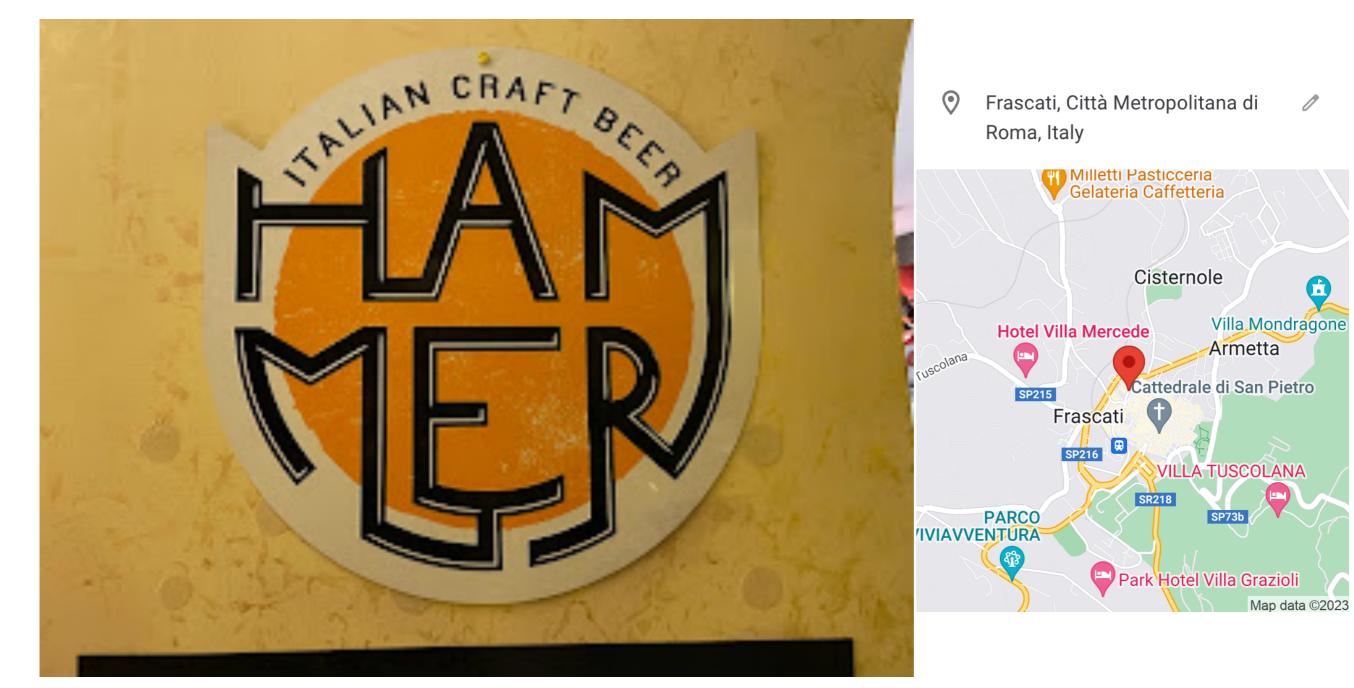
Fitting tools for semileptonics



Patrick Owen

Fitting in semileptonics

- A few features of fitting for semileptonics:
 - Don't tend to parameterise functions analytically.
 - PDF shapes depend on decay model.
 - Multi-dimensional fits common to extract most info.
 - Large yields.
 - Low purity strong dependence on background.
 - Simulation size leads to non-negligible systematics.

Fitting in semileptonics

• A few features of fitting for semileptonics:

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- Don't tend to parameterise functions analytically.
- PDF shapes depend on decay model. |Shape variations during minimisation
- Multi-dimensional fits common to extract most info. |1D projections insufficient for fit quality
 - Large yields. Efficient data management needed
- Low purity strong dependence on background.
- Simulation size leads to non-negligible systematics.

