

Territorial ratemaking with unsupervised embeddings

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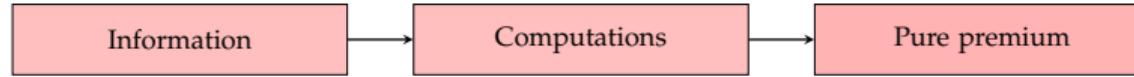
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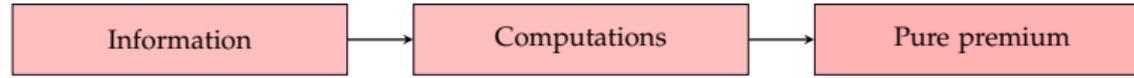


Quantact

Classical ratemaking



Territorial homeowners ratemaking



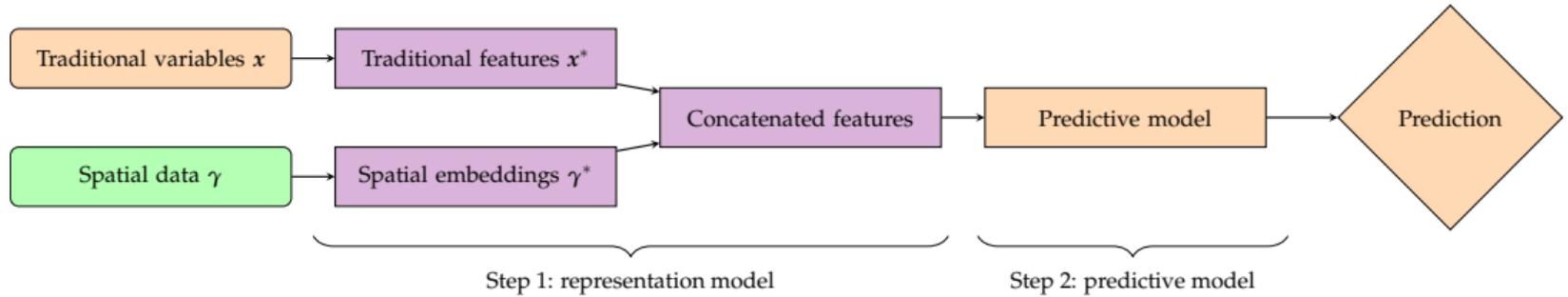
Territorial ratemaking

- Typical method for territorial ratemaking: spatial interpolation
 - 1 Smoothing risks based on coordinates (loess, splines, kriging, spRF)
 - 2 Averaging past losses based on polygons
- Control flexibility: compromise between credibility and homogeneity
 - ▶ Too flexible: few observations = noisy data, overfit
 - ▶ Too rigid: fail to capture local variations in spatial risk

Territorial ratemaking

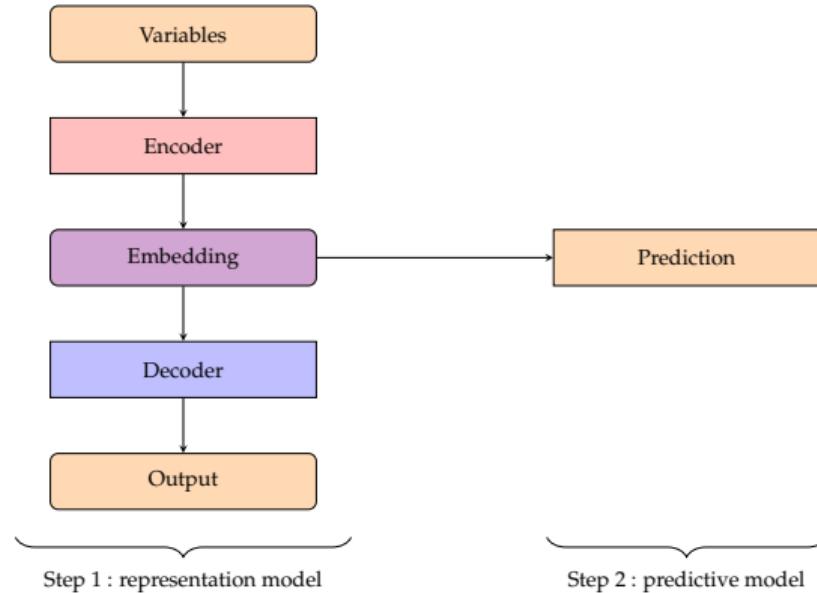
- Spatial embeddings = alternative to the credibility / homogeneity compromise
- Focus instead on what actually generates spatial **risk**
 - ▶ Landform
 - ▶ Weather
 - ▶ People
- Assumption: Since **people** generate **risk**, the spatial distribution of **people** relates to the spatial distribution of **risk**

