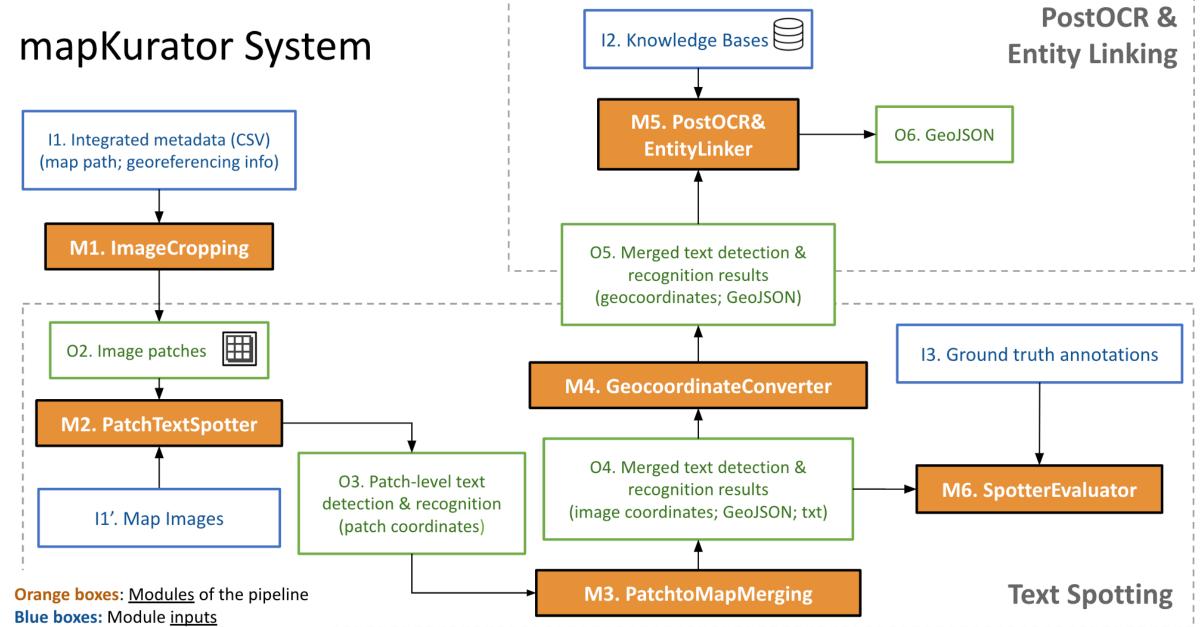
	Model	MRR	R@1	R@5	R@10
	BERT _{Base}	.400	.289	.559	.635
	RoBERTa _{Base}	.326	.232	.446	.540
	SpanBERT _{Base}	.164	.138	.201	.213
ated	$LUKE_{Base}$.306	.188	.440	.547
	SimCSE _{BERT-Base}	.453	<u>.371</u>	.547	.628
ies	SimCSE _{RoBERTa-Base}	.227	.188	.264	.301
	SpaBERT _{Base}	.515	.338	.744	.850

*Details for automatically computing the annotation info can be found in the paper

Complete map after overlaying text



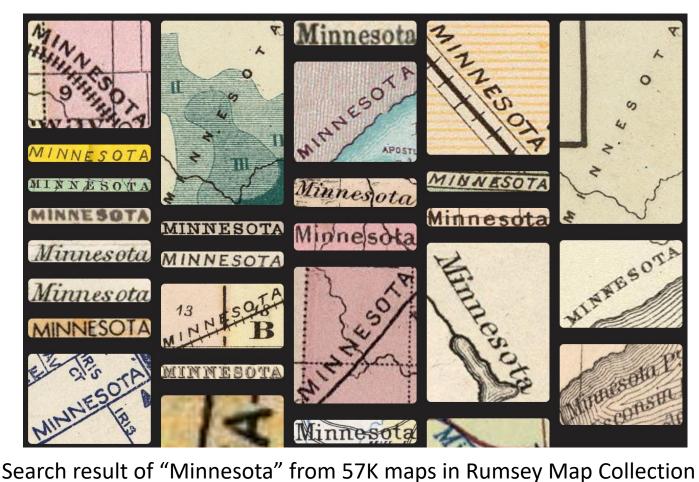








rom "Map Of California And Nevada" by Geological Survey of California



Conclusion

References

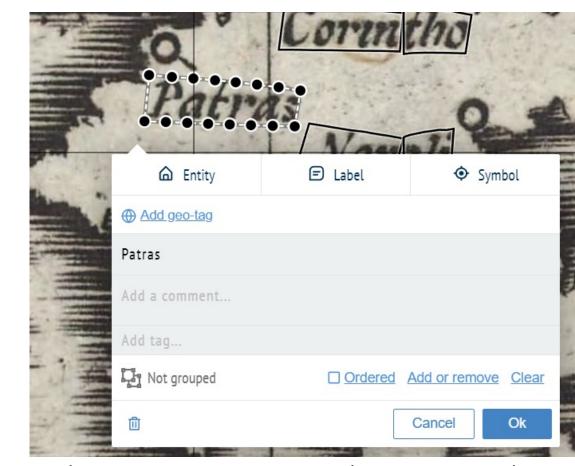






mapKurator: Historical Map Understanding

Inputs: Historical map images (.png/.geotiff) **or** metadata providing map path **Outputs:** Recognized text labels & label bounding polygons & Identifier to OSM



Display mapKurator spotting result in Recogito web interfa

SynthMap, a dataset of synthetic historical map images generated from OSM tiles using cycleGAN to help improve text detection.

SpaBERT, a BERT-based language model to capture the relations between 2D geo-entities and produce spatial-context-aware features.

mapKurator, a machine learning system for historical map understanding.

[1] Li, Zekun. "Generating historical maps from online maps." Proceedings of the 27th ACM SIGSPATIAL International Conference on Advances in Geographic Information Systems. 2019.

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[3] Li, Zekun, et al. "SpaBERT: A Pretrained Language Model from Geographic Data for Geo-Entity Representation." Proceedings of the EMNLP. 2022.

[4] Li, Zekun, et al. "An automatic approach for generating rich, linked geo-metadata from historical map images." Proceedings of the 26th ACM SIGKDD. 2020.

POWERED BY LUNA IMAGING

Acknowledgement

JARTOGRAPHY ASSOCIATES





David Rumsey Map Collection

Knowledge Computing Lab UNIVERSITY OF MINNESOTA

The

Institute

Alan Turing