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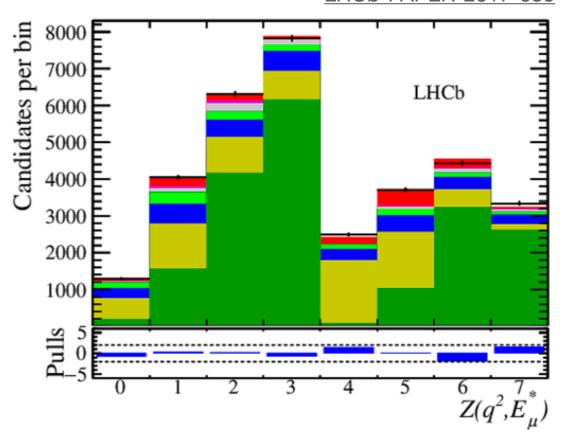
Template fitting

Control samples and ad-hoc variations

Barlow-Beeston or bootstrapping

RooFit implementation

- RooFit still most common tool.
 - Templates provided by RooHistPDF.
 - Limited to three explicit dimensions.
- Why histograms and not KDE?
 - Much faster and resolution wide c.f. bin width.
- RooHistFactory builds upon RooFit.
 - Provides Barlow-Beeston-lite implementation.
 - Also allows ad-hoc systematic variations.
 - Nice tutorial from Phoebe (old but gold) [link]



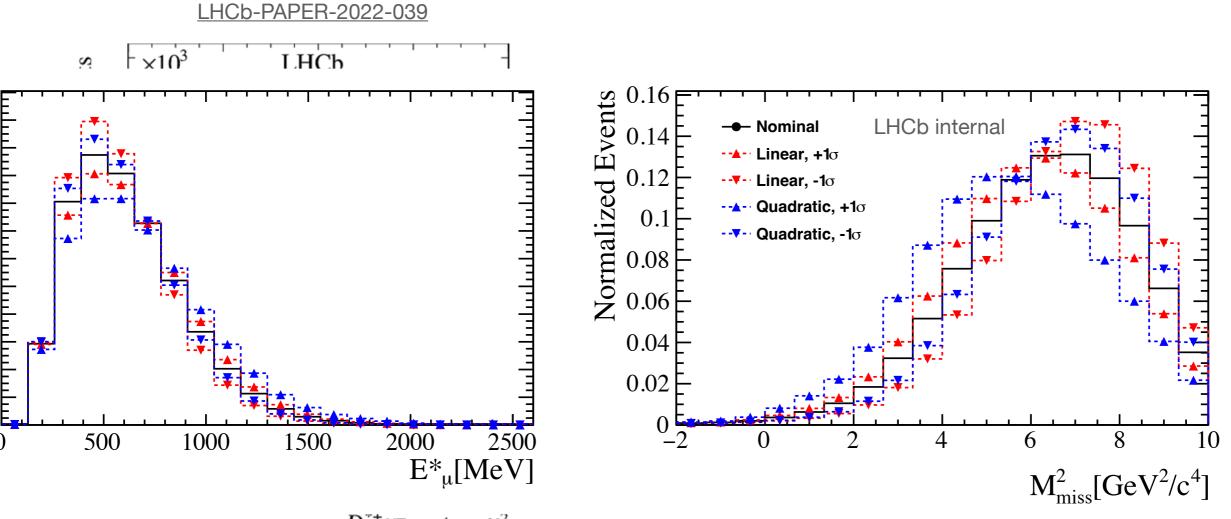
if(cfg->MCstat())

DstTauNu.ActivateStatError(); DplusTauNu.ActivateStatError(); DplusMuNu.ActivateStatError(); DstMuNu.ActivateStatError(); Bd2DD.ActivateStatError(); Bu2DD.ActivateStatError(); Bu2Dstst.ActivateStatError();

• If one only has one missing neutrino, more options open up.

Systematic shape variations

- Partial reconstruction imprints shape dependence.
- Decay model dealt with separately (see later), but what about detector effects such as data/MC agreement?
- HistFactory systematic variations allow for polynomial interpolation between 1σ variations.