Spatial embeddings



[Blier-Wong et al., 2020], [Blier-Wong et al., 2021b]

Spatial embeddings



[Blier-Wong et al., 2021a]

Spatial embeddings

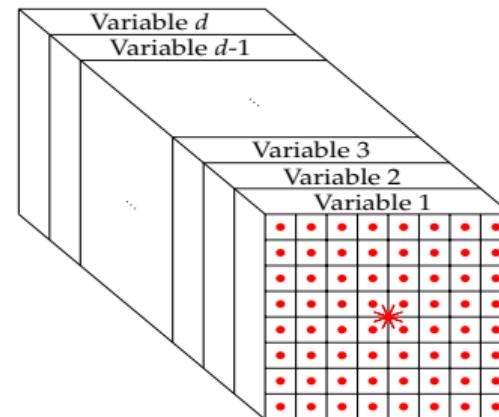
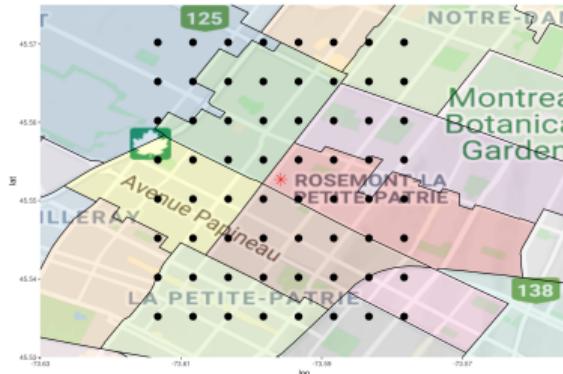
Desirable attributes of spatial embeddings

- We define three desirable attributes for spatial embeddings
 - 1 Spatial embeddings must follow Tobler's first law of geography
 - 2 Spatial embeddings are coordinate-based
 - 3 Spatial embeddings encode relevant external information
- A simple ratemaking model using spatial embeddings as input will inherit the desirable attributes

Spatial embeddings

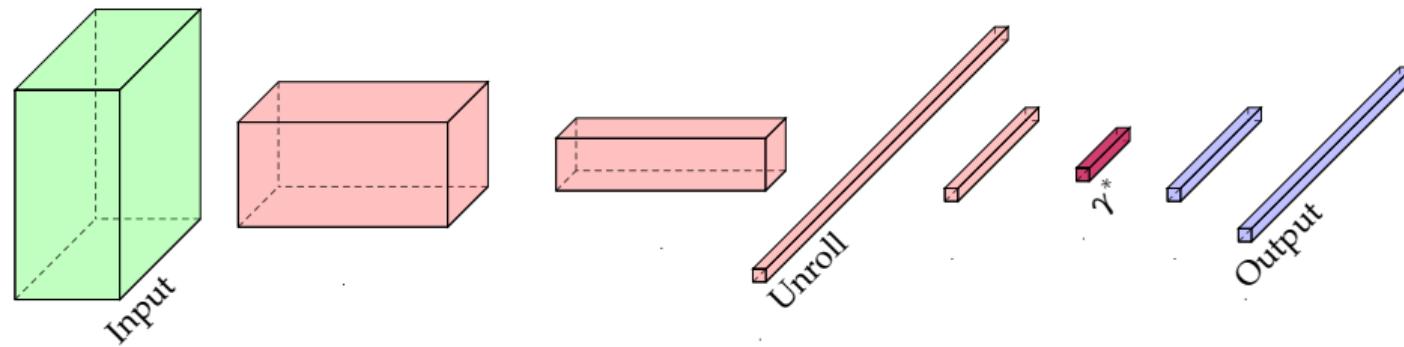
Input to the encoder

Input data = geographic data square cuboid



Spatial embedding construction

Encoder + decoder



Application on insurance data

Spatial ratemaking

We compare GAM (bivariate spline) with GLM (with spatial embeddings)

- Accident frequency prediction
- Home insurance in Québec
- Over 2 000 000 contracts

