

Analysis presentation V_{cb}

Part 1: Overview and tag side

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March 16, 2021



Overview

- Part 1 (Kilian): Why V_{cb} ? How V_{cb} ? Measurements at Belle.
- Part 2 (Kilian): **Tagging calibration** for $|V_{cb}|$ measurement with $B \rightarrow D^* l \bar{\nu}_l$ (but the calibration looks at $B \rightarrow X l \nu$)
- Part 3 (Markus): $|V_{cb}|$ measurement with $B \rightarrow D^* l \bar{\nu}_l$

Caveats:

- Not a physics talk: We'll **cut corners** and **go on tangents** about doing analysis
- I'm a bit out of the loop with recent $|V_{cb}|$ papers. . .
- Not a rockstar presentation :/

Part 1

1 Why?

2 How?

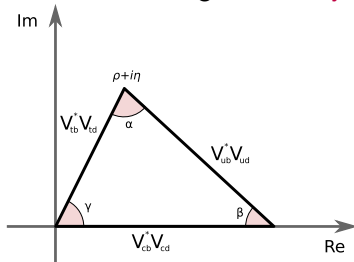
- Measuring $|V_{cb}|$
- Exclusive vs Inclusive Analyses
- Tagging

Why $|V_{cb}|$?

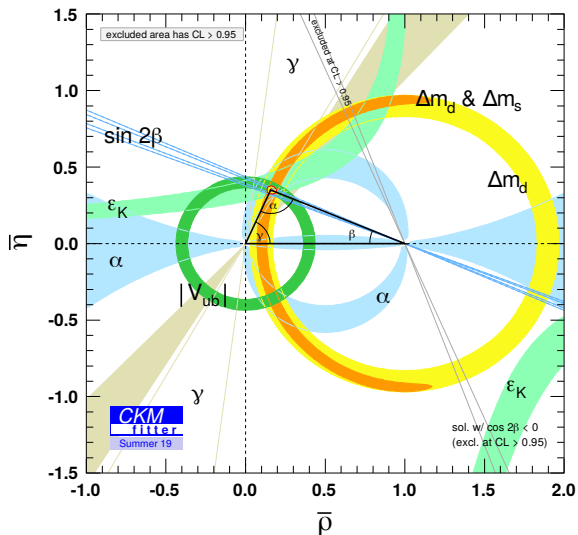
- The CKM matrix $V = \begin{pmatrix} V_{ud} & V_{us} & V_{ub} \\ V_{cd} & V_{cb} & V_{cs} \\ V_{td} & V_{ts} & V_{tb} \end{pmatrix}$ is **unitary** in the SM, i.e.

$$VV^\dagger = 1 = V^\dagger V$$

- Thus $\sum_k |V_{ik}|^2 = 1$ for all k (weak unitarity) and
- $\sum_k V_{ik} V_{jk}^* = 0$ for all i, j
- Most popular: $V_{ub}^* V_{ud} + V_{cb}^* V_{cd} + V_{tb}^* V_{td} = 0$
- Powerful test** of the SM
- Can also be visualized as a triangle **“unitarity triangle”**



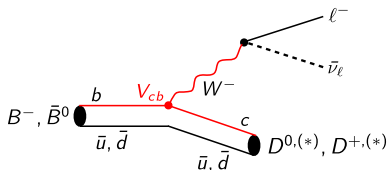
[wiki]

Why $|V_{cb}|$ 

Measuring $|V_{cb}|$

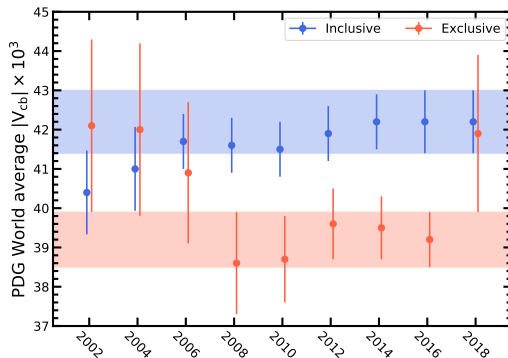
- Pure leptonic: $B_c \rightarrow \ell \bar{\nu}$. **Unavailable @B-factories**
- Pure hadronic: **Hard theoretically**
- **Semileptonic:**
 - Theory: **Small EW corrections; QCD uncertainties under control**
 - Experiment: **Only one neutrino missing, good BRs ($\approx 10\%$)**
 - \implies Best opportunity to measure $|V_{cb}|$
 - Exclusively: Either $B \rightarrow D^* \ell \bar{\nu}_\ell$ (this analysis) or $B \rightarrow D \ell \bar{\nu}_\ell$
 - Inclusively: $B \rightarrow X_c \ell \nu$

Exclusive decays



Exclusive vs Inclusive Analyses

- Tension between **inclusive** ($B \rightarrow X_{cl\nu}$) and **exclusive** ($B \rightarrow D^{(*)}l\bar{\nu}_l$) determination of V_{cb} ! $> 3\sigma$ (2018)



[1902.09412]

- Speculation about tension being related to FF parametrization in exclusive mode (CLN vs BGL)

How to extract $|V_{cb}|$ (exclusive)

