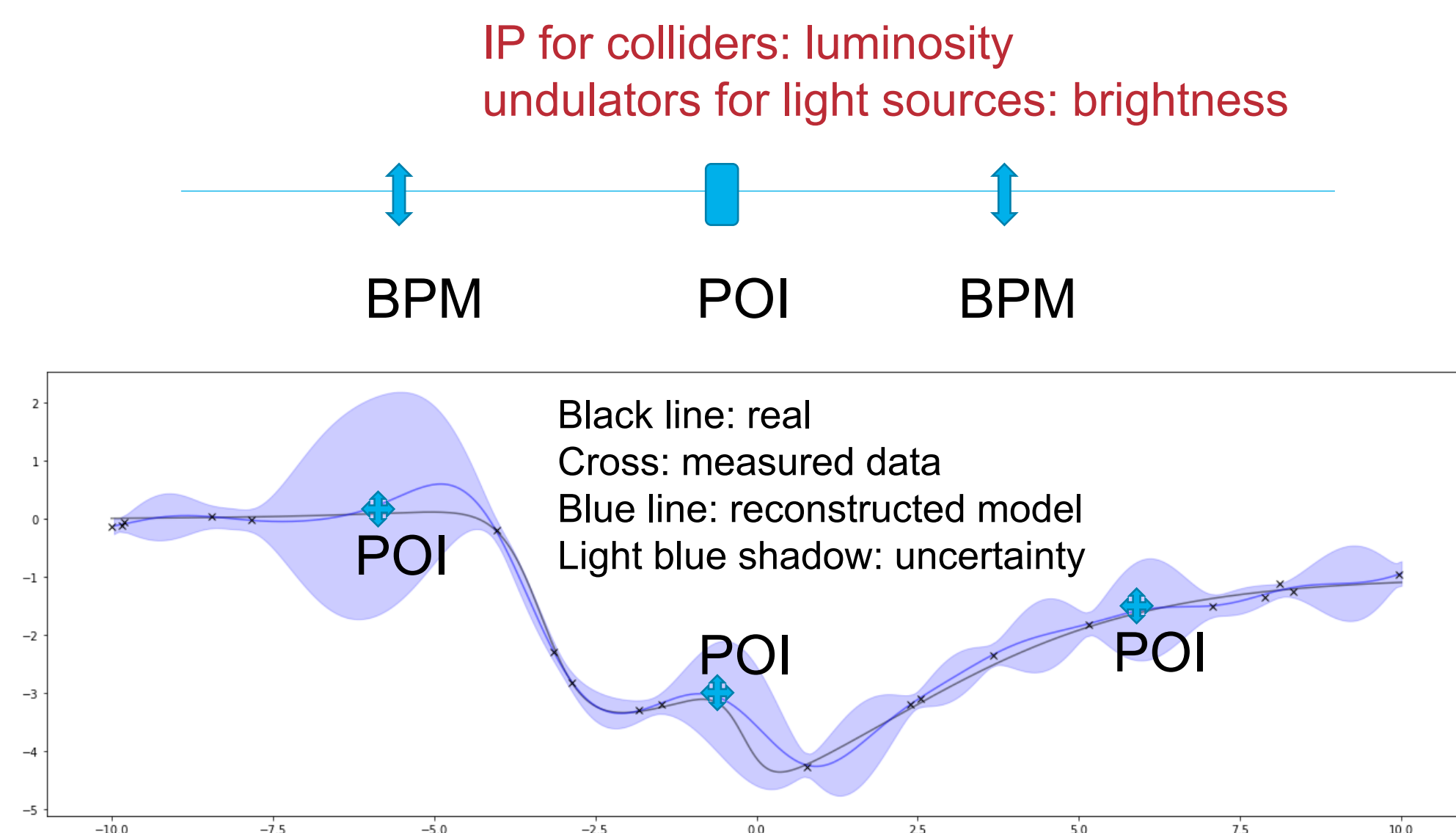


Analysis of BPM requirements with Bayesian Gaussian regression

Yongjun Li, Robert Rainer and Weixing Cheng | Brookhaven National Lab
Yue Hao | Michigan State University

