A sensitivity analysis concerning spatial density estimation with MNO data

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Repository and references

Research Intent

Mobile Network Operator (MNO) data offer a rich source for estimating the spatial distribution of mobile phones at a given point in time. We compare two kinds of estimation strategies: (1) the traditional **Voronoi tessellation and variants thereof**, which exclusively assume that every mobile phone connects to its nearest cell (deterministic), (2) recently introduced estimators that are based on **radio propagation modeling (RPM)** through additional cellular network information (probabilistic). We know that RPM pays off in terms of spatial accuracy if the model information is of perfect quality, however, how high-quality does the model information need to be?

For this, we have developed two research questions:

- ❖ *RQ1: How robust are estimation strategies to network model mismatching errors (information quality)?*
- ❖ *RQ2: How sensitive are estimation strategies to network characteristics, such as cell density?*

Simulation Study Design

Toyworld Generation

Estimation

Evaluation

- 100m x 100m grid (1,600 km², right-tailed population distribution, Munich)
- For RQ2, **two synthetic cellular networks** (dense and sparse)
- Mimicking cell counters and developing the ground truth RPM

- **Voronoi estimators:** Tower (VOR_t), antenna offset (VOR_o), barycenter (VOR_b)
- **RPM estimators:** MLE, DF
- For RQ1, **model mismatch** introduced through spatially sensitive random noise

- **Kantorovitch-Wasserstein distance (KWD)** to measure the degree of horizontal spatial error
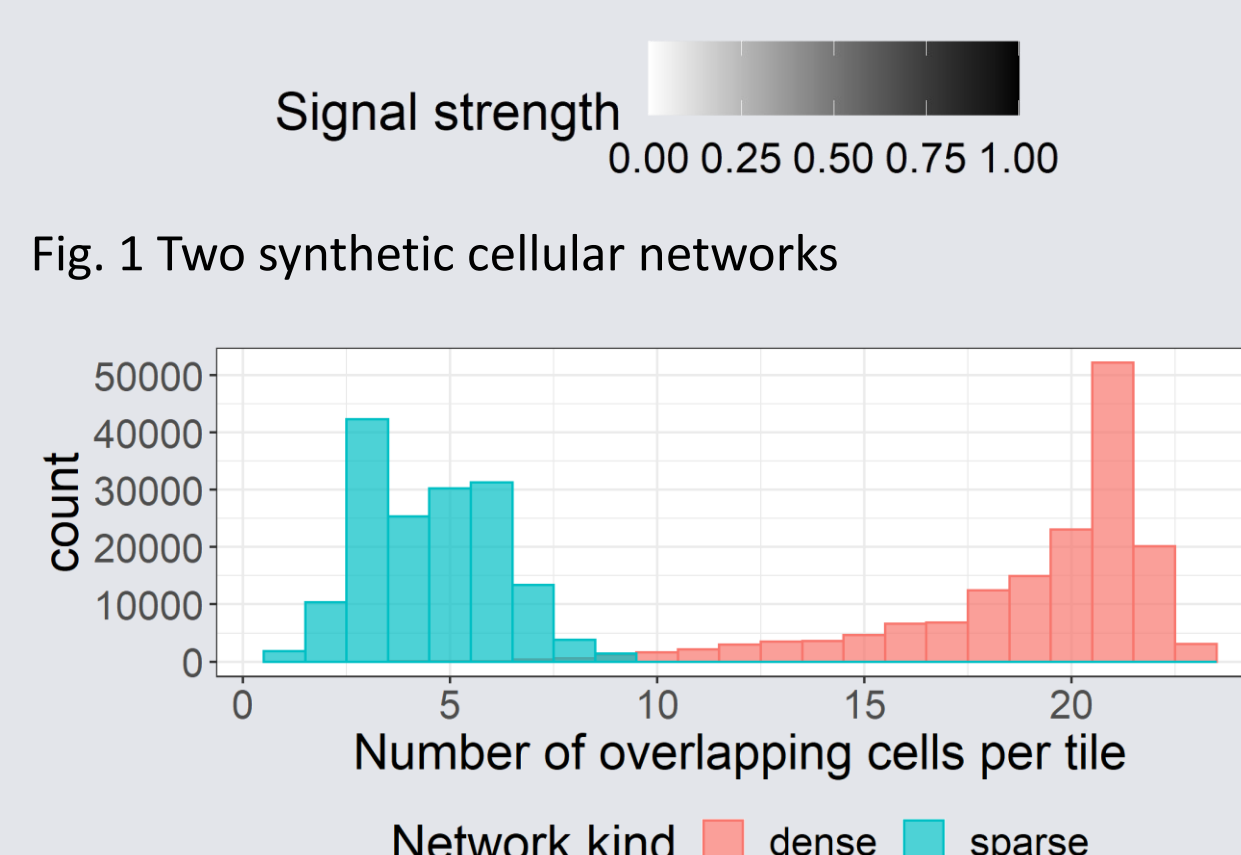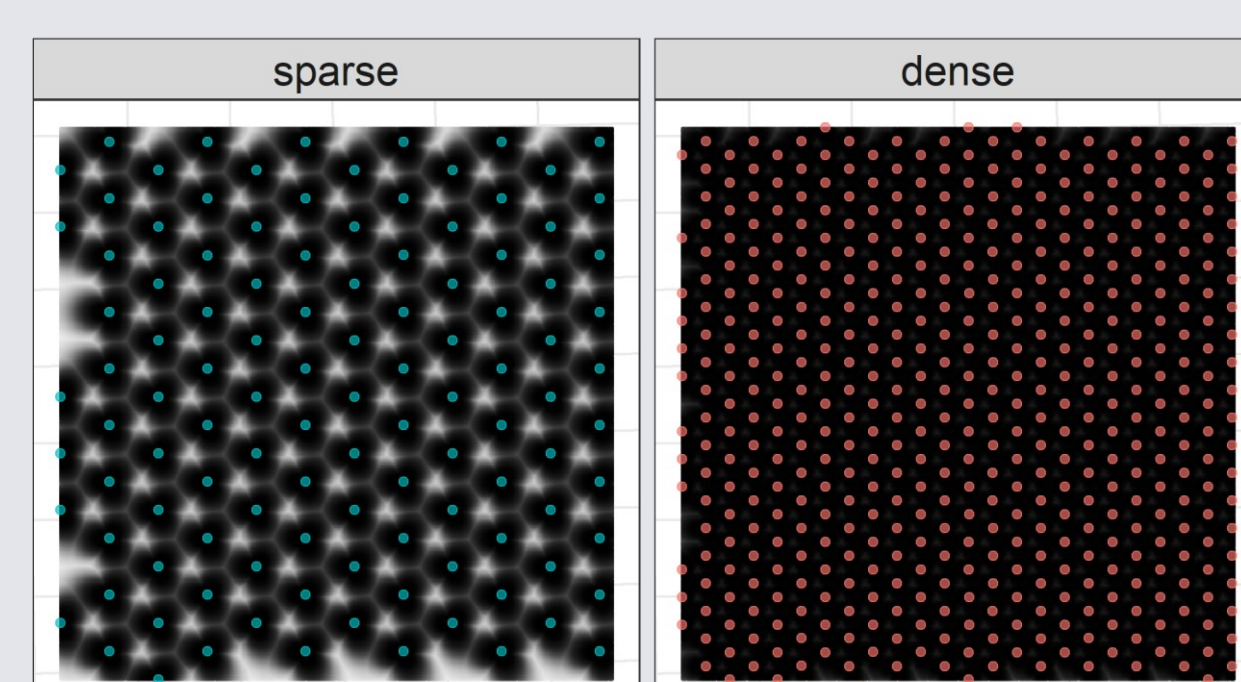- Spatial density maps