- Say we want to extract $|V_{cb}|$ from $B \rightarrow D^* \ell \bar{\nu}_\ell$

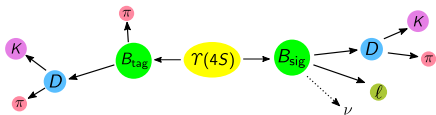
$$\frac{d\Gamma(B \rightarrow D^* \ell \bar{\nu}_\ell)}{dw} = \text{stuff we know} \cdot \text{FF}^2(w) \cdot |V_{cb}|^2,$$

- $w = \vec{v}_B \vec{v}_{D^{(*)}} = \frac{m_B^2 + m_{D^{(*)}}^2 - q^2}{2m_B m_{D^{(*)}}}$, $q^2 = (p_B - p_{D^{(*)}})^2$
- $\text{FF}^2(w)$: **form factors**
- in order to get $|V_{cb}|^2$ we need to know something about $\text{FF}^2(w)$
- Lattice QCD can only confidently tell us $\text{FF}^2(1)$ (but this corresponds to $q^2 = 0$ and we don't have data there)
- Only solution: Use parameterized model for $\text{FF}^2(w)$ (model dependency! CLN vs BGL etc.), fit parameters, extrapolate to $w = 1$ and then use lattice result to get $|V_{cb}|^2$
- In other words: need to **fit form factor parameters** \implies **measure kinematic distributions**
- $B \rightarrow D \ell \bar{\nu}_\ell$ is similar

Tagged vs Untagged Analyses

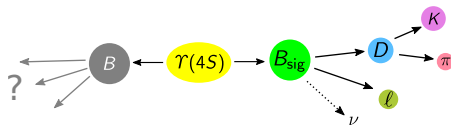
Tagged Analysis

- + High purity
- Low efficiency
(0.3% @Belle \rightarrow 0.55% @Belle II)
- + Get B meson momentum from tag side!



Untagged Analysis

- Low purity
- + High efficiency
(11% @Belle \rightarrow 20% @Belle II)



Part 2: $|V_{cb}|$ from hadronically tagged $B \rightarrow D^* l \bar{\nu}_l$ at Belle: Tag side & tagging calibration

3 Introduction

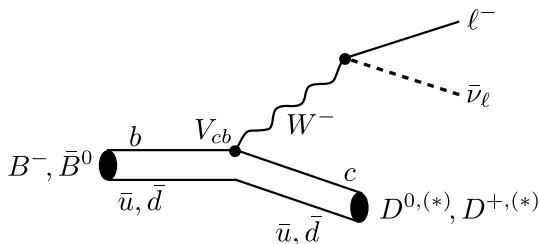
4 Dataset

5 Fitting

6 Results

7 Summary

Goals

Reanalysis of hadronically tagged $B \rightarrow D^* \ell \bar{\nu}_\ell$ at Belle

- Previous measurement: [\[1702.01521\]](#) (preliminary)
- Extract $|V_{cb}|$, form factors, kinematic distributions (separately for e/μ , potentially also investigating model dependencies)
- Belle dataset still exceeds Belle II dataset by large margin
- Expect around $2\times$ efficiency when using Belle II software (FEI)

⇒ Use Belle II software on Belle data! (b2bii)

Improved hadronic tagging @Belle II

[1808.10567]

Hadronic tagging @Belle II:

Around 5000 channels!

 $2.5\times$ efficiency!

Algorithm	MVA	Efficiency	Purity
Belle v1	Cut	—	—
Belle v3	Cut	0.1%	0.25%
Belle NB	NB	0.2%	0.25%
Belle II FEI	BDT	0.5%	0.25%