NAPAC19:THXBA2

Problem



Methodology

A Bayesian Gaussian regression approach can determine the probability distribution of the predictive errors at **point of interest (POI)**, which can be used to conversely analyze the BPM system requirements

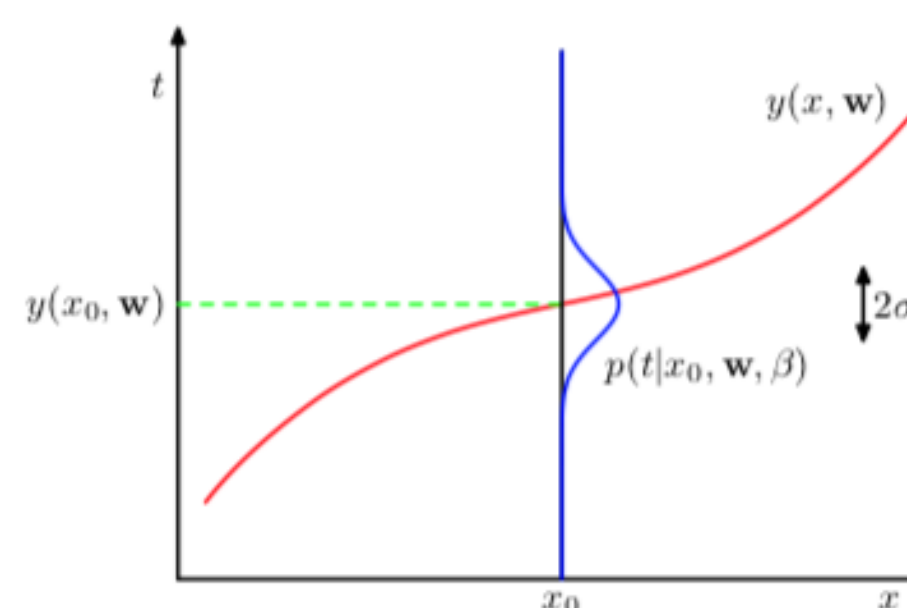
Bayes' Theorem

$$p(\Delta K|\beta) = \frac{p(\beta|\Delta K)p(\Delta K)}{p(\beta)}$$

$$\propto p(\beta|\Delta K)p(\Delta K).$$

Measured Optics

$$\mathcal{N}(\beta|\bar{\beta}, \sigma_{\beta}^2) = \frac{1}{\sqrt{2\pi}\sigma_{\beta}} \exp\left[-\frac{(\beta - \bar{\beta})^2}{2\sigma_{\beta}^2}\right].$$