Poisson GAM:

$$\ln(E[Y_i]) = \beta_0 + \ln \omega + \underbrace{\sum_{j=1}^p x_{ij}\alpha_j}_{\text{traditional component}} + \underbrace{f_k(\text{lon}_i, \text{lat}_i)}_{\text{spline component}} + \underbrace{\sum_{j=1}^{\ell} \gamma_{ij}^* \beta_j}_{\text{embedding component}}$$

Spatial ratemaking

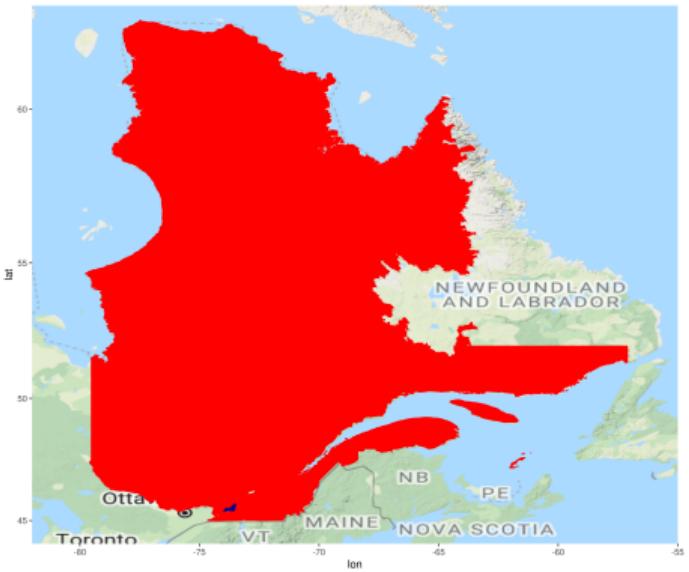
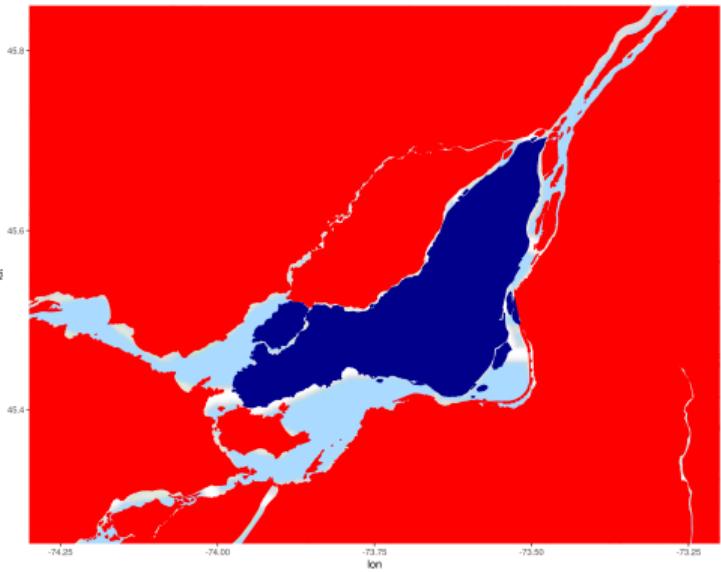
GLM vs GAM in Québec

Test deviance in Québec

k	Without embeddings		With embeddings	
	Test	DoF	Test	DoF
0	–	–	80663	19
3	80766	7.18	80785	26.16
5	80767	18.63	80780	35.18
8	80837	43.97	80777	60.43
10	80771	64.89	80748	82.86
15	80853	126.60	80825	138.64
20	80852	181.65	80825	191.97

Spatial ratemaking

Out-of-sample prediction



Spatial ratemaking

Out-of-sample prediction

- OOT : Out-of-territory + embeddings (red)
- GAM : Within territory + spatial splines (blue)
- C.P. : Center of population

C.P.	OOT	GAM	C.P.	OOT	GAM
Montréal	14364	14400	Sherbrooke	2754	2782
Québec	3787	3801	Saguenay	958	963
Laval	4046	4043	Levis	2500	2500
Gatineau	6406	6495	Trois Rivières	2952	2961
Longueuil	3203	3185	Terrebonne	2656	2677

Conclusion

Conclusion

- Easily incorporate spatial effects within predictive models
- Spatial embeddings improve (not replace) spatial models
- Other applications of spatial embeddings
 - ▶ in insurance: fraud detection, claims management, life insurance
 - ▶ in general: urban planning, crime prediction, election forecasting
- Arxiv paper: [Blier-Wong et al., 2021b]
- Thanks for your attention!
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- Icons from Freepik and fjsstudio on flaticon.com

References I

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