B^+ modes	B^0 modes	D^+ , D^{*+} , D_s^+ modes	D^0 , D^{*0} modes
$B^+ \rightarrow \bar{D}^0 \pi^+$	$B^0 \rightarrow D^- \pi^+$	$D^+ \rightarrow K^- \pi^+ \pi^+$	$D^0 \rightarrow K^- \pi^+$
$B^+ \rightarrow \bar{D}^0 \pi^+ \pi^0$	$B^0 \rightarrow D^- \pi^+ \pi^0$	$D^+ \rightarrow K^- \pi^+ \pi^+ \pi^0$	$D^0 \rightarrow K^- \pi^+ \pi^0$
$B^+ \rightarrow \bar{D}^0 \pi^+ \pi^0 \pi^0$	$B^0 \rightarrow D^- \pi^+ \pi^+ \pi^-$	$D^+ \rightarrow K^- K^+ \pi^+$	$D^0 \rightarrow K^- \pi^+ \pi^+ \pi^-$
$B^+ \rightarrow \bar{D}^0 \pi^+ \pi^+ \pi^-$	$B^0 \rightarrow D_s^+ D^-$	$D^+ \rightarrow K^- K^+ \pi^+ \pi^0$	$D^0 \rightarrow \pi^- \pi^+$
$B^+ \rightarrow D_s^+ \bar{D}^0$	$B^0 \rightarrow D^+ \pi^-$	$D^+ \rightarrow K_S^0 \pi^+$	$D^0 \rightarrow \pi^- \pi^+ \pi^0$
$B^+ \rightarrow \bar{D}^0 \pi^+$	$B^0 \rightarrow D^+ \pi^- \pi^0$	$D^+ \rightarrow K_S^0 \pi^+ \pi^0$	$D^0 \rightarrow K_S^0 \pi^0$
$B^+ \rightarrow \bar{D}^{*0} \pi^+ \pi^0$	$B^0 \rightarrow D^+ \pi^- \pi^+ \pi^-$	$D^+ \rightarrow K_S^0 \pi^+ \pi^+ \pi^-$	$D^0 \rightarrow K_S^0 \pi^+ \pi^-$
$B^+ \rightarrow \bar{D}^{*0} \pi^+ \pi^+ \pi^-$	$B^0 \rightarrow D^+ \pi^- \pi^+ \pi^+ \pi^- \pi^0$	$D^{*+} \rightarrow D^0 \pi^+$	$D^0 \rightarrow K_S^0 \pi^+ \pi^- \pi^0$
$B^+ \rightarrow \bar{D}^0 \pi^+ \pi^+ \pi^- \pi^0$	$B^0 \rightarrow D_s^+ D^-$	$D^{*+} \rightarrow D^+ \pi^0$	$D^0 \rightarrow K^- K^+$
$B^+ \rightarrow D_s^+ \bar{D}^0$	$B^0 \rightarrow D_s^+ D^{*-}$	$D_s^+ \rightarrow K^+ K_S^0$	$D^0 \rightarrow K^- K^+ K_S^0$
$B^+ \rightarrow D_s^+ \bar{D}^{*0}$	$B^0 \rightarrow D_s^+ D^{*-}$	$D_s^+ \rightarrow K^+ \pi^+ \pi^-$	$D^{*0} \rightarrow D^0 \pi^0$
$B^+ \rightarrow \bar{D}^0 K^+$	$B^0 \rightarrow J/\psi K_S^0$	$D_s^+ \rightarrow K^+ K^- \pi^+$	$D^{*0} \rightarrow D^0 \gamma$
$B^+ \rightarrow D^- \pi^+ \pi^+$	$B^0 \rightarrow J/\psi K^+ \pi^+$	$D_s^+ \rightarrow K^+ K^- \pi^+ \pi^0$	
$B^+ \rightarrow J/\psi K^+$	$B^0 \rightarrow J/\psi K_S^0 \pi^+ \pi^-$	$D_s^+ \rightarrow K^+ K_S^0 \pi^+ \pi^-$	
$B^+ \rightarrow J/\psi K^+ \pi^+ \pi^-$		$D_s^+ \rightarrow K^- K_S^0 \pi^+ \pi^+$	
$B^+ \rightarrow J/\psi K^+ \pi^0$		$D_s^+ \rightarrow K^+ K^- \pi^+ \pi^+ \pi^-$	
$B^+ \rightarrow D^- \pi^+ \pi^+ \pi^0$	$B^0 \rightarrow D^- \pi^+ \pi^0 \pi^0$	$D_s^+ \rightarrow \pi^+ \pi^+ \pi^-$	
$B^+ \rightarrow \bar{D}^0 \pi^+ \pi^+ \pi^- \pi^0$	$B^0 \rightarrow D^- \pi^+ \pi^+ \pi^- \pi^0$	$D_s^{*+} \rightarrow D_s^+ \pi^0$	
$B^+ \rightarrow \bar{D}^0 D^+$	$B^0 \rightarrow \bar{D}^0 \pi^+ \pi^-$	$D^+ \rightarrow \pi^+ \pi^0$	$D^0 \rightarrow K^- \pi^+ \pi^0 \pi^0$
$B^+ \rightarrow \bar{D}^0 D^+ K_S^0$	$B^0 \rightarrow D^- D^0 K^+$	$D^+ \rightarrow \pi^+ \pi^+ \pi^-$	$D^0 \rightarrow K^- \pi^+ \pi^+ \pi^- \pi^0$
$B^+ \rightarrow \bar{D}^0 D^+ K_S^0$	$B^0 \rightarrow D^- D^0 K^+$	$D^+ \rightarrow \pi^+ \pi^+ \pi^- \pi^0$	$D^0 \rightarrow \pi^- \pi^+ \pi^+ \pi^-$
$B^+ \rightarrow \bar{D}^0 D^{*+} K_S^0$	$B^0 \rightarrow D^+ D^0 K^+$	$D^+ \rightarrow K^+ K_S^0 K_S^0$	$D^0 \rightarrow \pi^- \pi^+ \pi^0 \pi^0$
$B^+ \rightarrow \bar{D}^{*0} D^{*+} K_S^0$	$B^0 \rightarrow D^+ D^0 K^+$	$D^{*+} \rightarrow D^+ \gamma$	$D^0 \rightarrow K^- K^+ \pi^0$
$B^+ \rightarrow \bar{D}^0 D^0 K^+$	$B^0 \rightarrow D^- D^+ K_S^0$	$D_s^+ \rightarrow K_S^0 \pi^+$	
$B^+ \rightarrow \bar{D}^{*0} D^0 K^+$	$B^0 \rightarrow D^+ D^+ K_S^0$	$D_s^+ \rightarrow K_S^0 \pi^+ \pi^0$	
$B^+ \rightarrow \bar{D}^0 D^0 K^+$	$B^0 \rightarrow D^- D^+ K_S^0$	$D_s^{*+} \rightarrow D_s^+ \pi^0$	
$B^+ \rightarrow \bar{D}^{*0} D^{*0} K^+$	$B^0 \rightarrow D^+ D^+ K_S^0$		
$B^+ \rightarrow \bar{D}^{*0} \pi^+ \pi^0 \pi^0$	$B^0 \rightarrow D^+ \pi^- \pi^+ \pi^0$		

