

Summary of the Workshop

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Probe

For vertex $\{E, \mathbf{r}, t_0\}$, the PE counts received on the J_{th} PMT follows an inhomogeneous Poisson process from a time-dependent function R_j .

- Probe function: $R_j(t; E, \mathbf{r}, t_0) \rightarrow R(x, y, z, t)$
- Expected PE counts: $\lambda(x, y, z) = \int_0^T R(x, y, z, t) dt$. (marginal distribution)

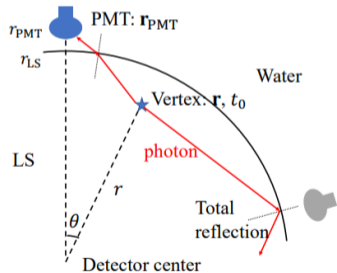


Fig 1: Geometric model for the probe

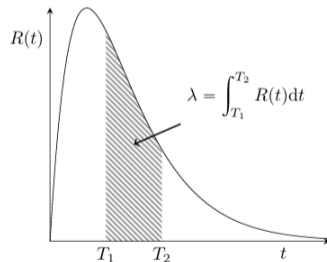


Fig 2: Schematic curve of $R(t)$

Generalized Additive Model(GAM) fitting

regression formulation of the expected PE counts $\lambda(x, y)$ and Probe function $R(x, y, t)$

$$nPE \sim \text{Poisson}(\lambda), nPE \sim \text{Poisson}(\lambda_{tbin})$$

$$\log \lambda = \text{te}(x, y, z) + \log(nEV)$$

$$\log R = \text{te}(x, y, z, t) + \log(nEV) + \log(tbinwidth)$$

- We use big MC dataset via **generalized additive model (GAM)** to fit the Probe function.
- We use the **bam** function of the **mgcv** package in R to perform the fitting.
- **bam** is a function that supports GAM for very large datasets.

limitation

As shown in the regression formulation, λ and R require separate fitting procedures, which may lead to insufficient accuracy in the fitting results.

Problem

- According to the mathematical relationship between λ and R , if we fit the integral of R , we only need one fitting to get them.
- The integral $\int_0^t R(\vec{r}, \tau) d\tau$ must increase strictly monotonically. But the `mgcv` package does not support monotonic spline fitting for tensor product functions.

Goal

- Realize the monotonic spline for tensor product in `bam` function by modifying the source code.

About **bam**

- Read two articles, Understood how it supports the GAM for very big datasets. Briefly, they used **iteration algorithm** for **blocked** matrix to deal with the big matrix.
- Reviewed key code blocks about the iteration algorithm.

About **scam**

- Found another package similar to mgcv, named scam. It provides the shape constrained GAM model.
- Got the constraint principles of scam via reading article and code.
- Found that scam supports tensor product smoothing constructor for a bivariate function monotone increasing in a certain covariate.
- Had successful experiments on 1-D and 2-D datasets.

2-D experiment results

$$\text{nPE} = \text{te}(z, t).$$

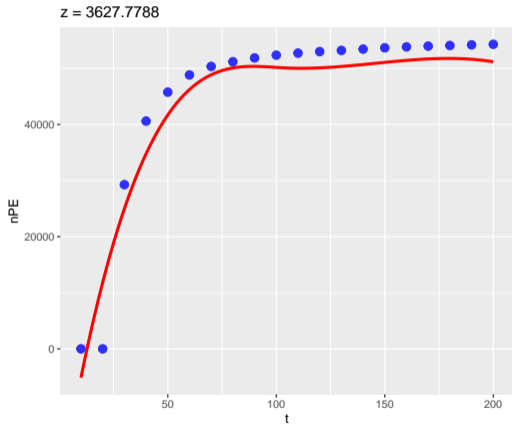


Fig 3: B-spline

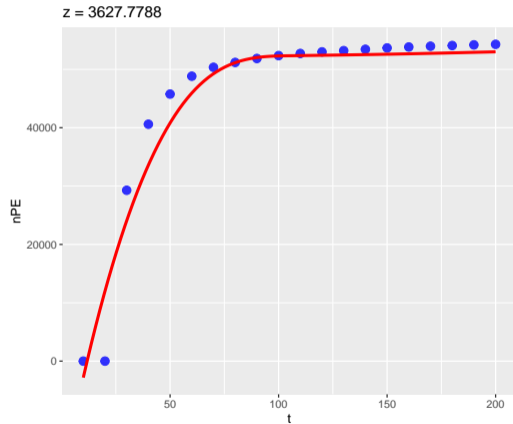


Fig 4: Monotonic spline in t-direction with scam

TODO

- 1 Focus on modifying the code of **scam** to realize the multivariate form.
- 2 Integrate **scam** modifications with **mgcv**'s **bam** function to support very large datasets.

Thank you for listening!
Q & A