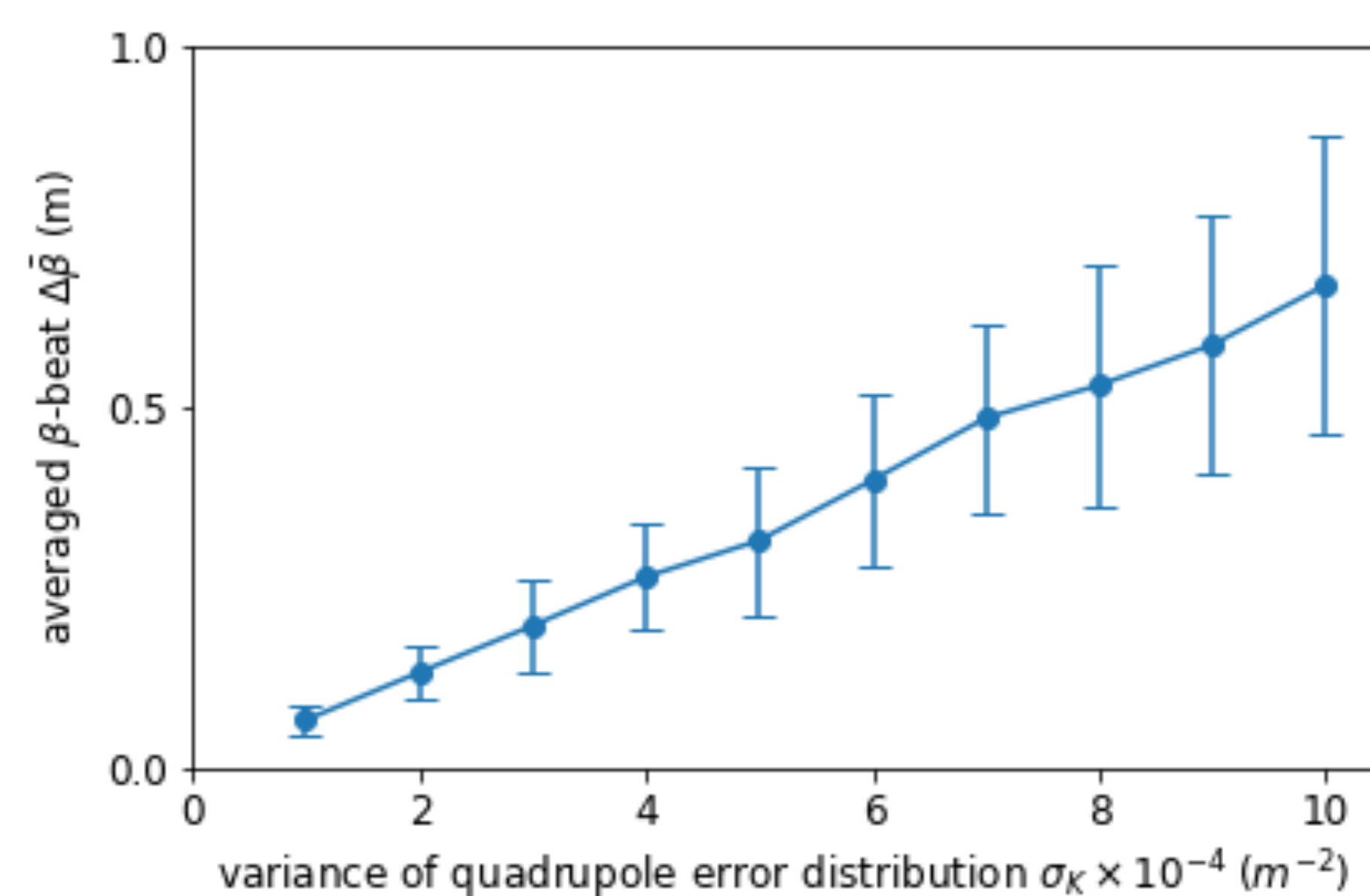
Quad's Errors

$$\mathcal{N}(\Delta K|0, \sigma_K^2) = \frac{1}{\sqrt{2\pi}\sigma_K} \exp\left[-\frac{\Delta K^2}{2\sigma_K^2}\right]$$

Works Cited

- Y. Li et al., arXiv:1904.05683 (2019)
- Y. Li et al., Phys. Rev. Accel. Beams **22** (2019)