Calibration for excl. $B \rightarrow D^{(*)} \ell \bar{\nu}_\ell$: $B \rightarrow X \ell \nu$ (systematics limited)

Calibration of hadronic tagging

Tagging

- The FEI returns a list of B_{tag} candidates with a **likelihood** and the **decay channel**
- Often we need to know the **efficiency** of this selection
- Unfortunately this efficiency can be **different between MC and data** \implies calibration!

How?

- Reconstruct $\Upsilon(4S) \longrightarrow B_{\text{tag}} B_{\text{sig}}$ on MC and data
- **Idea:** Find B_{sig} channel with large BF that we **trust**
- Compare yields and blame all differences on B_{tag} reconstruction, i.e. the FEI
- Here: Inclusive $B_{\text{sig}} \longrightarrow Xlv$
- Need to do this in bins of the decay channel and the likelihood output

Dataset & MC fixes

The dataset

- Full Belle (1!) dataset (710 fb^{-1} , $771 \times 10^6 B\bar{B}$) + 1 stream of MC
- In total there are 10 streams of Belle MC (1 stream = MC equivalent of total data taken)
- “Generic” MC types: Charged (B^+B^-), Mixed ($B^0\bar{B}^0$), Charm ($c\bar{c}$), UDS ($q\bar{q}$)
- There is also special/rare/signal MC for special purposes

MC updates

- Belle MC was produced long ago \implies need to update to match better understanding
- Most simple updates: branching fractions (BFs)
- Only need to do this for most relevant BFs: $B \longrightarrow D^{(*)}$

In practice: MC reweighting

Simplest case

- Let's say our MC was generated with $BF(B \rightarrow XYZ) = 0.5$ (this is written in the decay file), but it's actually 0.25!
- Do we need to regenerate our MC? 🙄
- Easy fix: Simply add **weight** of 0.5 to all $B \rightarrow XYZ$ events

Uncertainties

- What if we only know $BF(B \rightarrow XYZ) = 0.5 \pm 0.1$?
- Uncertainty needs to be propagated to later results
- Add varied weight columns, i.e. `--weight--`, `--weight_xyz_up--`, `--weight_xyz_down--`
- Either repeat analysis for different weights or have other way to incorporate them (\rightarrow fitting)


Form factors (FF)

- Can also set weights to adjust *differential* cross sections! \implies Update FFs and more!

Dataset & MC fixes

- Lepton **PID** corrections
- Updated **BFs** for $B \rightarrow D^{(*)}$, **FF** reweighted for $B \rightarrow D, D^*, D^{**}$
- Generate dedicated **D^{**} samples** (family of poorly known particles)
- Big issue: The **gap**
 - Sum up all exclusive $B \rightarrow X\ell\nu$ modes in MC, e.g.
 $B \rightarrow D^{(*)}\ell\bar{\nu}_\ell, \dots$
 - Much smaller than direct measurement of $B \rightarrow X\ell\nu \implies$ Gap between sum of exclusive and inclusive
 - We need to fill this gap somehow \rightarrow educated guesses + large uncertainties
- Update Belle **gap filling** with dedicated samples $D_1(\rightarrow D\pi\pi)\ell^+\nu_\ell$, $D^*\eta\ell^+\nu_\ell$ and $D^{(*)}\pi\pi\ell^+\nu_\ell$ (currently 100% scaling uncertainty for total gap component)

In practice: Generating signal MC

-  You normally shouldn't generate MC yourself → data production liaison
- MC generation is governed by decay files → read them to understand your MC!
- There's a **online book lesson** about it

```

Define dm 0.507e12

Alias myB0 B0
Alias myAnti-B0 anti-B0
ChargeConj myB0 myAnti-B0

Decay Upsilon(4S)
1.00000 B0 anti-B0 myB0 myAnti-B0 VSS_BMIX dm;
Enddecay

Alias myD_1- D_1-
Alias myD_1+ D_1+
ChargeConj myD_1- myD_1+

Decay myB0
1.000 myD_1- e+ nu_e PHOTOS ISGW2;
Enddecay
CDecay myAnti-B0

Decay myD_1-
1.000 anti-D*0 pi- PHOTOS VVS_PWAVE 0.0 0.0 0.0 0.0 1.0 0.0;
Enddecay
CDecay myD_1+

End

```

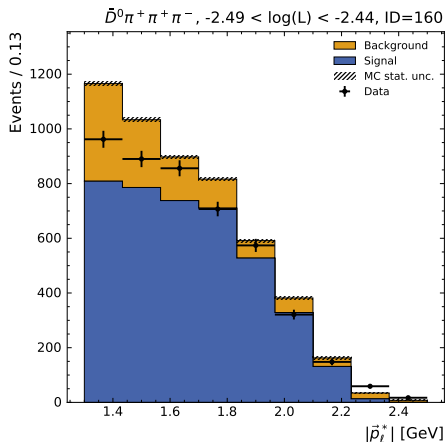
 In practice: Cuts & Best candidate selection

