



Robust Pose Graph Optimization Without An Accurate Measurement Covariance Model

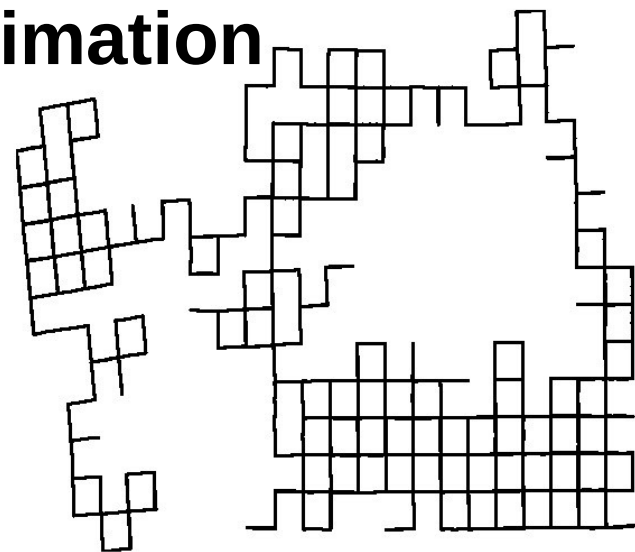
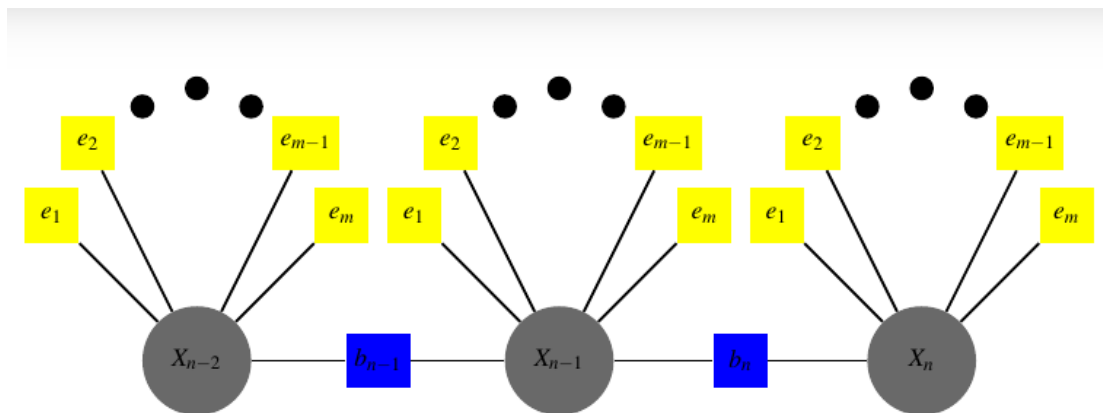
- **ATR Center Summer Workshop**
- **9,10 August 2017**

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Mentors: Robert C. Leishman, Clark N. Taylor



Introduction: What is a F.G.

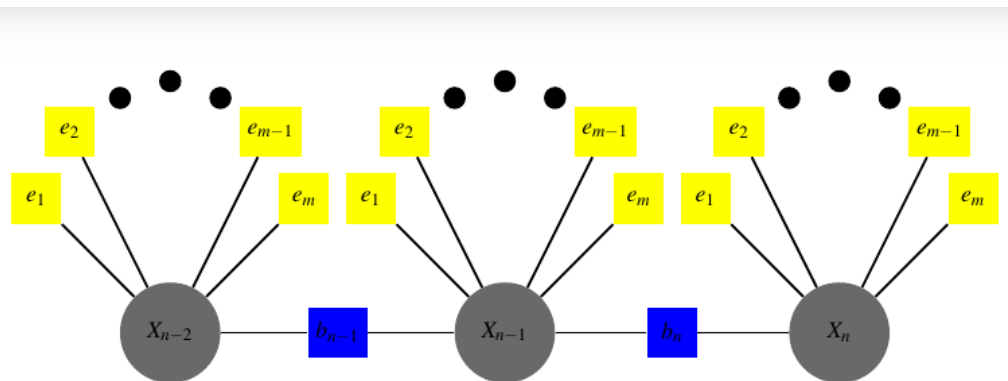
- What is a factor graph?
 - Tool to factorize a function of many variables into a product of smaller subsets
 - Factorization represented as a bipartite graph $G=(F,X,E)$
 - Can be applied to MAP estimation





Introduction: Solving a F.G.

