



Inclusive B -meson tagging for an $R(D^*)$ measurement at Belle II

IMPRS Interview

Sofia Palacios Schweitzer
LMU München

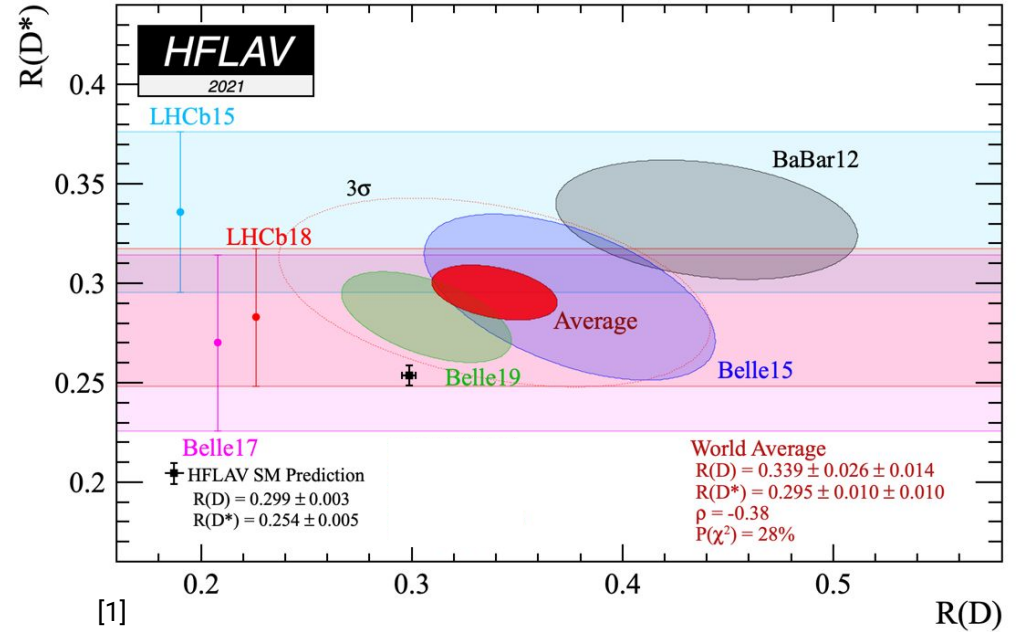
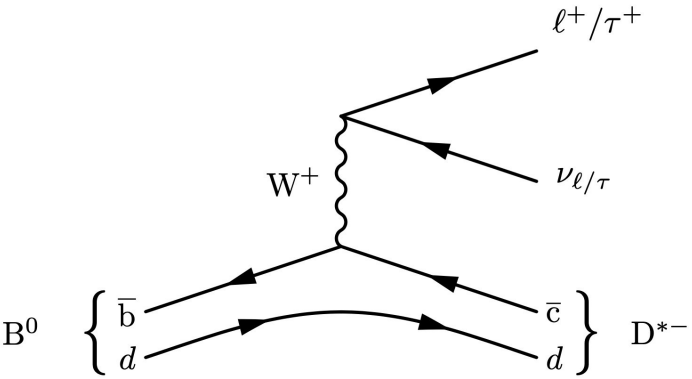
18. July 2022





A Road to New Physics

$$R(D^*) = \frac{\mathcal{B}(B \rightarrow D^* \tau \nu_\tau)}{\mathcal{B}(B \rightarrow D^* l \nu_l)} \quad l = \mu, e$$



→ Measured values in 3.3 σ tension with SM prediction

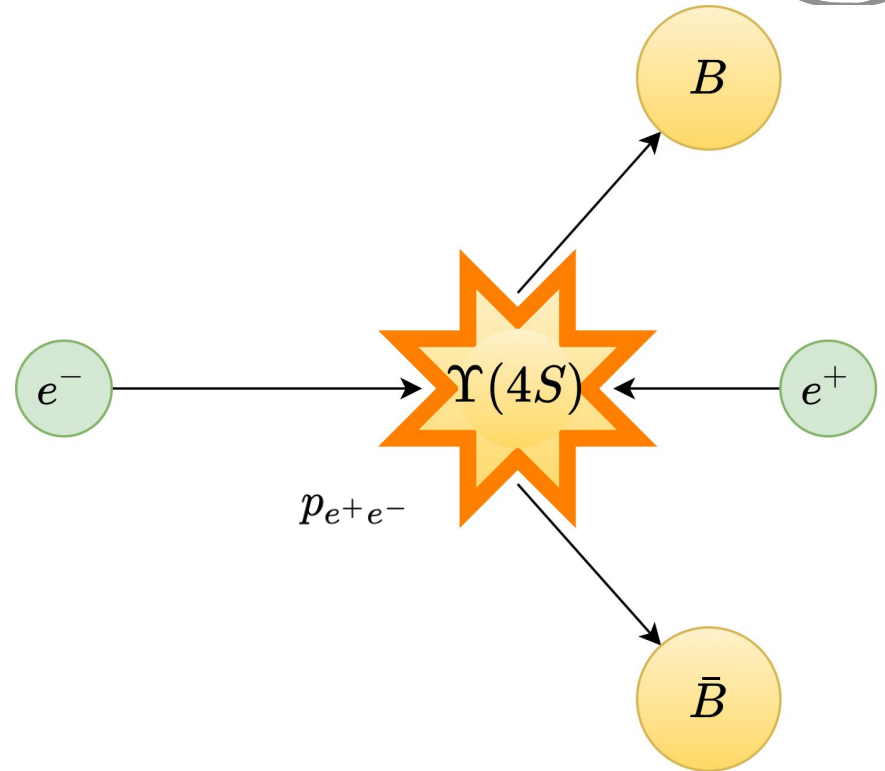


How to get a B-Meson

In the Belle II Experiment at SuperKEKB :

1. Collision of electron & positron ($\sqrt{s}=10.58$ GeV)
(precisely known)
2. Production of $\Upsilon(4S)$
3. Decay into B-Meson pair