- Cuts & best candidate selection drastically reduce memory/disk requirements
- **Dilemma**: Do them “online” (using basf2) or “offline” (e.g. pandas)
- **Online**: Change cut value needs reprocessing with basf2 to (takes a lot of time)
- **Offline**: Dealing with a lot of data (memory & disk space issues for Belle dataset – Belle 2 data is smaller, so you might be fine) → consider looking into dask and friends for an **intermediate step**
- Makes sense to use soft online cuts and harder offline cuts
- Similar problem: Which **columns** to write out? → columnar data formats might at least help with fast reading of partial column set
- No golden rule, but **deliberate choices** are better

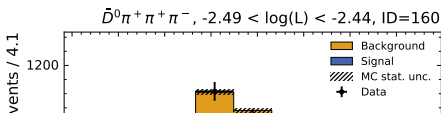
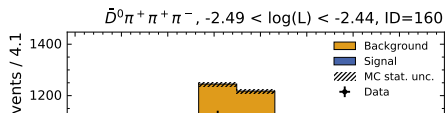
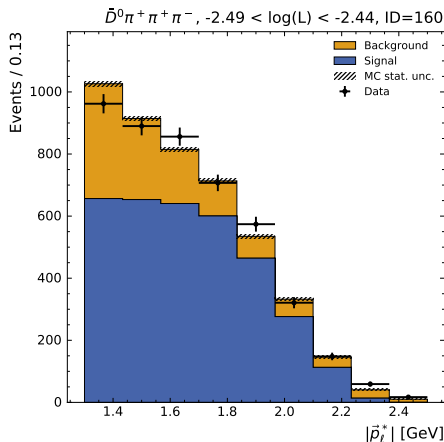
Example fits

$$B_{\text{sig}} \rightarrow X \ell \nu, B_{\text{tag}} \rightarrow \bar{D}^0 \pi^+ \pi^+ \pi^-$$

Before fitting



Fitted



In practice: Fitting frameworks

- Many fitting frameworks: RooFit, zFit, pyHF, ... → **choose wisely!**
 - Vary a lot in usability and quality of documentation!
 - Good to think ahead! (Features? How are uncertainties handled? Integration in your code? Help?)
- Different ways to **incorporate uncertainties**: Either
 - fits with fixed templates/weights → repeat fits with varied templates/weights → collect results and calculate uncertainty from differences
 - have more complicated likelihood where uncertainties are **nuisance parameters**

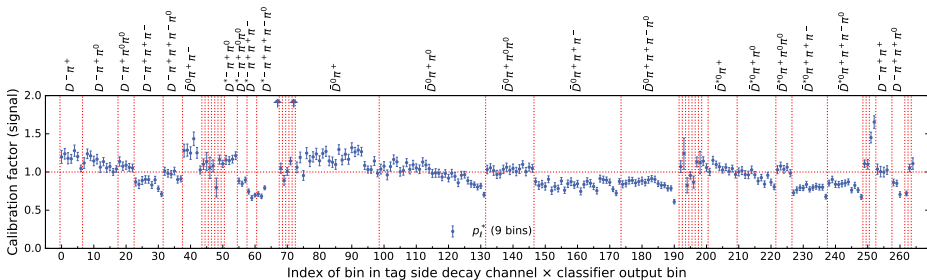
$$\mathcal{L} = \prod_i \mathcal{P}(n_i; \nu_i) \times \prod_k \mathcal{G}_k(\vec{0}, \vec{\theta}_k, \Sigma_k)$$

$$\nu_i = \sum_k \eta_k \prod_s (1 + \tilde{\theta}_{ks}) \frac{\eta_{ik}(1 + \theta_{ik})}{\sum_j \eta_{jk}(1 + \theta_{jk})},$$

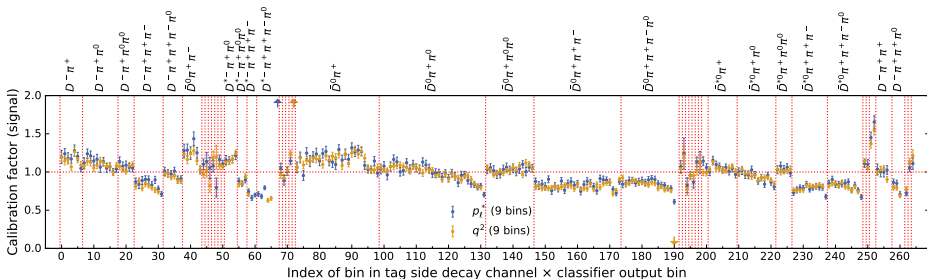
The calibration factors

CFs and uncertainties

Calibration factors for all 265 bins



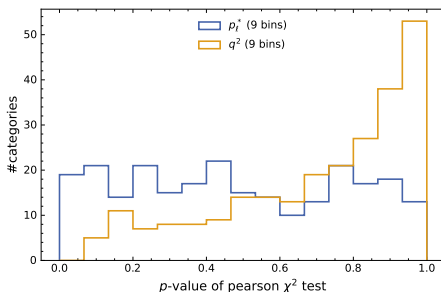
Calibration factors for all 265 bins



In practice: Validating fits

How do we know to trust our fits?