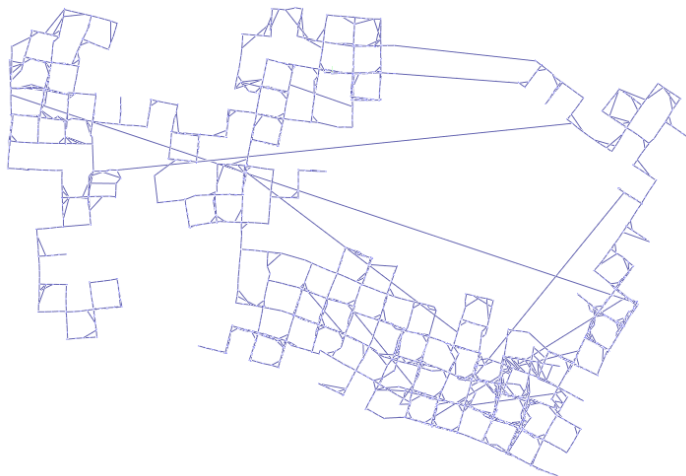
- When The Gaussian Assumption Holds ::



$$e_i = H_i(X_i) - Z_i$$

$$\hat{X} = \operatorname{argmin}_i \sum ||e_i||_{\Sigma}^2$$

- When The Gaussian Assumption Does Not Hold ::

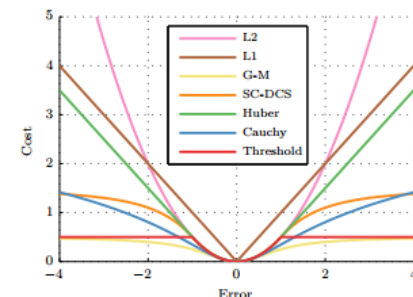




Previous Robust Methods

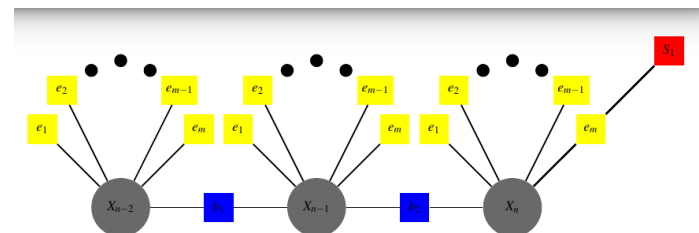
- **M-Estimator**

- Replace the traditional L^2 cost with a modified cost function



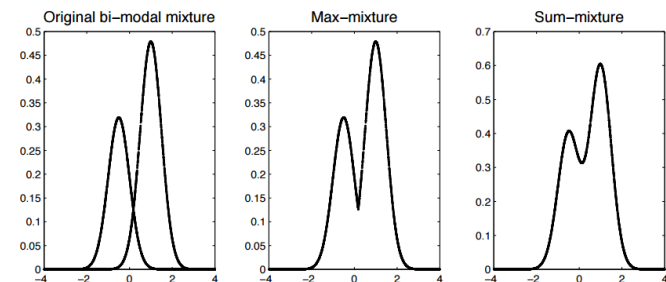
- **Switchable Constraints**

- The topology of the pose graph should be subject to the optimization.



- **Max-Mixtures**

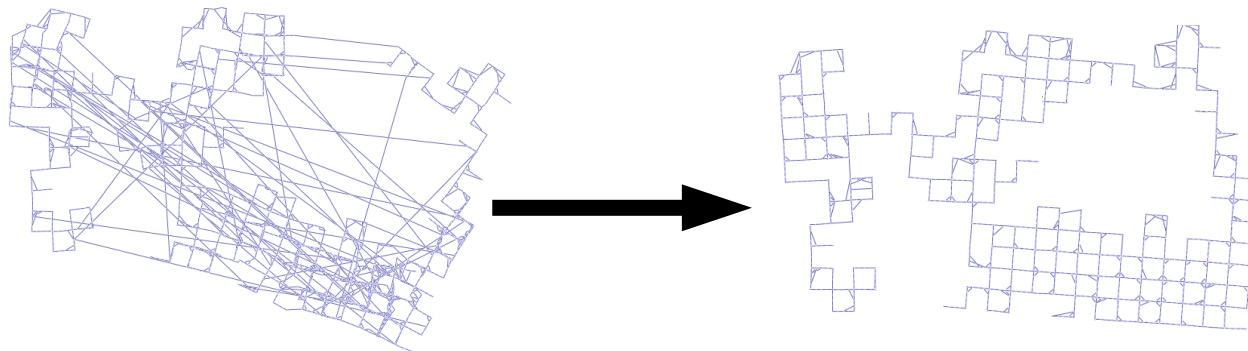
- Allows for the adoption of more realistic noise models through the max operator