Prior from Model



$$\sigma_{\Delta K} \sim \kappa \cdot \Delta\beta$$

Predictive errors

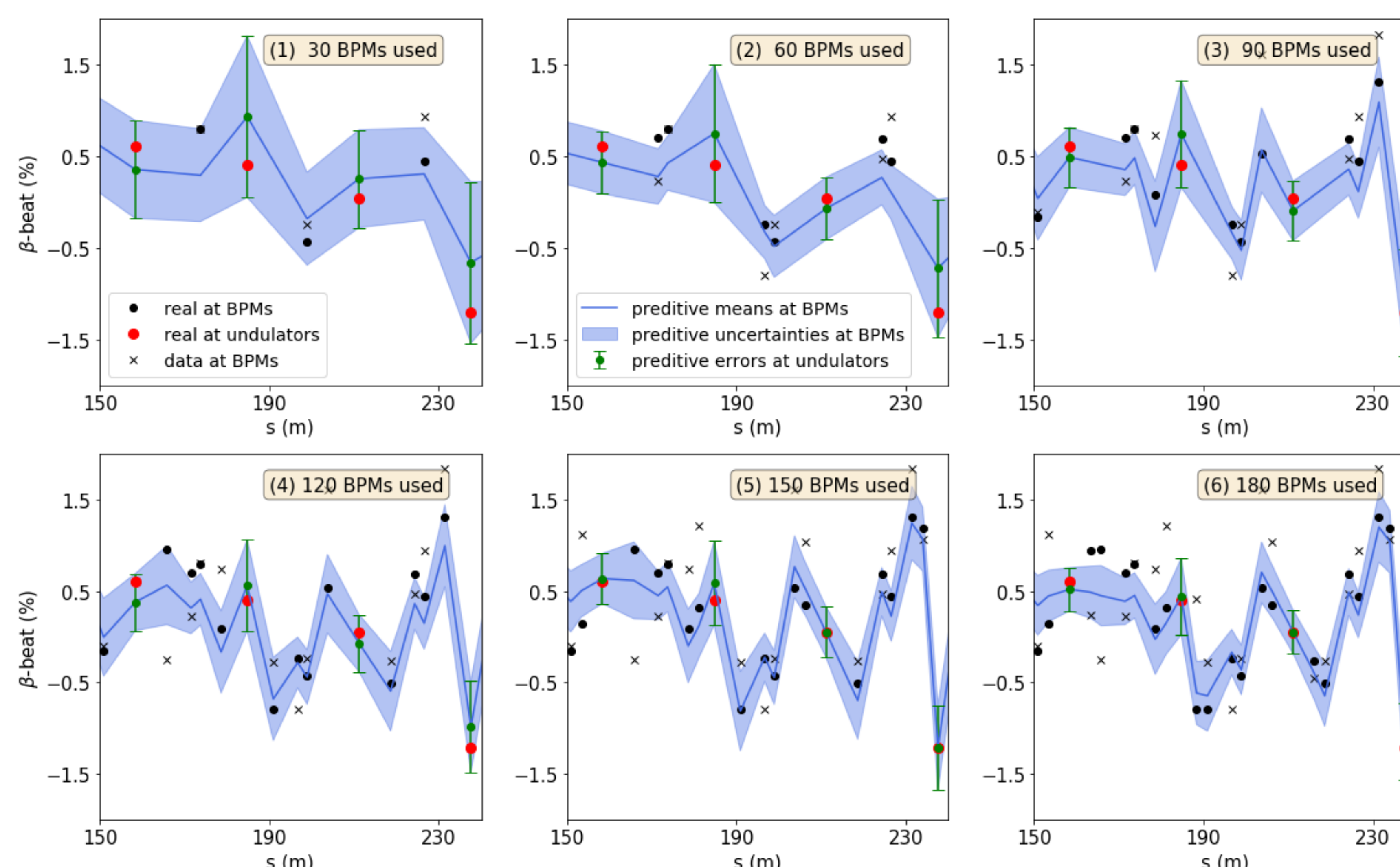
$$m_* = \sigma_{\beta}^{-2} M_* A^{-1} M^T \Delta\bar{\beta}$$