- Expect to repeat fits very, very, often. . .
- Simple validation: **Linearity check**
 - Generate artificial dataset from MC “bootstrapping” with different signal strengths
 - Fit this
 - Expect to get back signal strength that we put in
- Can also look at p -value distribution (should be flat!)



Summary

$|V_{cb}|$

- $|V_{cb}|$ needed to **test unitarity** of CKM matrix
- $\text{BR}(B \rightarrow D^{(*)} \ell \bar{\nu}_\ell) = \text{stuff we know} \cdot \text{FF}^2(w) \cdot |V_{cb}|^2 \rightarrow$ **exclusive measurement**
- Form factors $\text{FF}^2(w)$ make trouble
- Other method of extraction: **Inclusive measurement**; in tension with exclusive measurements

This analysis: $|V_{cb}|$ from hadronically tagged $B \rightarrow D^* \ell \bar{\nu}_\ell$ at Belle

- B_{tag} reconstructed by the **FEI**
- The FEI has **different efficiencies** on data and MC \rightarrow Calibration
- For calibration need something we trust: $B_{\text{sig}} \rightarrow X \ell \nu$
- Reconstruct $\mathcal{Y}(4S) \rightarrow B_{\text{tag}} B_{\text{sig}}$; blame efficiency difference on FEI \rightarrow calibration factors

Backup

Cuts & best candidate selection

Electrons, Muons

e	μ
$dr < 2 \text{ cm}$	
$ dz < 4 \text{ cm}$	
$ p_T^* > 0.1 \text{ GeV}/c$	
$p > 0.3 \text{ GeV}/c$	$p > 0.6 \text{ GeV}/c$
$eIDBelle > 0.6$	$muIDBelle > 0.9$
$muIDBelleQuality = 1$	

Photon

$goodBelleGamma = 1$

Event

Number of tracks with $dr < 2 \text{ cm}$ and

$|dz| < 4 \text{ cm}$: ≤ 17

Number of FS photons: < 18

 B_{tag}

$M_{bc} > 5.27 \text{ GeV}/c^2$

signal probability > 0.001

CS output > 0.2 (CS via BDT)

Best Candidate Selection

- Pick B_{tag} with smallest ΔE
- Pick lepton with highest $|p_\ell^*|$
- Still multiple candidates \rightarrow pick random

Fit strategies

- Decay channels vary significantly in number of events \implies number of bins in FEI classifier depends on decay channel (currently aiming for > 4500 events per bin)
- Perform a fit to extract yields for signal
- **Calibration factor** = fitted yield / MC yield
- **Binned maximum likelihood fit** (systematics included in likelihood \implies pull on the shape of the distributions \rightarrow Backup)

Current strategy: 2 component fit

- Fake leptons + secondary leptons + continuum are fitted as one component (background) \rightarrow Requires additional systematic uncertainty \rightarrow todo
- Fitted with **2 components**: Signal + background
- Number of bins fixed to 9 (will be made dynamic later)

Fitting and systematics

Maximize binned likelihood, allowing systematics to pull on the shapes:

$$\mathcal{L} = \prod_i \mathcal{P}(n_i; \nu_i) \times \prod_k \mathcal{G}_k(\vec{0}, \vec{\theta}_k, \Sigma_k)$$

$$\nu_i = \sum_k \eta_k \prod_s (1 + \tilde{\theta}_{ks}) \frac{\eta_{ik}(1 + \theta_{ik})}{\sum_j \eta_{jk}(1 + \theta_{jk})},$$

where

- \mathcal{P} : Poisson distribution
- n_i : Measured events in bin i
- ν_i : Expected events in bin i
- k : Component of fit
- η_k : Yield of fit template k
- η_{ik} : Bin content of fit template k , bin i
- θ_{ik} : Additive nuisance parameter bin i , fit component k
- $\tilde{\theta}_{is}$: Multiplicative nuisance parameter from source s for fit component k
- \mathcal{G} : Multivariate Gaussian with covariance matrix Σ_k

Fitted: $\eta_k, \theta_{ik}, \tilde{\theta}_{ks}$

Reconstruction signal side

π , K meson

Identification via PID likelihood ratio, impact parameters.

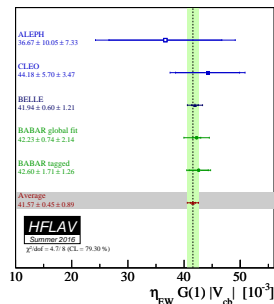
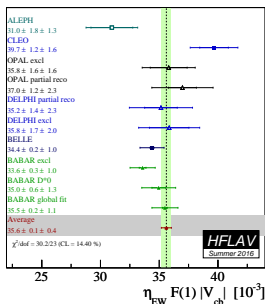
π^0 : From γ candidates (clusters in calorimeter not matched to any track)