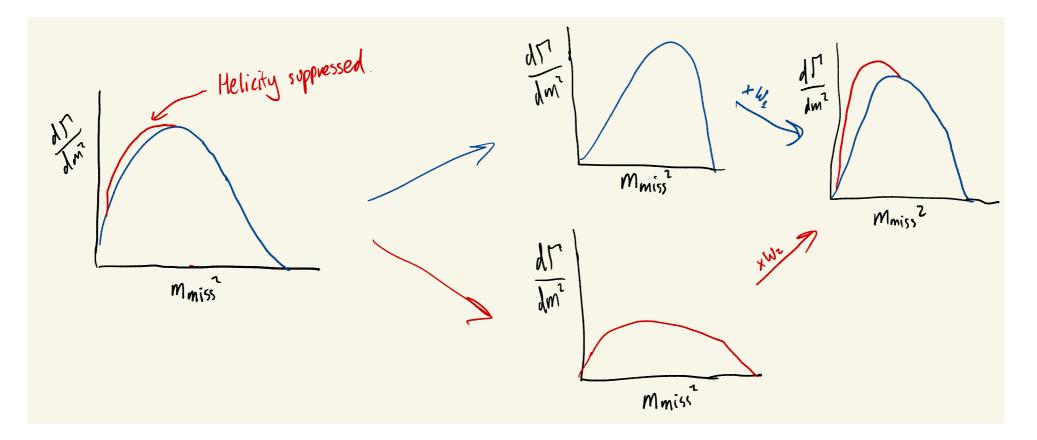
 $D^{*+}\mu^{-}$ vertex χ^2

Decay model

- What about the decay model? Could also use HistFactory variations.
 - This ignores correlations in the uncertainties.
 - Fine for systematics but what if the shapes are parameters of interest? e.g. form factors or Wilson Coefficients.
- Brute force solution event-by-event reweighing for every likelihood call.
 - Reweighting exact, no assumptions on correlations or interpolation.
 - Prohibitively slow, even for smaller size yields.
- One get out: If we measure form factors of muonic mode, uncertainties should be sub-leading and then one doesn't need to vary them in the fit.