$$\Sigma_*^2 = M_* A^{-1} M_*^T,$$

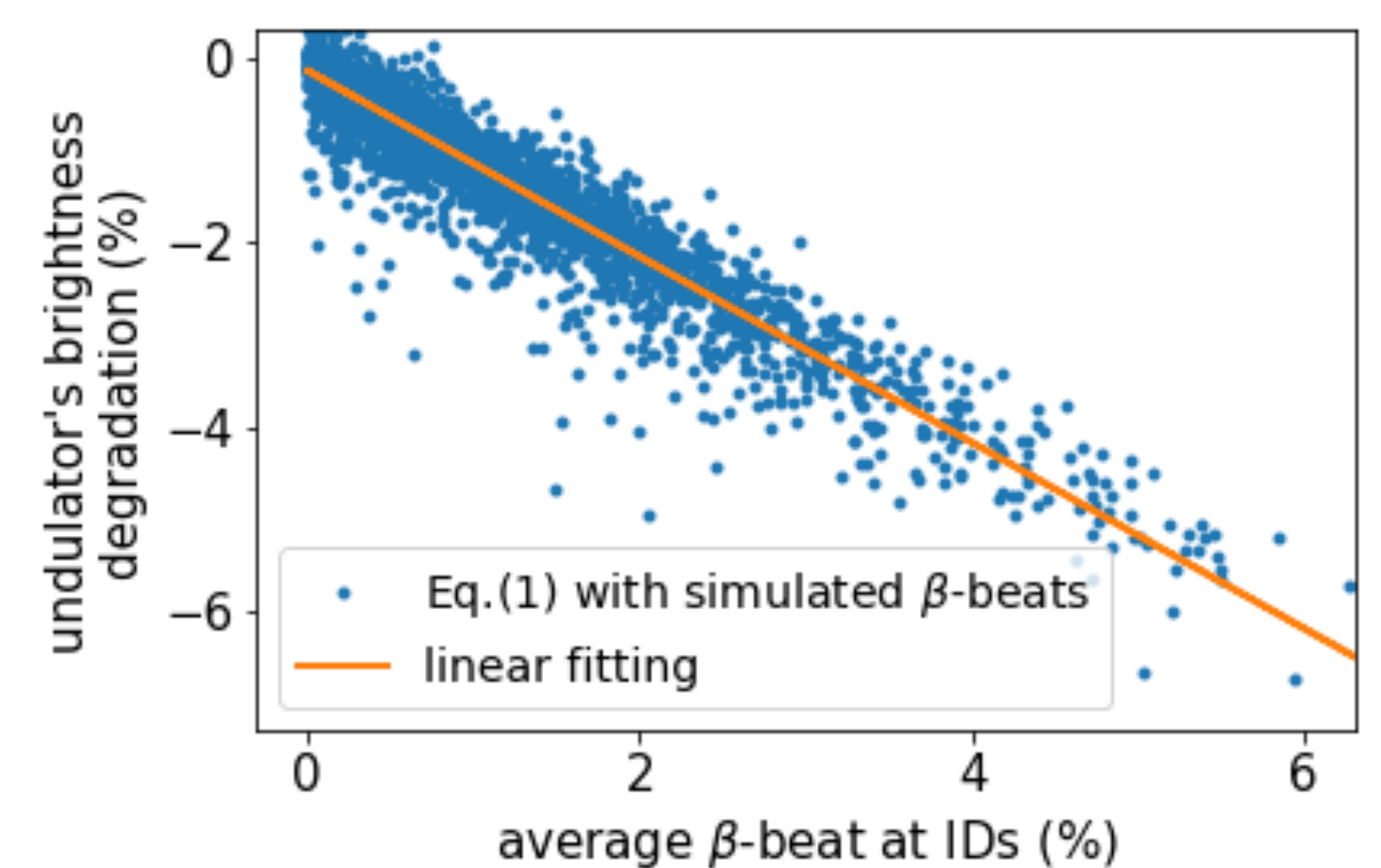
$$A = [\sigma_{\beta}^{-2} M^T M + \sigma_{\Delta K}^{-2} I]$$

- From BPM errors to model errors, then predictive errors at POIs
- Inverse problem: **How are the BPM system technical requirements determined in order to observe whether a ring achieves its desired performance or not.**