D , D^* meson

arXiv	Signal	Tag	D^0 modes	D^+ modes	D^{*-} modes
[1010.5620]	$D^* \ell \bar{\nu}_\ell$	No	$K^- \pi^+$		$D^- \pi^-$
[1809.03290]					
[1702.01521]	$D^* \ell \bar{\nu}_\ell$	Had.	$K^- \pi^+(\pi)(\pi)$	$K^- \pi^+ \pi^+$	$\bar{D}^0 \pi^-$, $D^- \pi^0$
[1510.03657]	$D \ell \bar{\nu}_\ell$	Had.	$K^- \pi^+(\pi)(\pi)$, $K_S^0 \pi^+ \pi^- (\pi^0)$, $K_S^0 \pi^0$, $K^+ K^-$, $\pi^+ \pi^- (\pi^0)$, $K_S^0 K_S^0$, $\pi^0 \pi^0$, $K_S^0 \pi^0 \pi^0$, $K^- \pi^+ \pi^+ \pi^- \pi^0$	$K^- \pi^+ \pi^+ (\pi^0)$, $K_S^0 \pi^+ (\pi^0)$, $K^+ K^- \pi^+$, $K_S^0 K^+$, $K_S^0 \pi^+ \pi^+ \pi^-$, $\pi^+ \pi (\pi)$, $K^- \pi^+ \pi^+ \pi^+ \pi^-$	

Results and Outlook @Belle

Link	Channel	Tag	CLN	BGL	Unfold	Notes
[1010.5620]	$D^* \ell \bar{\nu}_\ell$	No				
[1809.03290]	$D^* \ell \bar{\nu}_\ell$	No	38.4 ± 0.9	42.5 ± 1.0		
[1702.01521]	$D^* \ell \bar{\nu}_\ell$	Had.	37.4 ± 1.3		Prelim.	Soon: Separate results $\ell = e$ and $\ell = \mu$
[1510.03657]	$D \ell \bar{\nu}_\ell$	Had.	39.86 ± 1.33	40.83 ± 1.13		



Projection in bins of kinematic variable

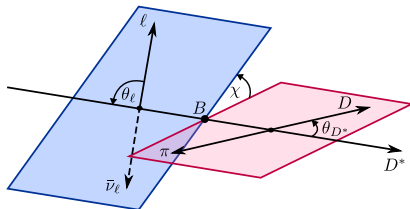
$$B \rightarrow D\ell\bar{\nu}_\ell$$

10 equal-size bins in w .

Good resolution (0.005) vs bin width (0.06) \implies Bin migration neglected

$$B \rightarrow D^*\ell\bar{\nu}_\ell$$

- 10 equal size bins in w , χ , $\cos\theta_\ell$, $\cos\theta_{D^*}$ (Projections)
- Correlation between the 4 distributions (\rightarrow toy experiments)
- Finite resolution \implies **Migration!**
 - \rightarrow Mig. matrix from truth vs reco MC.
 - \rightarrow Fold theory (easy) or unfold measurement (hard)



Fit variables

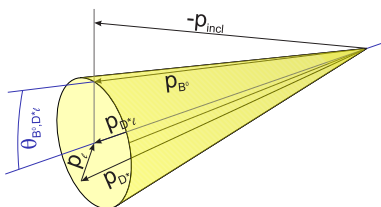
Tagged analyses

Fit variable: $m_{\text{miss}}^2 := (p_B - p_{D^{(*)}} - p_\ell)^2$ with $p_B = p_{\text{LER}} + p_{\text{HER}} - p_{\text{tag}}$

Correct reco \implies Peak at 0; Missed particles \implies Peak > 0 ; Particles from tag side \implies Peak < 0

Untagged analyses

Fit variables:



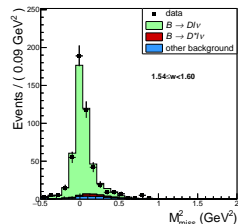
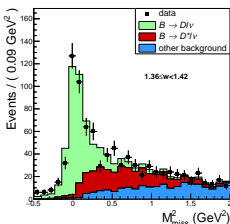
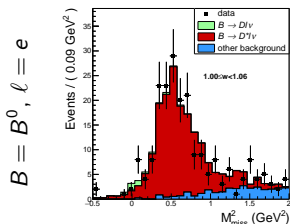
$$\cos \theta_{B, D^{*l}} := \frac{2E_B^* E_{D^{*l}}^* - m_B^2 - m_{D^{*l}}^2}{2|\vec{p}_B^*||\vec{p}_{D^{*l}}^*|}$$

Correct reco $\implies -1 \leq \cos \theta_{B, D^{*l}} \leq 1$

$$\Delta m = m_{D^{*}} - m_D$$

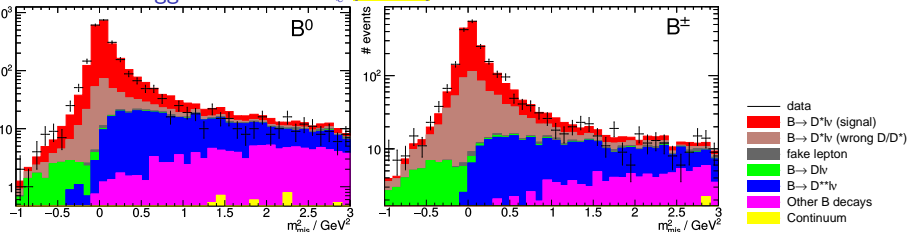
$$p_\ell$$

Fit strategies

Belle tagged $B \rightarrow D\ell\bar{\nu}_\ell$ [1510.03657]

