

# Search for the Lepton Flavour Violating decays $B^0 \rightarrow e^\pm \mu^\mp$ and $B_s^0 \rightarrow e^\pm \mu^\mp$ with LHCb Run 2 data

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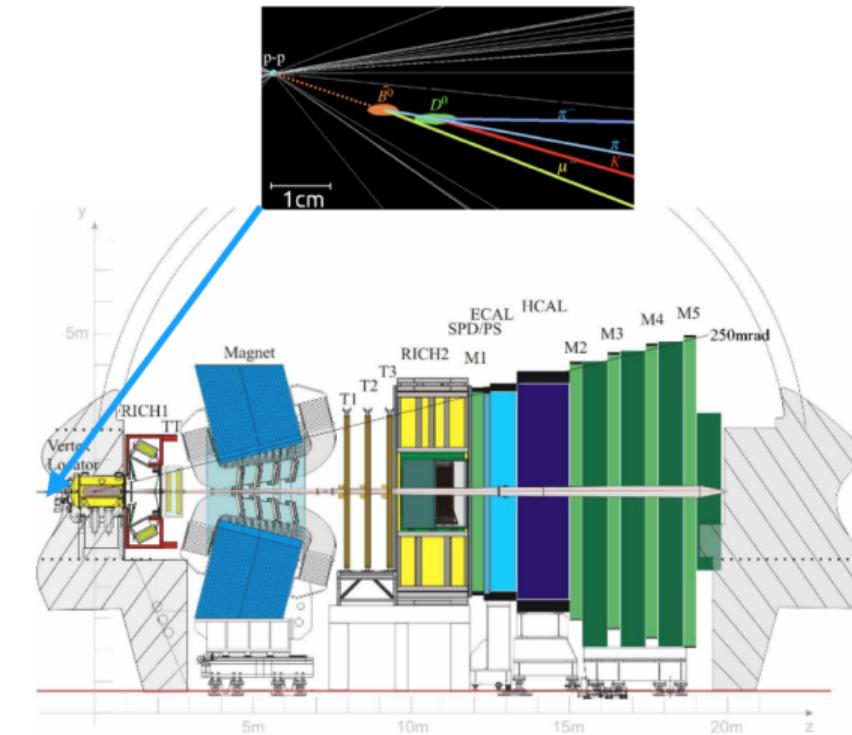
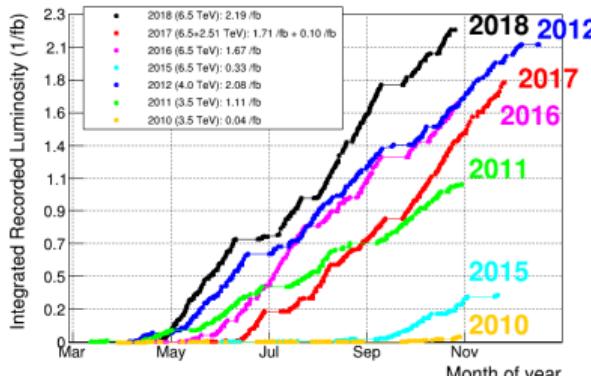


SPS annual meeting

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# The LHCb experiment

- **Asymmetric forward spectrometer (  $2 < \eta < 5$  )**  
→ designed for b and c physics
- excellent vertex, mass and momentum resolution
- very good particle identification
- recorded integrated luminosity:  
→ Run 1:  $3.23 \text{ fb}^{-1}$   
→ Run 2:  $5.85 \text{ fb}^{-1}$



# Search for Lepton Flavour Violating decays

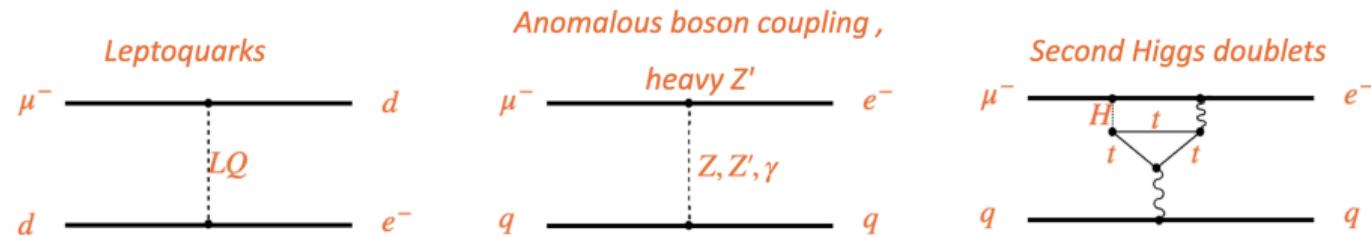
- **Lepton flavour violation (LFV)**

- observation of neutrino oscillations implies LFV
- not observed in the charged lepton sector

- ▶ [1] D. Bečirević et al, Phys. Rev. D 94, 115021
- ▶ [2] I.de Medeiros Varzielas et al, JHEP 06 (2015) 072
- ▶ [3] A. Crivellin et al, Phys.Rev.D 92 (2015) 5, 054013
- ▶ [4] R.A. Diaz et al, Eur.Phys.J.C 46 (2006) 403-405