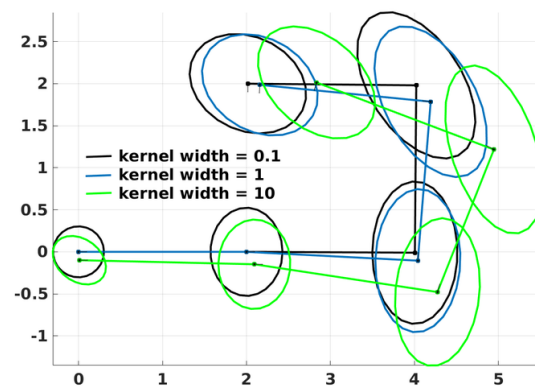


Motivation / Problem Statement

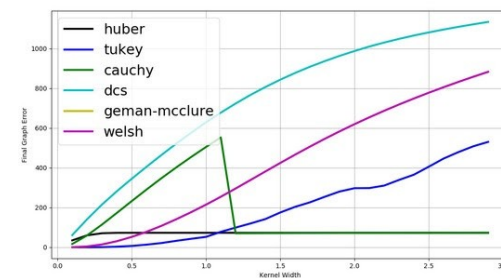
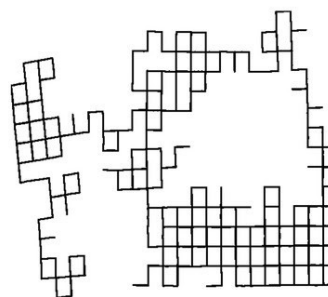
- **Robust**



- **Accurate Covariance**



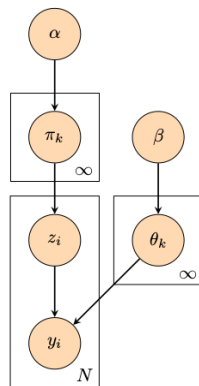
- **Insensitive to Hyper-Parameter**





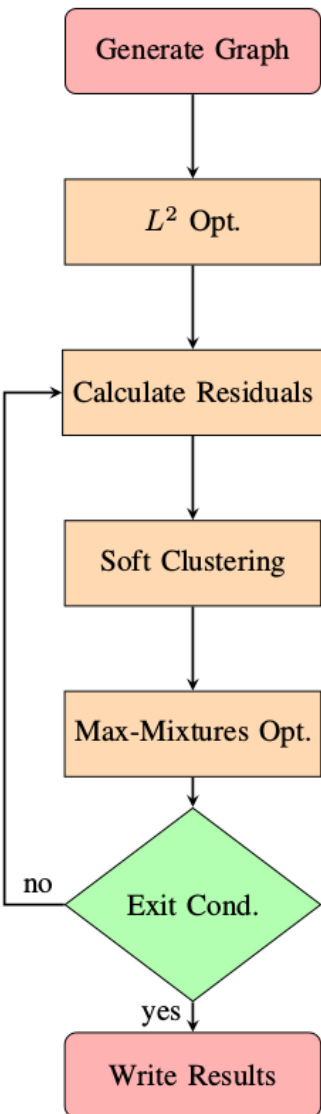
Our Approach

- Using collapsed Gibb's sampling to estimate Gaussian Mixture Model.



Define $p(\pi|\alpha)$ and $p(\mu_k, \Sigma_k|\beta)$ in such a way that we can analytically integrate out latent variables and only sample parameters of interest through collapsed Gibbs sampling

- Now, we have an n-component mixture that characterizes our measurement covariance.
- Now, all constraints are iteratively tested against the model, which allows the information and Jacobian matrices to be scaled accordingly.