→ in 8 years : $5 \cdot 10^{10}$ B-Meson pairs

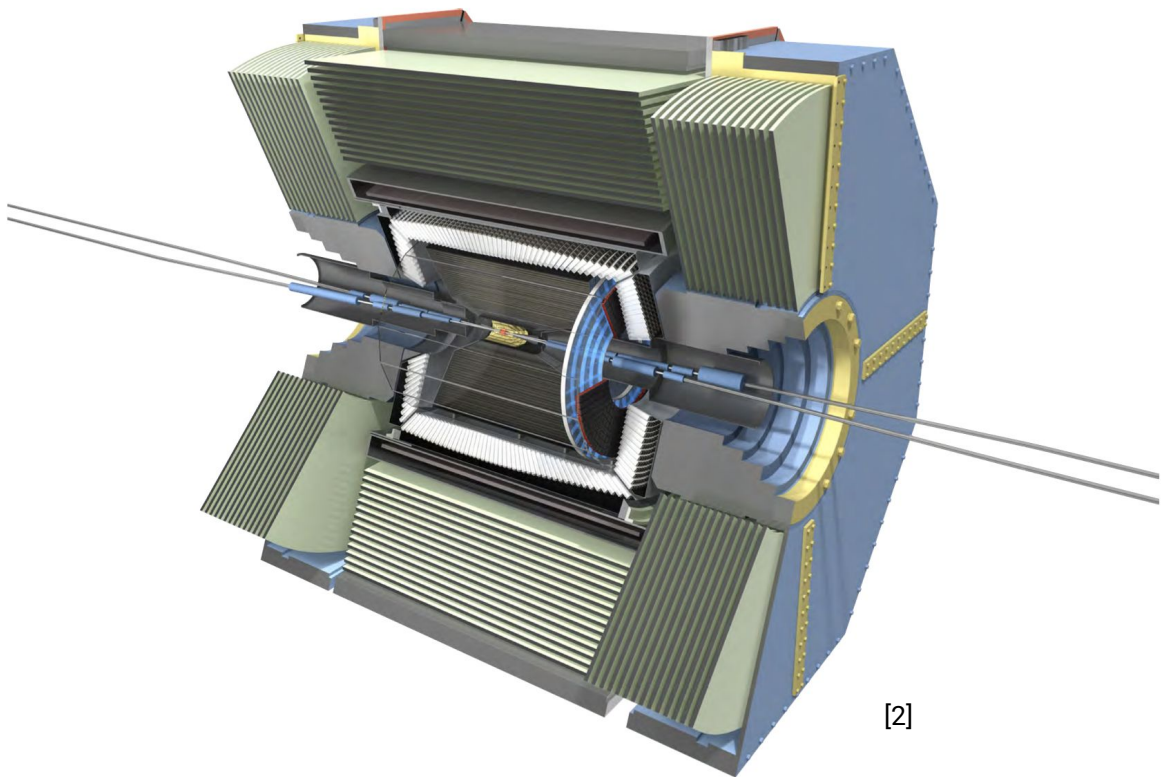


How to detect

Belle II Detector

1. Vertex detector
2. Central Drift Chamber
3. Particle identification detectors
4. EM Calorimeter inside solenoid
5. K_L & μ detector

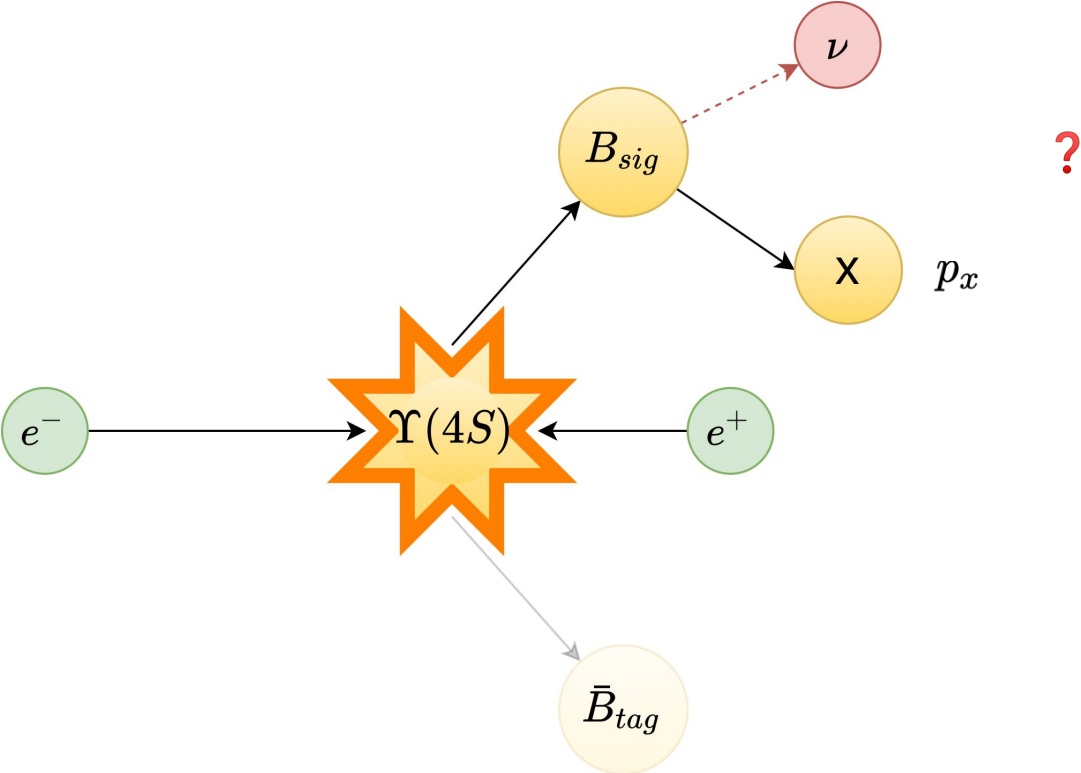
→ Neutrinos?