Key Results

A) Network characteristics

To reserach the effect of network characteristics, we developed two networks of different directional cell-density levels (Fig. 1). Each network leads to a different distribution of overlapping cells per 100m x 100m tile (Fig. 2).

Fig. 1 Two synthetic cellular networks

Fig. 2 Cell overlap per tile

Fig. 4 KWD values for all simulated versions

B) Model mismatch

To research the effect of cellular network information quality, we introduce spatially sensitve, random noise that distorts the modeled cell coverage profiles. These noised distortions resemble shaded versions of information quality.
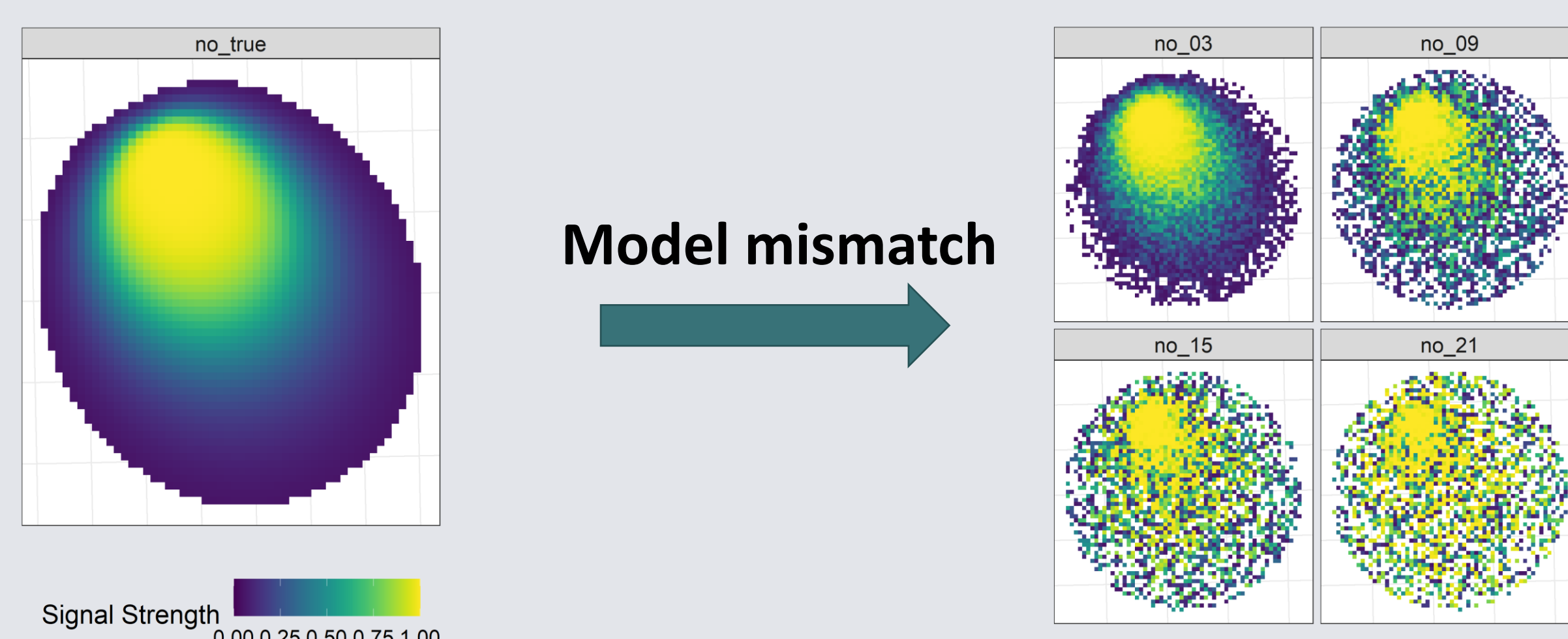

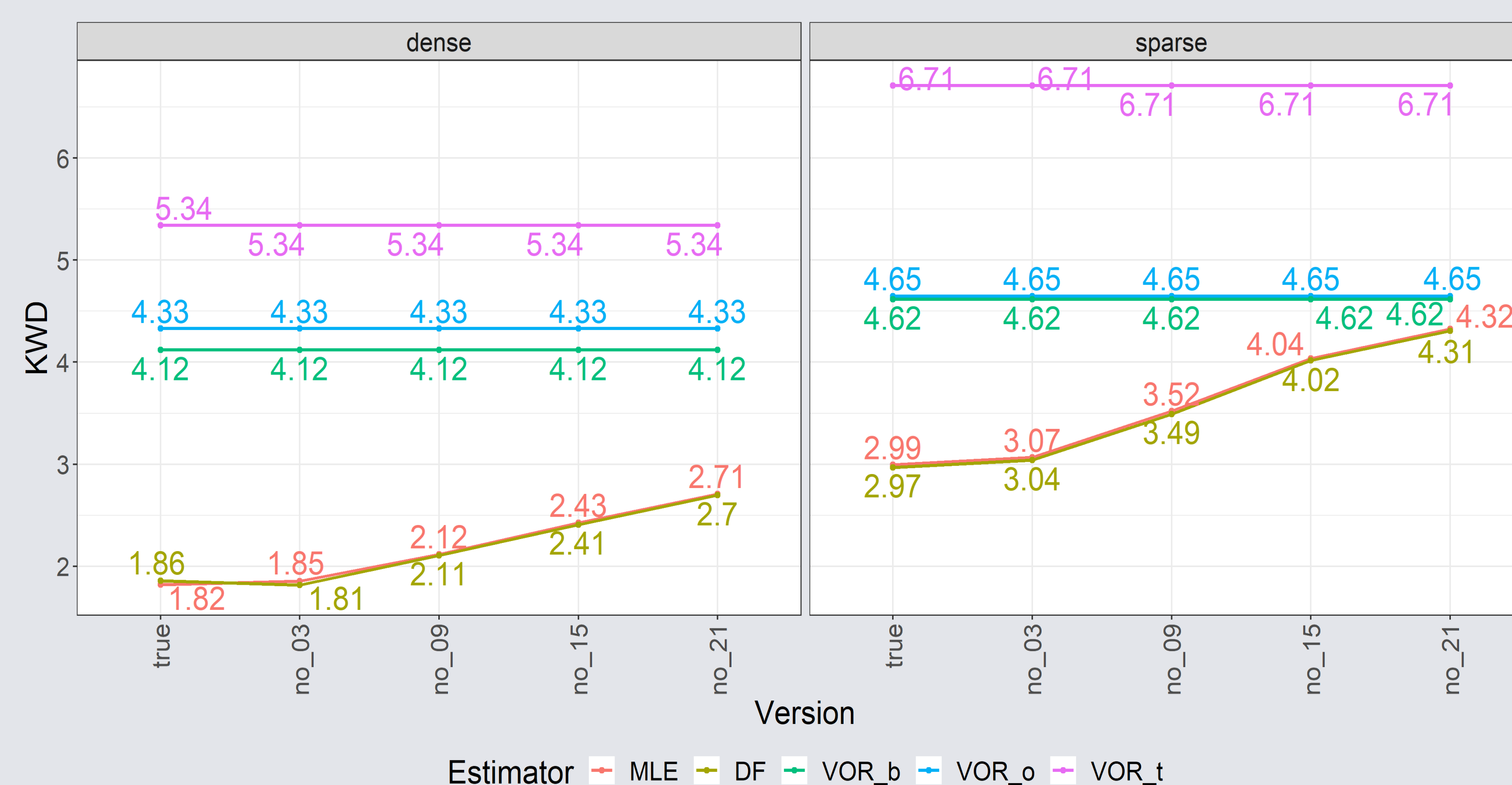Fig. 3 Model mismatch on exemplary cell profile through random noise