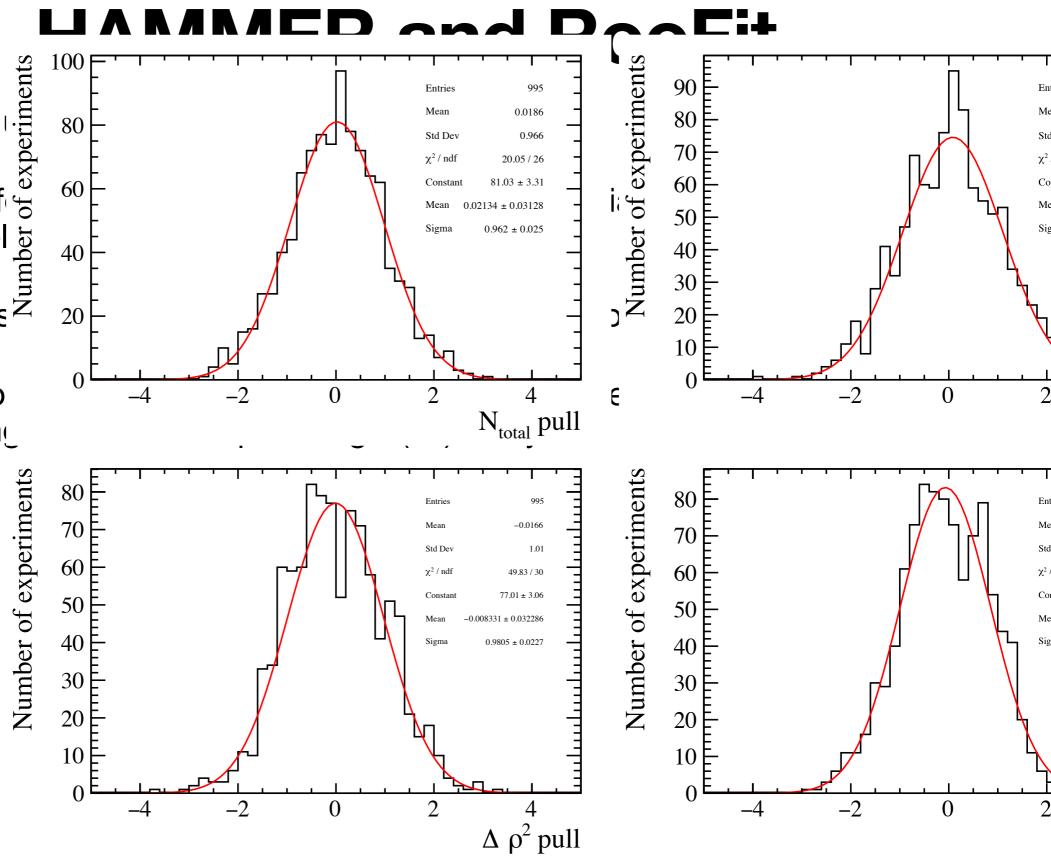
HAMMER

- Helicity Amplitude Module for Matrix Element Reweighting [Eur. Phys. J. C 80, 883 (2020)].
- Decompose amplitude into sub-parts, each with different linear dependence on parameters of interest.
 - $\frac{d\Gamma^{\{s\}}}{d\mathcal{PS}} = \sum_{\alpha,i,\beta,j} c_{\alpha} c_{\beta}^{\dagger} F_{i} F_{j}^{\dagger}(\{q\}) \mathcal{A}_{\alpha i}^{\{s\}} \mathcal{A}_{\beta j}^{\dagger\{s\}}(\{q\})$ $= \sum_{\alpha,i,\beta,j} c_{\alpha} c_{\beta}^{\dagger} F_{i} F_{j}^{\dagger}(\{q\}) \mathcal{W}_{\alpha i\beta j},$
- Oversimplified example: scalar part of a histogram.



Best place to find documentation is on the webpage: https://hammer.physics.lbl.gov/

- HAMMER being HAMMER being Sometimes just f debugged model and the some analyses analyses of the source of the sourc
- - **Developed Ro** Currently being



Overall works well and gives good pulls in toys. Increases likelihood call time by factor ~100, worth digging into once this round is over.

Simulation uncertainties

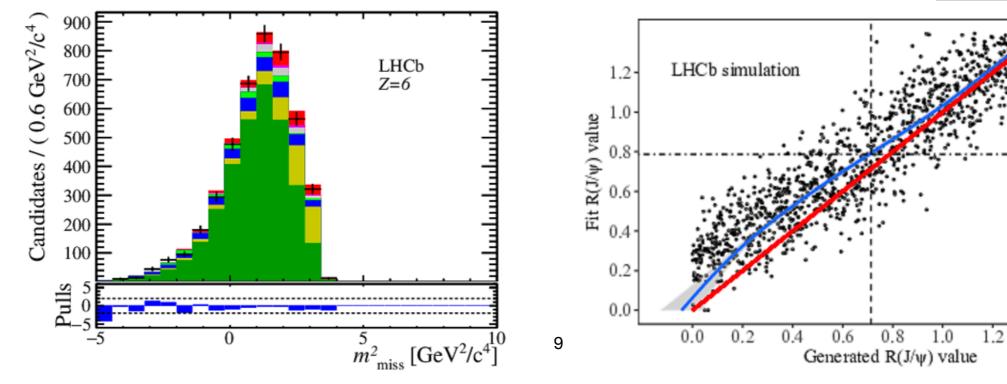
- Despite our best efforts, simulation size still seems creep up as systematic uncertainty.
 - Lose a large fraction of effective size from the various re-weighting schemes.
- Barlow-Beeston-lite can undercover this uncertainty.
 - Necessary to check with boostraps.
 - Is Barlow-Beeston worth it? Fits faster and more stable without it.
 - As long as have enough simulation to explore disagreement, perhaps not needed.



