Modeling Smooth Backgrounds at Collider Experiments With Log Gaussian Cox Processes

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Motivation

•BKG modeling is a key element in New Physics searches

• Typical method: analytic function fitting – possibly inefficient and complicated

•LGCP is fast, automated, with 1D count data input

Markov Chain Monte Carlo Fits

 Markov Chain Monte Carlo estimates non-normalized probability distributions

•A 2D MCMC for the marginal likelihood estimates the hyperparameters



Results

 Pull plots with 1000 seeds comparison for LGCP and MLE reference

• Ideally mean at 0, uncertainty [-1,1]





•the optimized hyper-parameters are defining the multivariate normal used for the posterior MCMC





non-homogeneous Poisson process, with log likelihood $\log p(X|\lambda) = \sum_{i=1}^{n} \log \lambda(x_i) - \int_{X_a}^{X_b} \lambda(x') dx'$ •Bayes' law yields

kernel - two hyper parameters

Monte Carlo integration

•The 16,50 and 84 percentiles of the chain defines the fit and 1 σ uncertainty band





• Further uncertainty estimations

Spurious Signal and Injection Test

• Test feasibility for smoothing

•Implementing in very low statistics analyses, like the ditau diphoton