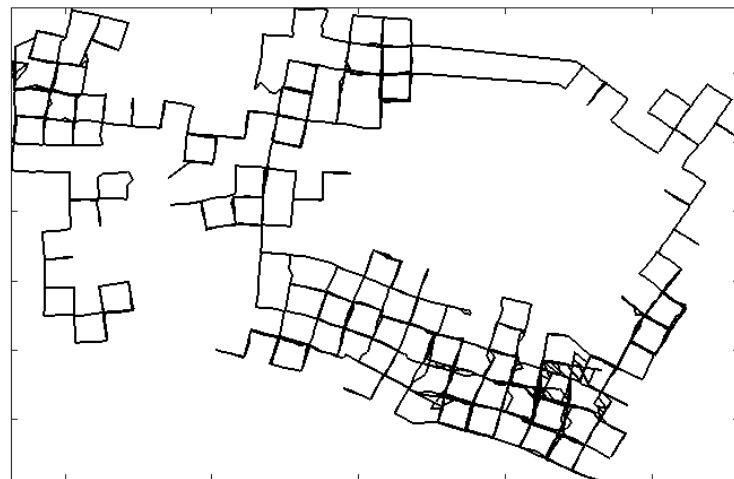
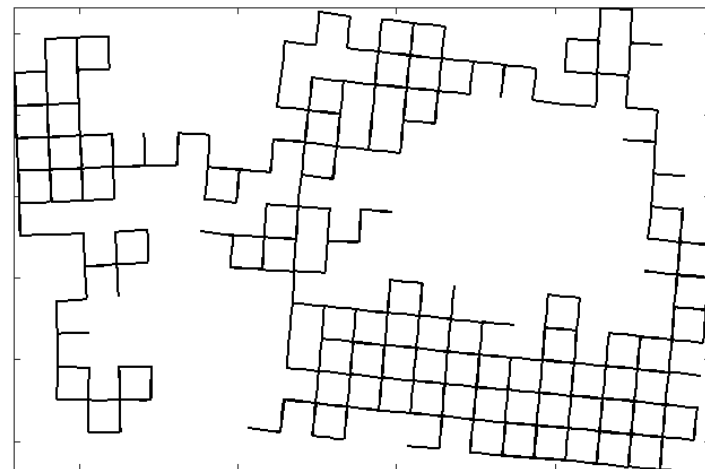




Results :: Robust



Soft Clustering
→
Robust Opt.