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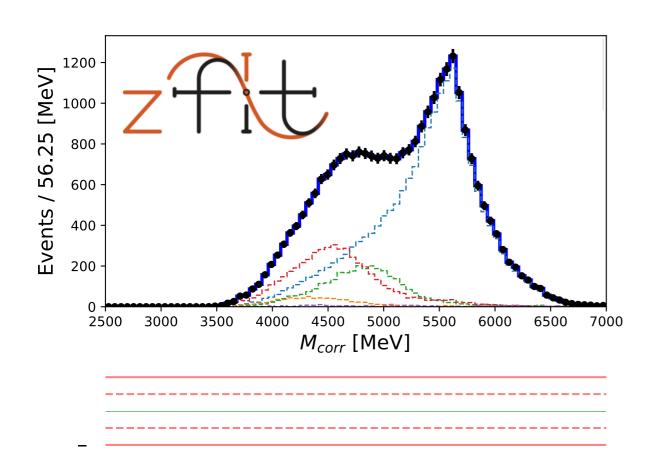
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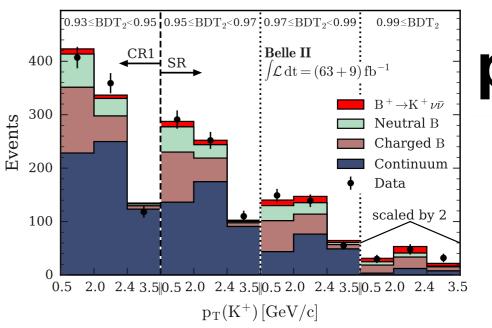
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Beyond ROOT

- ROOT starting to compete with large ecosystem of analysis in python.
- Original plan for $R(\Lambda_c)$ was to implement it in zFit [SoftwareX 11 (2020) 100508].
 - How to efficiently interface HAMMER here would be a big undertaking.
 - Pyhf also an option, as the python replacement for HistFactory [JOSS, 6(58), 2823]
- HistFactory works, do we stick with it, try to improve timing, or move to python.
 - Or both?





<u>Phys. Rev. Lett. 127, 181802 (2021)</u>



Summary Slide:

Title: Fitting Tools for Semileptonics - Key Points

- 1. Fitting in semileptonics:
 - Multi-dimensional fits
 - Large yields, low purity
 - Simulation size systematics
- 2. Template fitting with RooFit:
 - Efficient data management
 - 1D projections insufficient
 - Barlow-Beeston or bootstrapping
- 3. HAMMER & RooFit integration:
 - Increasing use at LHCb
 - RooHammerModel for interfacing
 - Slower likelihood calls
- 4. Addressing simulation uncertainties:
 - Barlow-Beeston-lite & bootstrap methods
 - Empty bins in low stats fits
- 5. Future directions:
 - HistFactory vs. Python alternatives
 - Efficient HAMMER integration in Python

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- 1. How do you foresee improvements in simulation techniques impacting the fitting process \bigcirc \bigcirc for semileptonic decays in the near future?
- 2. Given the increasing popularity of Python in data analysis, do you think the community should prioritize transitioning from RooFit to Python-based alternatives like zFit or pyhf?
- 3. Are there any specific limitations of the Barlow-Beeston method that could be addressed to make it more reliable in handling simulation uncertainties?
- 4. How can researchers optimize the balance between computing resources and the accuracy of the fitting process in semileptonic decays?
- 5. In the context of HAMMER integration with RooFit, are there any plans to improve the likelihood call time or implement additional optimizations to reduce the computational burden?
- 6. Can you discuss any recent advances or ongoing research in semileptonic decay fitting that may help address the challenges outlined in your presentation?
- 7. How do you envision the role of machine learning and artificial intelligence in improving the fitting process for semileptonic decays in the coming years?