C1) Evaluation

Lower KWD values resemble **higher similarity** to the ground truth density in terms of spatial accuracy. Even in the sparse network scenario the RPM estimators with severely noised cell profile models perform much better (Fig. 4). The selected spatial density maps (Fig. 5) reveal how the model mismatch is translated into the estimation.
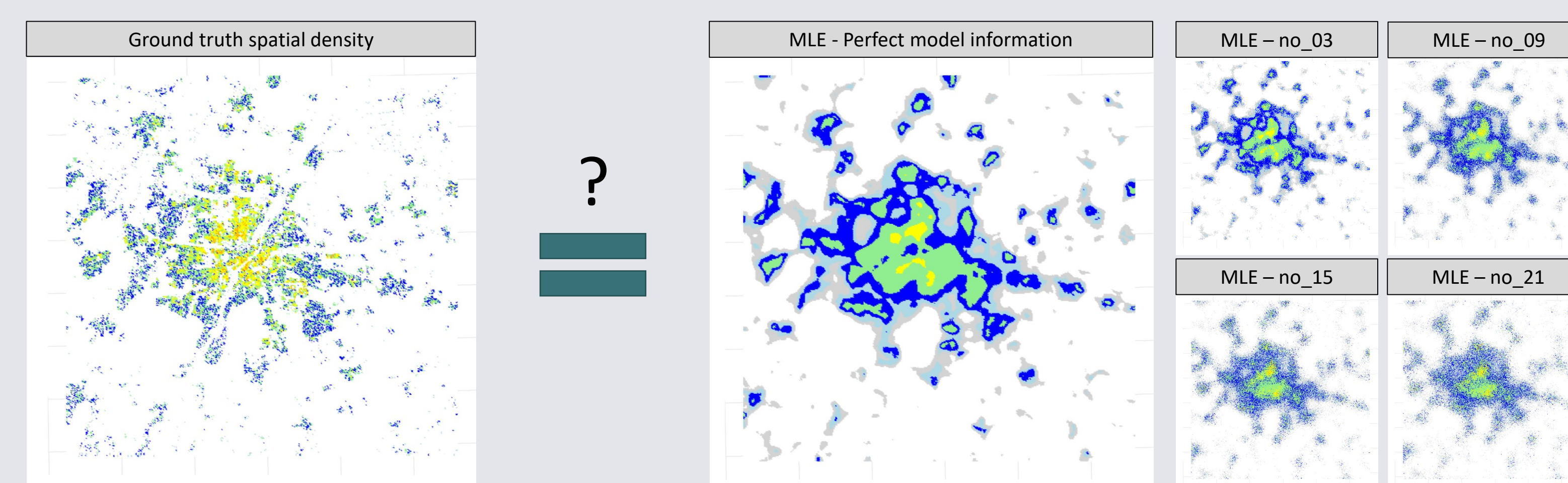

Fig. 5 Selected spatial density maps (comparing ground truth with MLE estimations of the dense network)

Key Takeaways

- *For RQ1: Even with severe model mismatch, RPM estimators lead always to spatially more accurate results than any Voronoi estimator.*

- *For RQ2: Higher network density offers more information and can therefore lead to spatially more accurate results than sparse networks. RPM estimators utilize this information more than Voronoi estimators.*

RPM is a complex task, however, in terms of spatial accuracy, it pays off because it allows to model overlapping cells. Our research shows that RPM estimators perform well under different network characteristics and are robust to severe model mismatching errors. Therefore, RPM estimators should be prefered as they utilize various information levels much better than Voronoi estimators. This research project will be extended through more networks (e.g. network layers) and mismatch versions (e.g. quantization, blurring).