



# BAYESIAN INVERSE PROBLEMS



► Inverse might not exist and if it does it might not be unique

▶ some images are more likely than others

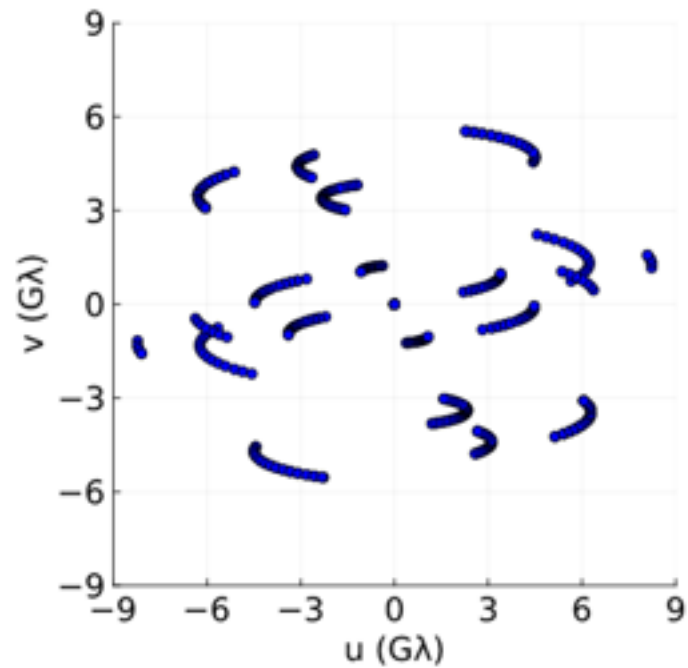
► **Solution:** Introduce a prior  $p(x)$  on the space of images  ~~$X$~~  corresponding to feasible

▶ e.g. log-gaussian priors with unknown kernel parameters

► Infer the image from the posterior  $p(x|y)$



- **Problem:** Data is sparse, noisy, temporally correlated, and in Fourier space



**Inference**



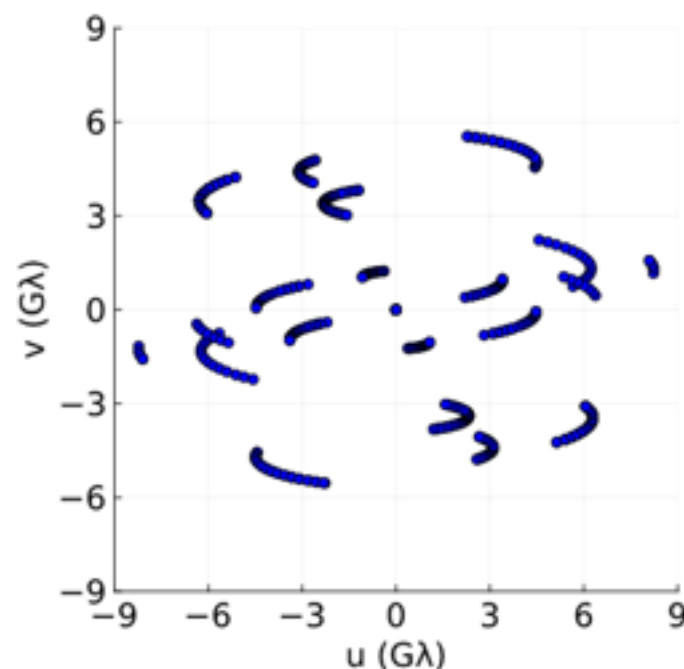
$$\sim p(x | y)$$

**M87, 2019**

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- ▶ **Problem:** Data is sparse, noisy, temporally correlated, and in Fourier space
  - ▶ Inverse might not exist and if it does it might not be unique
  - ▶ Some images are more likely than others
- ▶ **Solution:** Introduce a prior  $p(x)$  on the space of images  $\mathbb{X}$  corresponding to feasible
  - ▶ e.g. a log-gaussian process with unknown kernel parameters
  - ▶ Infer the image from the posterior  $p(x | y)$



Inference  
→



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$\sim p(x | y)$

# BAYESIAN WORKFLOW