[2]

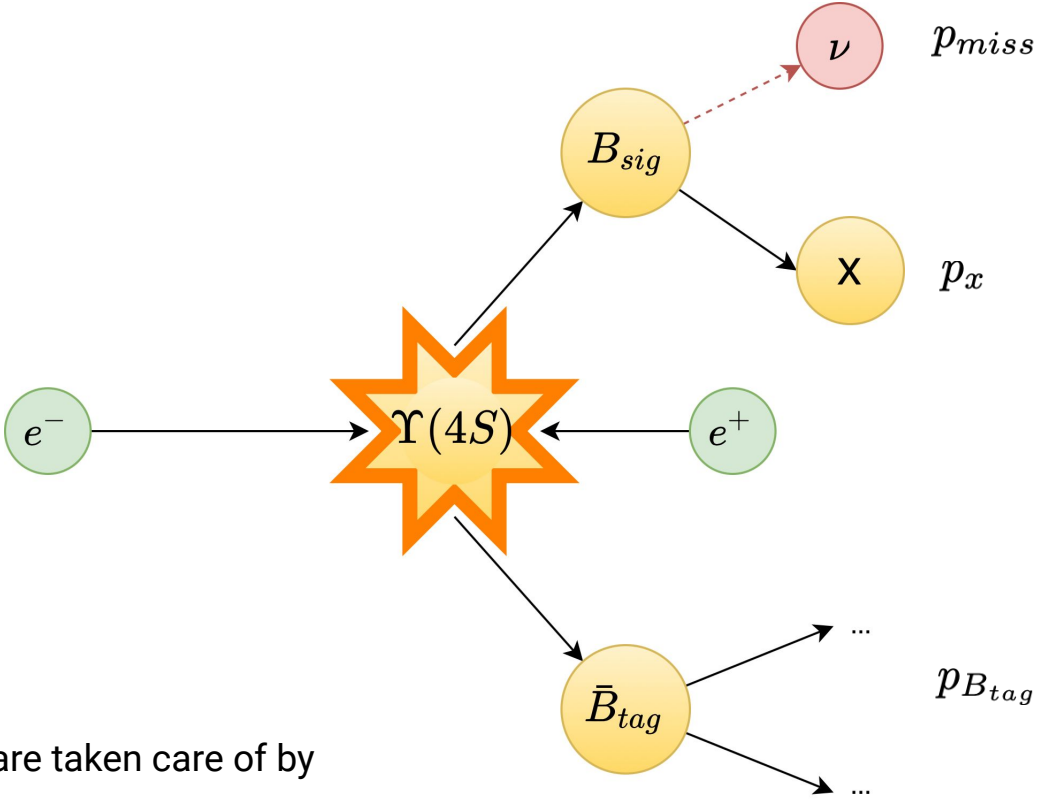


What about Neutrinos?





Let's just tag it

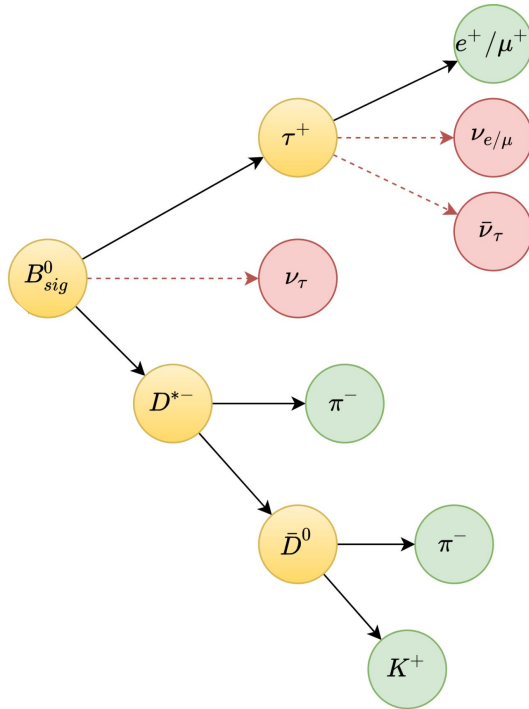


→ Neutrinos are taken care of by

$$p_{miss} = p_{e^+e^-} - p_{B_{tag}} - p_x$$



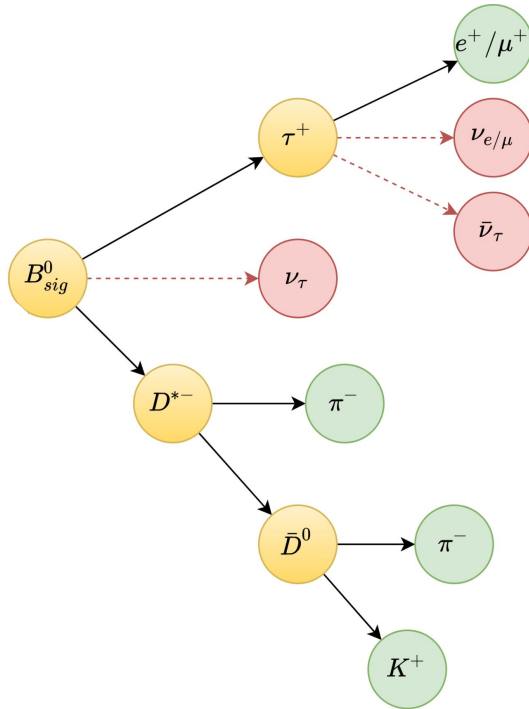
Reconstruction Strategy



1. Reconstruct signal side exclusively



Reconstruction Strategy



For tag side reconstruction:

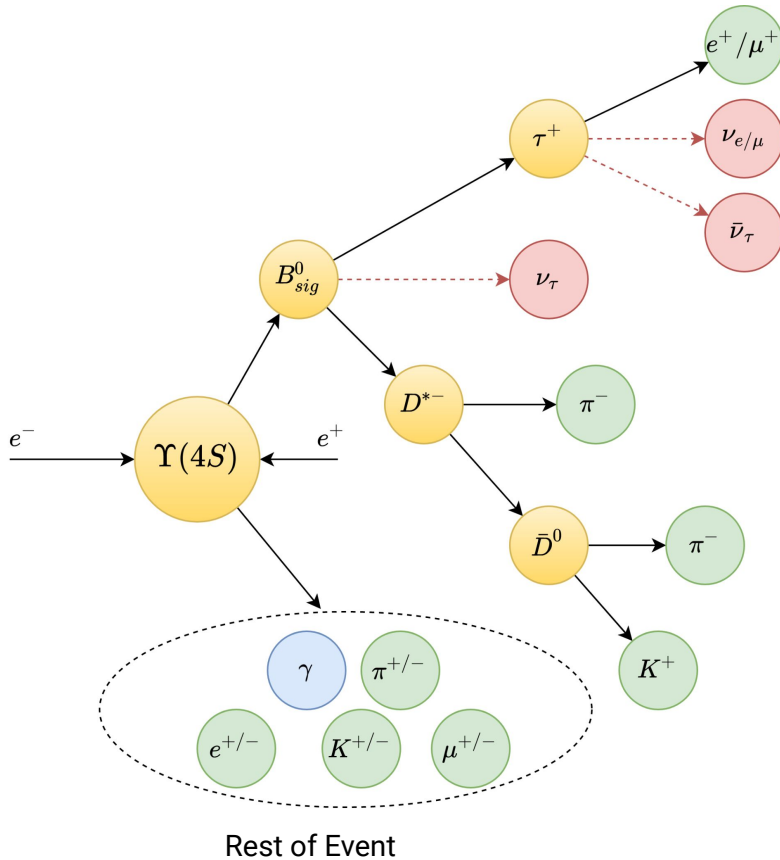
- iterating over multiple exclusive decay modes (default)

vs

- inclusively



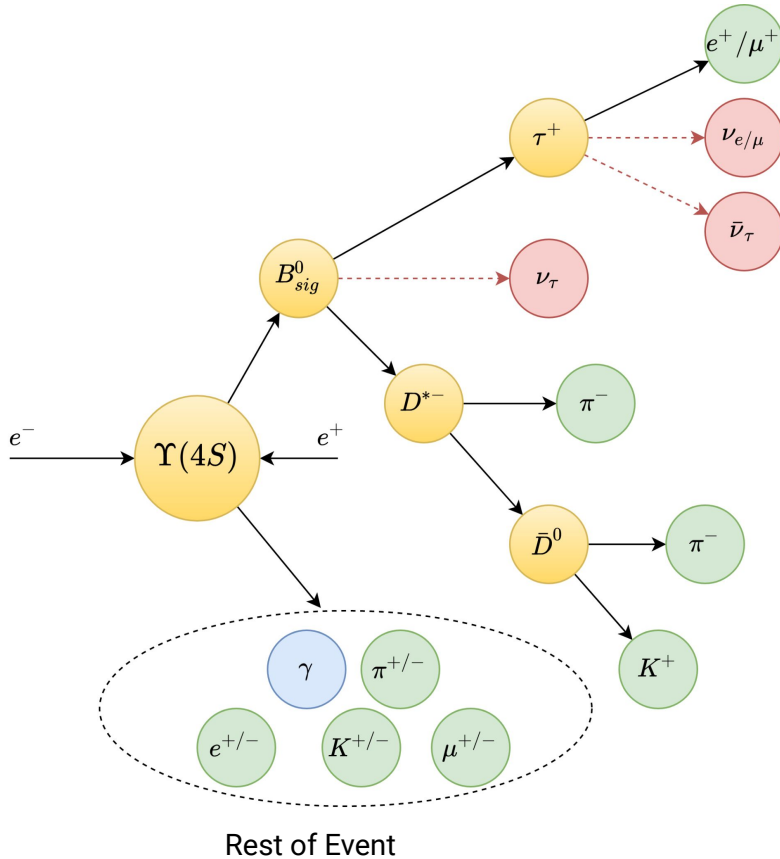
Reconstruction Strategy



2. Reconstruct tag side inclusively (hadronically)
- no intermediate daughters are reconstructed



Reconstruction Strategy



Belle II exclusive tagging algorithm

Tags	Decay Modes covered	Tag Side Efficiency
Semileptonic	4.0%	2.04%
Hadronic	1.1%	0.46%

[3],[4]