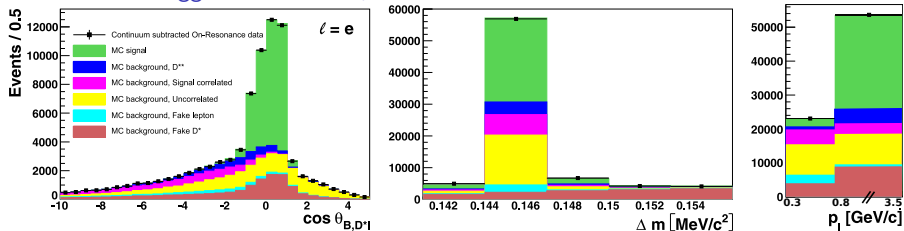
Fit	Binned extended likelihood fit (Barlow, Beeston, 1993)
Fit variable	$m_{\text{miss}}^2 := (p_B - p_D - p_\ell)^2$
Templates	From MC
Fixed Norm.	“other” background from MC
Float. Norm.	2: $B \rightarrow D\ell\bar{\nu}_\ell$ and $B \rightarrow D^*\ell\bar{\nu}_\ell$ normalization

Fit strategies

Belle tagged $B \rightarrow D^* l \bar{\nu}_l$ [1702.01521]

Fit	Unbinned likelihood fit
Fit variable	$m_{\text{miss}}^2 := (p_B - p_{D^*} - p_l)^2$
Templates	From MC
Fixed Norm.	Ratios of background normalizations from MC
Float. Norm.	2: Correctly reco. sig. + sig. with $D^{(*)}$ wrongly reco. norm.; total background normalization

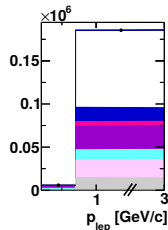
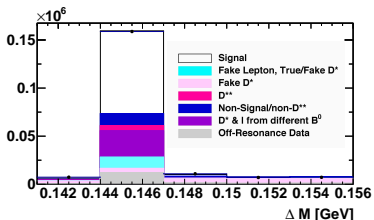
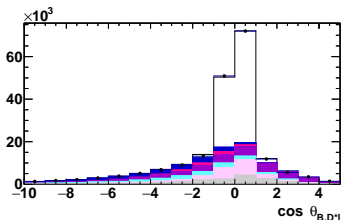
Fit strategies

Belle untagged $B \rightarrow D^* \ell \bar{\nu}_\ell$ [1010.5620]

Fit	Binned likelihood fit
Fit variable	$\cos \theta_{B,D^* \ell}$, Δm , p_ℓ
Templates	Continuum from off-resonance, rest from MC ¹
Fixed Norm.	Continuum from off-resonance (corrected for $1/s$ dependency)
Float. Norm.	6 : Normalizations for signal and backgrounds

¹For $\ell = \mu$: Shape of fake ℓ corr. with data from $K_S^0 \rightarrow \pi^+ \pi^-$; ℓ PID eff. corr. with data from $2\gamma \rightarrow e^+ e^- / \mu^+ \mu^-$

Fit strategies

Belle untagged $B \rightarrow D^* \ell \bar{\nu}_\ell$ (new) [1809.03290]

Fit	Binned likelihood fit
Fit variable	$\cos \theta_{B,D^* \ell}$, Δm , p_ℓ
Templates	Continuum from off-resonance, rest from MC ²
Fixed Norm.	Continuum from off-resonance (corrected for $1/s$ dependency, kinematics)
Float. Norm.	6 : Normalizations for signal and backgrounds

²Shape of fake ℓ corr. with data from $D^* \rightarrow D^0 \pi$, $D^0 \rightarrow K \pi$; lepton PID eff. corr. with data from $ee \rightarrow ee \gamma$, $ee \rightarrow \mu \mu (\gamma)$ and $J/\psi \rightarrow \ell^+ \ell^-$; low momentum track reco. eff. corr. with control sample of $B \rightarrow D^* \ell \bar{\nu}_\ell$

V_{cb} and form factors fit χ^2 fit

$$\chi^2 = \left(\vec{v}_{\text{sig}} - \vec{v}_{\text{sig}}^{\text{pred}} \right) C^{-1} \left(\vec{v}_{\text{sig}} - \vec{v}_{\text{sig}}^{\text{pred}} \right) + \chi_{\text{nuisance}}^2,$$

where:

- \vec{v}_{sig} yields in bins of kinematic variables
($D\ell\bar{\nu}_\ell$: w , $D^*\ell\bar{\nu}_\ell$: w , χ , $\cos\theta_\ell$, $\cos\theta_{D^*}$)
- $\vec{v}_{\text{sig}}^{\text{pred}} = (\epsilon_{\text{reco}}\epsilon_{\text{tag}})\mathcal{M}_{\text{mig}}\Delta\vec{\Gamma}$
 $\Delta\vec{\Gamma}_i$: theory expectation diff. CS in bin i
 (depends on FF param, $|V_{cb}|$),
 \mathcal{M}_{mig} : migration matrix
- C : Covariance matrix
- χ_{nuisance}^2 : Account for multiplicative factors degenerate with $|V_{cb}|$