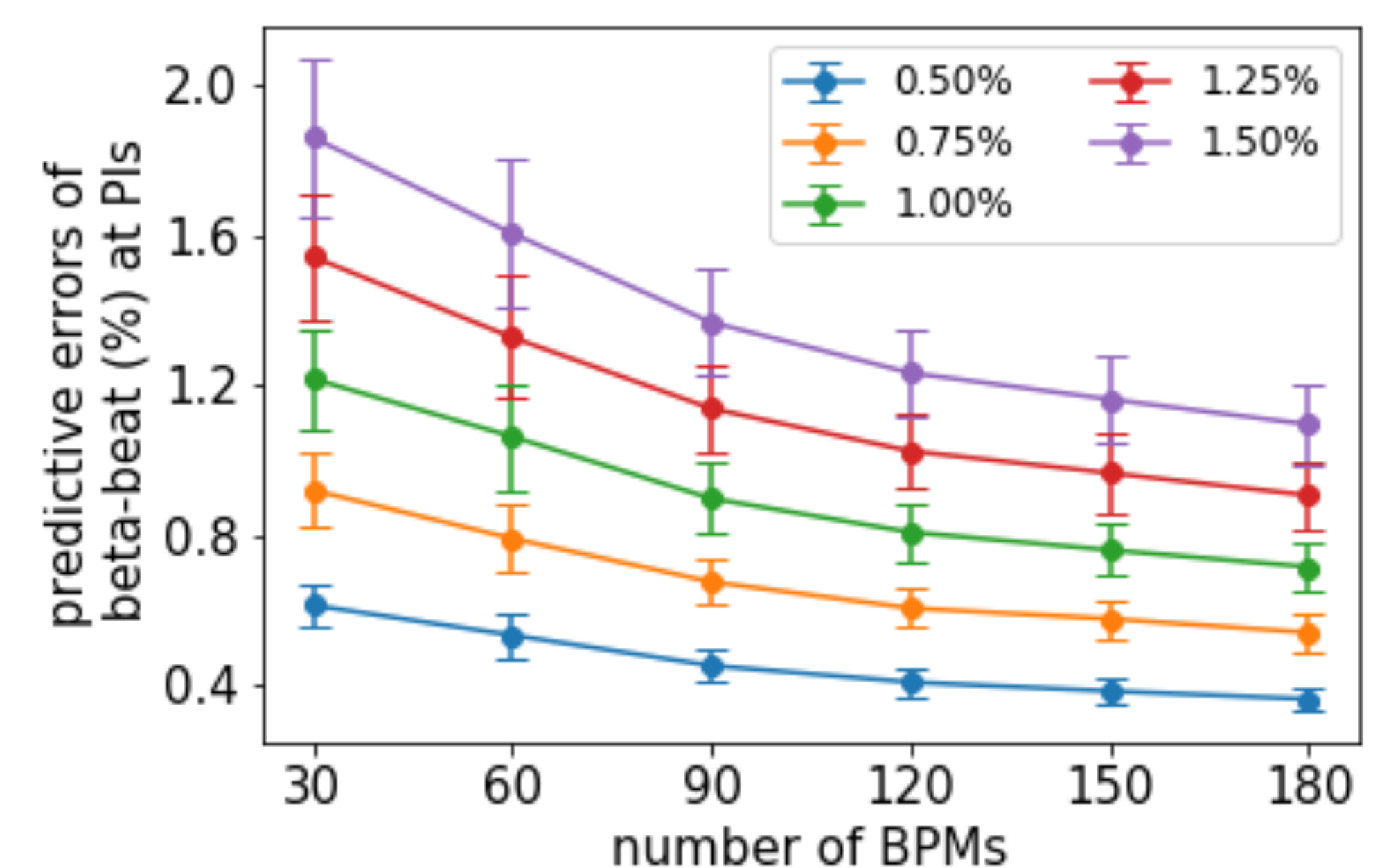
How many BPMs needed?



NSLS-II brightness



BPMs' resolution?



Summary:

- BPMs' quality is more important than their quantity, an optimal balance exists
- Specifications of BPM system can be quantitatively determined based on its ultimate performance

This research used resources of the NSLS-II, BNL under Contract No. DE-SC0012704. This work is also supported by NSF under Cooperative Agreement PHY-1102511, MSU