→ fully inclusive promises higher tag side efficiency

BUT: Very high background levels



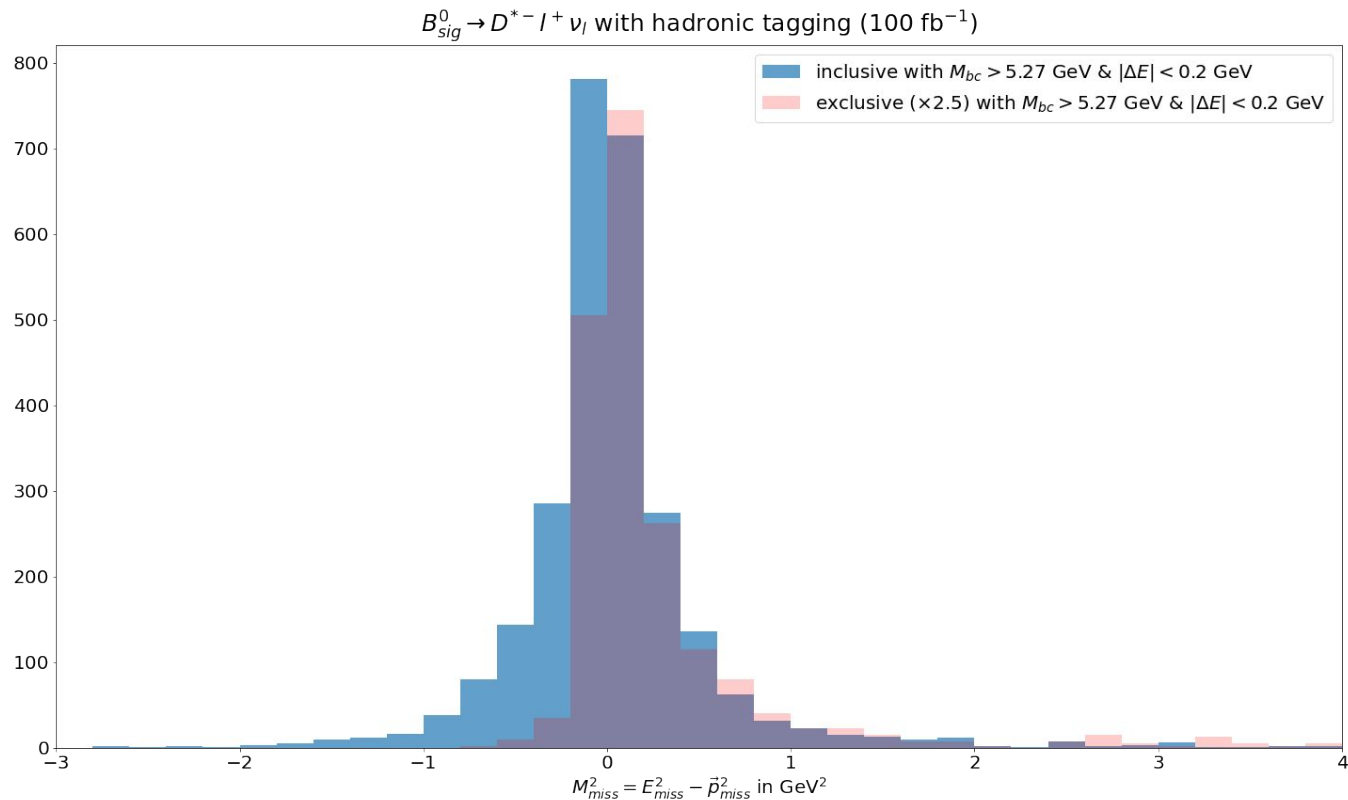
1. Reconstruction
 - a. optimize signal & tag side selection

2. Background Analysis
 - a. categorize background modes
 - b. build MVAs

3. Fit
 - a. 2D Fit in M_{miss}^2 and lepton momentum

4. Estimate Systematic Uncertainties

1. Reconstruction



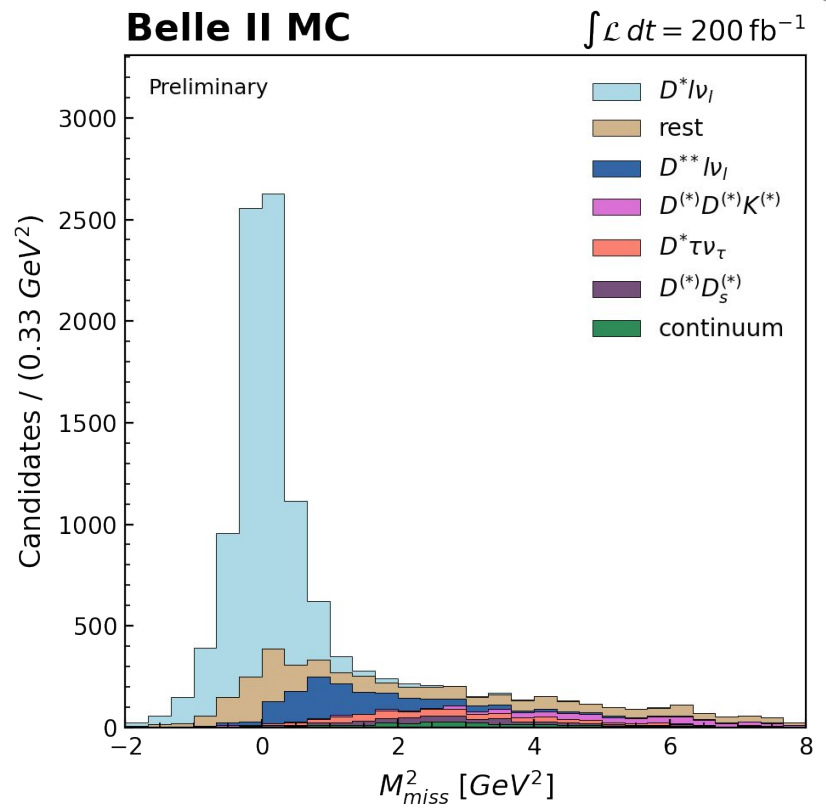
→ Reconstruction strategy works



2. Background Analysis

Reconstruction Modes:

- 62% normalization mode
- 3.5% signal mode
- 2.6% continuum background
- 32% combinatorial background