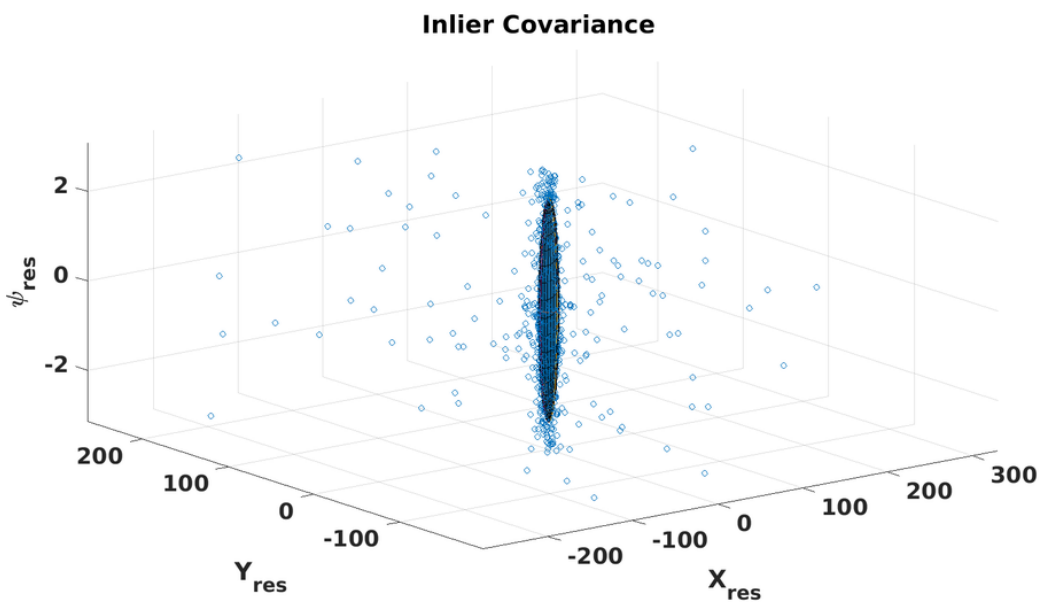


L^2 Opt.
→





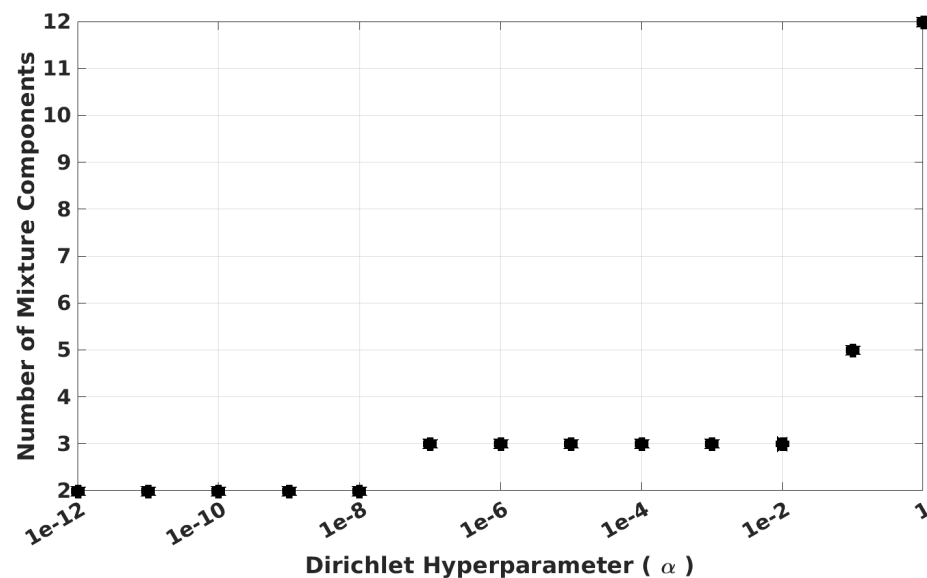
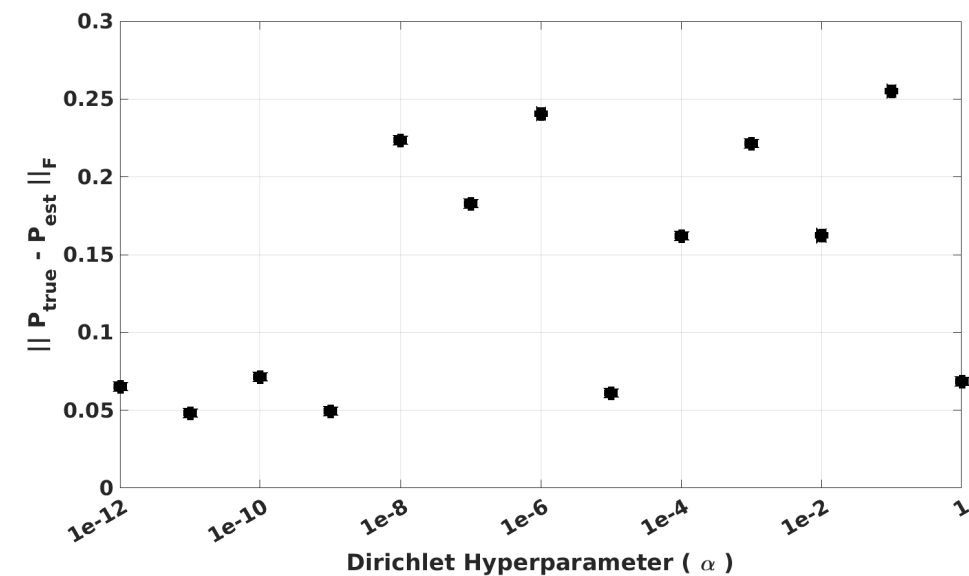
Results :: Accurate Covariance



	Inliers	Outliers	
Inliers	100.00%	0	100.00%
Outliers	2.00%	98.00%	100.00%
	102.00%	98.00%	



Results :: Hyper-parameter





Contributions, Future Work

- **Contributions:**
 - **Provided a robust pose-graph optimization routine that provides a reliable estimate of the inlier measurement covariance**
- **Future work:**
 - **Reduce run-time**
 - **Store residuals in a kd-tree**
 - **Replace compressed Gibbs Sampling with variational inference.**
 - **Scale final covariance using Nyman-Pearson lemma.**



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