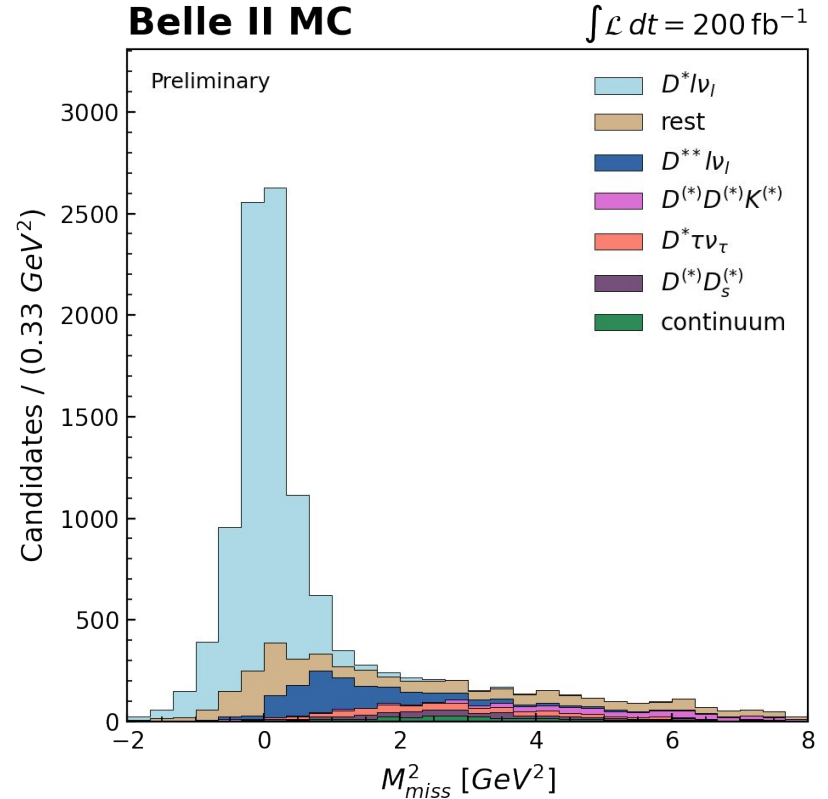
2. Background Analysis

For $M_{\text{miss}}^2 < 1 \text{ GeV}^2$:

Normalization Channel with purity of 82%

For $M_{\text{miss}}^2 > 1 \text{ GeV}^2$:

Signal Channel with purity of 12%



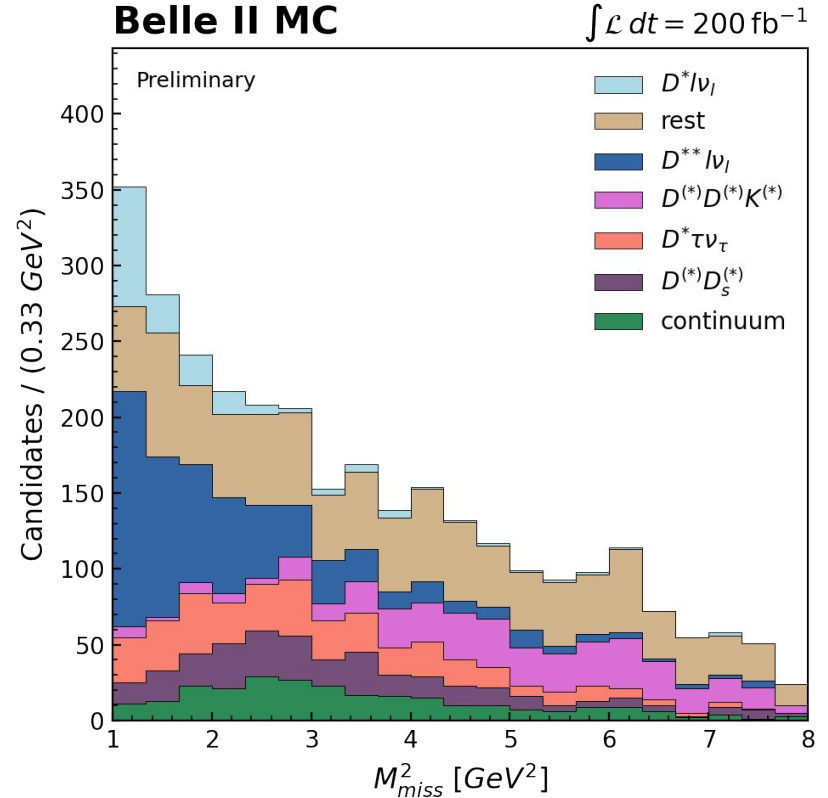


2. Background Analysis

Sideband Components

Background:

- 20% $D^{**} l \nu$ (higher excited D states)
 $\hookrightarrow D^* \pi$
- 13% $D^{(*)} D^{(*)} K^{(*)}$
- 10% $D_s^{(*)} D^{(*)}$
 $\hookrightarrow \tau \nu$
- 8.6% continuum
- 38% rest

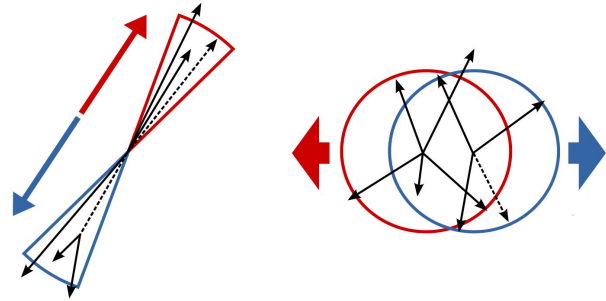


2. Background Analysis

Suppression

Idea: Build MVAs to suppress different background components

1. Continuum Background: Event Shape Variables
2. Combinatorial Background: ?



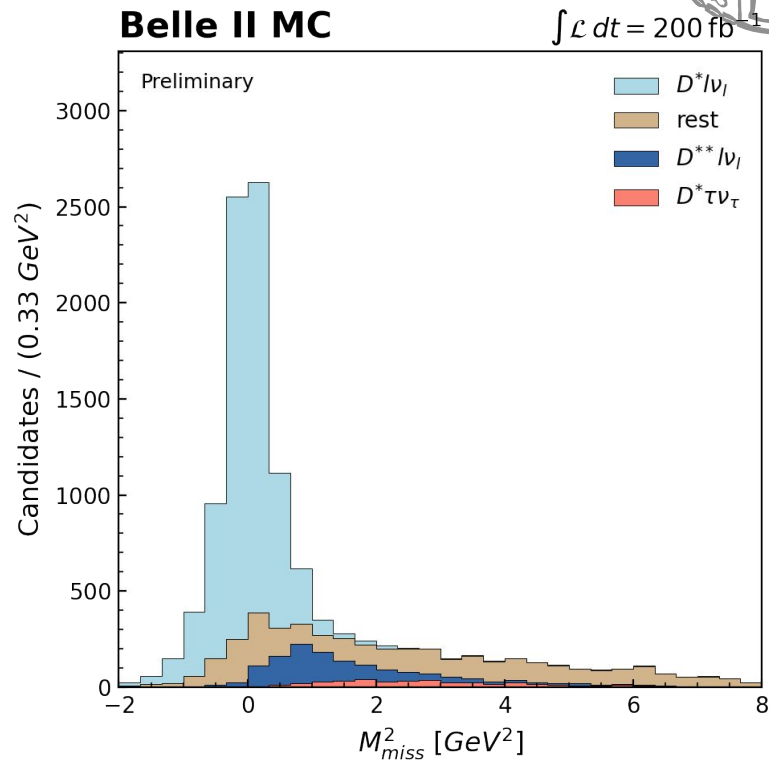
→ Suppression not optimal yet



3. Fit

- unbinned maximum likelihood fit
- pdfs: kernel density estimation from MC
- 2D: M_{miss}^2 & lepton momentum

- Fit Components:
1. normalization mode
 2. signal mode
 3. D^{**} background
 4. remaining background

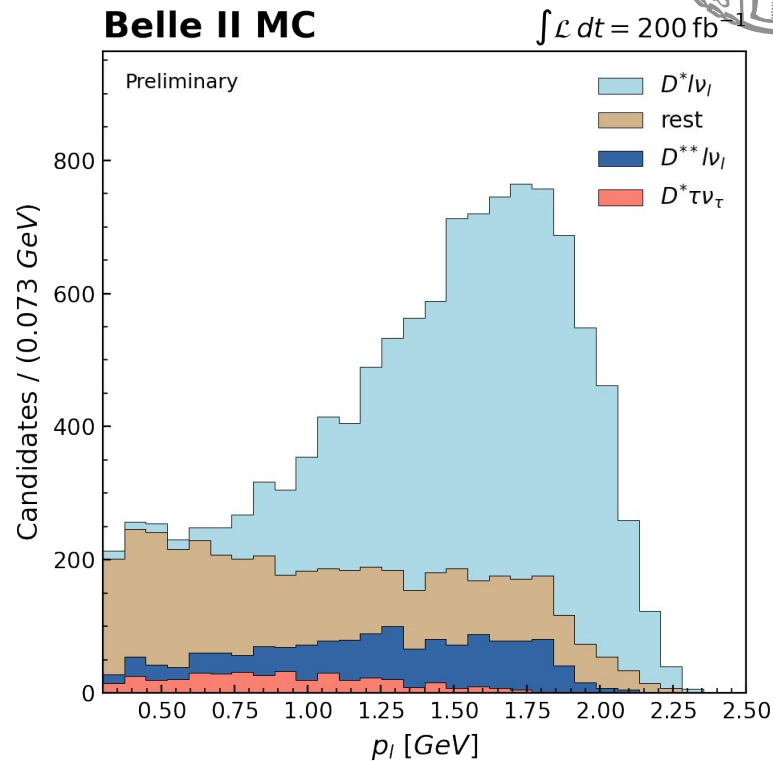




3. Fit

- unbinned maximum likelihood fit
- pdfs: kernel density estimation from MC
- 2D: M_{miss}^2 & lepton momentum

- Fit Components:
1. normalization mode
 2. signal mode
 3. D^{**} background
 4. remaining background





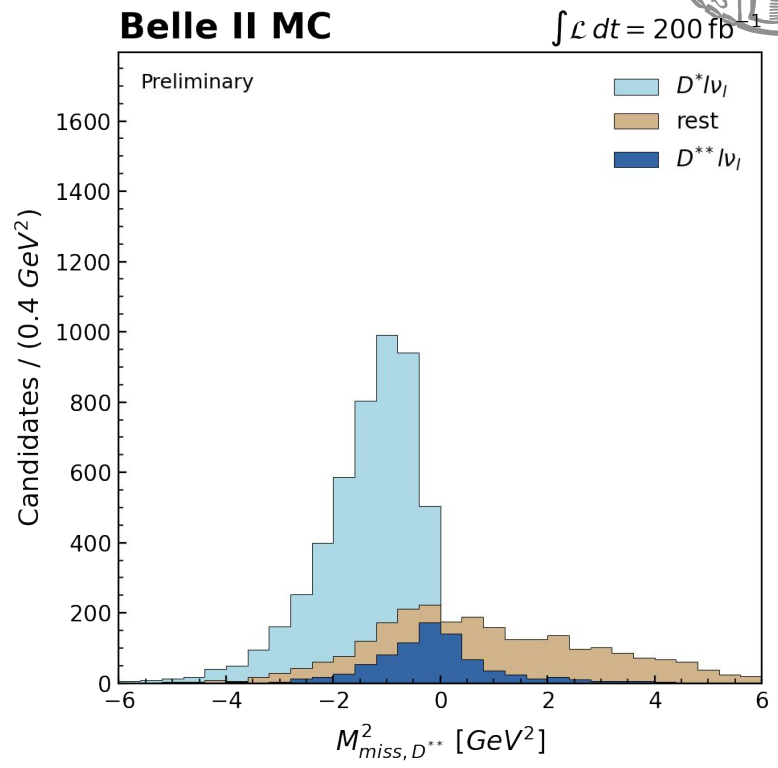
3. Fit

Investigate way to fix D** Background:

- Reconstruct D** in $B \rightarrow D^* \pi | \nu$
- Fit of M_{miss}^2 of D**
- Reconstruction efficiency

→ Use fit result to constrain D** component in signal fit

→ Sequential fit: 1d + 2d





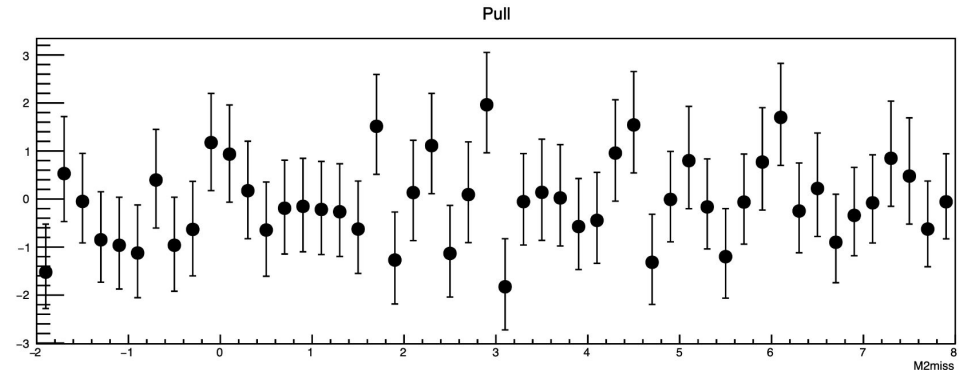
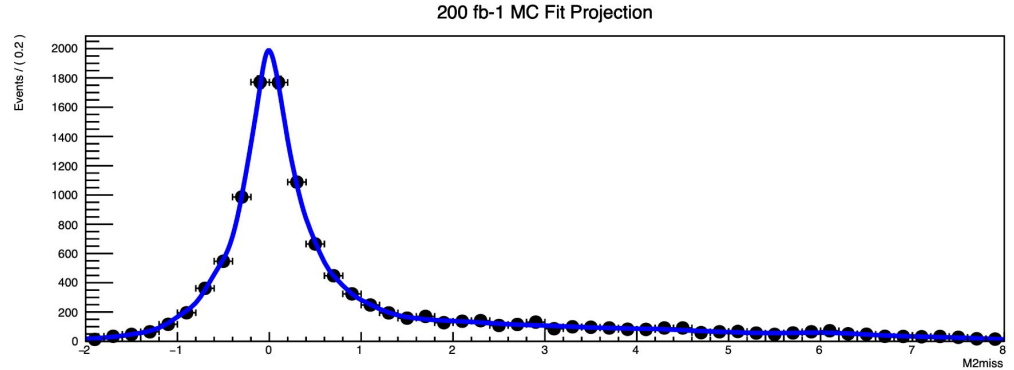
3. Fit

Test fit to MC sample:

$$R(D^*) = \frac{n_{sig}}{n_{norm}} \frac{\epsilon_{norm}}{\epsilon_{sig}} = 0.28 \pm 0.03$$

in MC: $R(D^*) = 0.258$

- toy test show unbiased fit with well described uncertainties





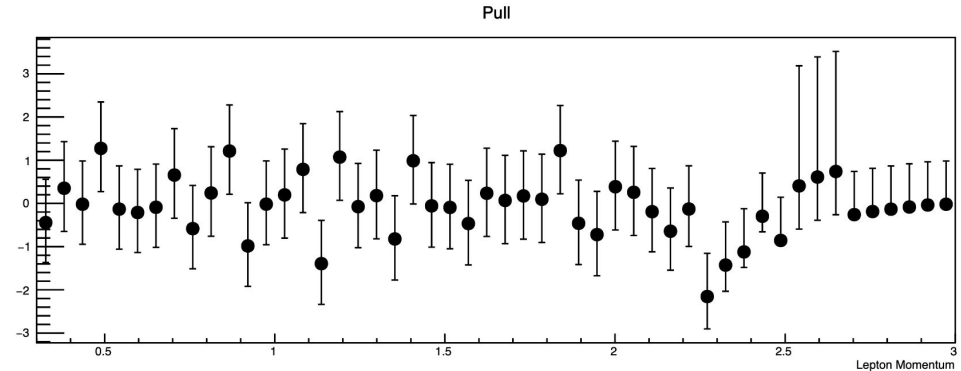
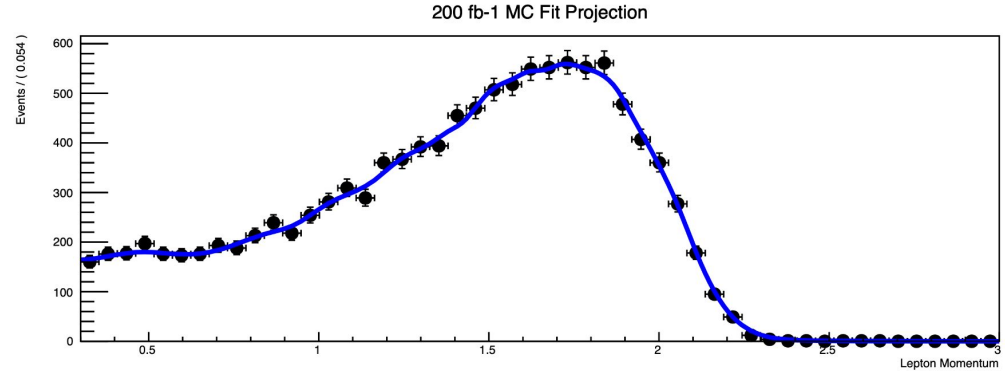
3. Fit

Test fit to MC sample:

$$R(D^*) = \frac{n_{sig}}{n_{norm}} \frac{\epsilon_{norm}}{\epsilon_{sig}} = 0.28 \pm 0.03$$

in MC: $R(D^*) = 0.258$

- toy test show unbiased fit with well described uncertainties



What now?



Is this approach compatible with exclusive tagging?

For now **yes**, available data statistically limited (currently $\sim 200 \text{ fb}^{-1}$)

BUT: statistical limit will be overcome eventually

To stay compatible:

- Optimize training
- Estimate systematic uncertainties



[5]



- [1] https://hflav-eos.web.cern.ch/hflav-eos/semi/spring21/r_dtaunu/rdrds_2021.png
- [2] <https://www.belle2.org/e21595/e21770/infoboxContent25428/BelleII3D.pdf>
- [3] T.Keck, “The Full Event Interpretation for Belle II” (2014), KIT Karlsruhe
- [4] T.Keck et al. “The Full Event Interpretation” (2019), [arXiv:1807.08680](https://arxiv.org/abs/1807.08680)
- [5] <https://stock.adobe.com/de/contributor/205527442/stichfiguren-de>