- **Search for forbidden b-hadron decays in the SM (e.g.  $\rightarrow e\mu, \rightarrow \tau\mu, \rightarrow e\tau$ )**

- Standard Model branching fraction is  $< 10^{-50}$
- can be enhanced by new mediating particles up to  $10^{-11}$
- several theoretical models predict LFV (leptoquarks , new gauge boson  $Z'$ , Higgs doublets) [1,2,3,4]



# Links to Lepton Flavour Universality

## ● Lepton flavour universality (LFU)

- scenarios opened by recent hints of LFU anomalies [1,2,3]
- links in some models between LFU and LFV [4,5]

$$R_K = \frac{\mathcal{B}(B \rightarrow K \mu^+ \mu^-)}{\mathcal{B}(B \rightarrow K e^+ e^-)}$$

$$\mathcal{B} \rightarrow K \mu^\pm e^\mp \sim 3 \cdot 10^{-8} \left( \frac{1-R_K}{0.23} \right)^2$$

$$\mathcal{B} \rightarrow K(e^\pm, \mu^\pm) \tau^\mp \sim 2 \cdot 10^{-8} \left( \frac{1-R_K}{0.23} \right)^2$$

$$\frac{\mathcal{B}(B_s \rightarrow \tau^+ (e^-, \mu^-))}{\mathcal{B}(B_s \rightarrow \mu^+ \mu^-)_{SM}} \sim 4 \left( \frac{1-R_K}{0.23} \right)^2$$

$$\frac{\mathcal{B}(B_s \rightarrow \mu^+ e^-)}{\mathcal{B}(B_s \rightarrow \mu^+ \mu^-)_{SM}} \sim 0.01 \left( \frac{1-R_K}{0.23} \right)^2$$

► [1] LHCb collaboration, Phys.Rev.Lett.115, 111803

► [2] LHCb collaboration, JHEP 08 (2017) 055

► [3] LHCb collaboration, Phys. Rev. Lett. 113, 151601

► [4] G. Hiller et al, arXiv: 1609.08895v2

► [5] S.L. Glashow et al, Phys. Rev. Lett. 114, 091801

# Search for $B_{(s)}^0 \rightarrow e^\pm \mu^\mp$

- Electrons produce bremsstrahlung

→ imperfect bremsstrahlung recovery

→ bremsstrahlung categories for  $B \rightarrow e\mu$ :  $0\gamma$ ,  $1\gamma$

- Current limits at 90(95) % CL (Run 1)

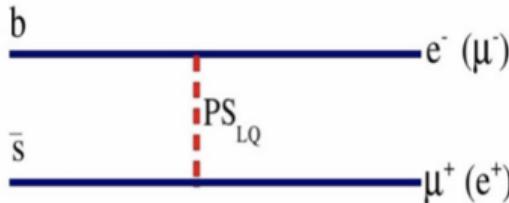
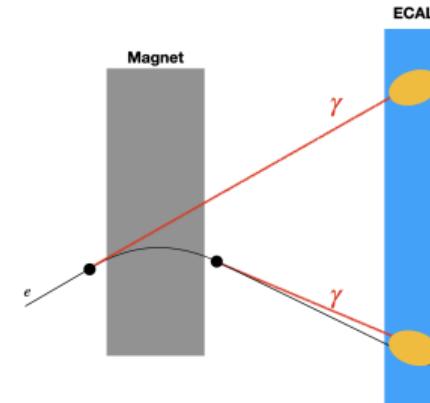
→  $\mathcal{B}(B_s \rightarrow e^\pm \mu^\mp) < 6.0(7.2) \times 10^{-9}$

→  $\mathcal{B}(B \rightarrow e^\pm \mu^\mp) < 0.9(1.2) \times 10^{-9}$

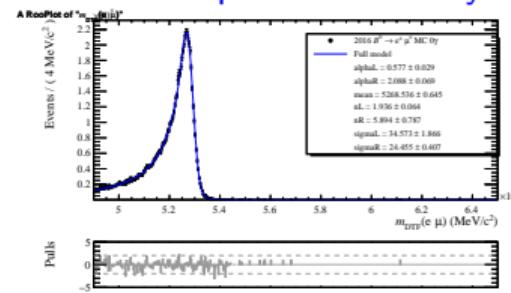
► LHCb collaboration, JHEP 03 (2018) 078

- Run 2 analysis: 2016 + 2017 + 2018

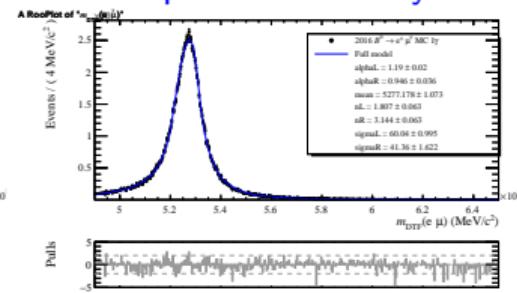
→ can expect factor  $\sim 2$  improvement from statistics



without photon recovery



with photon recovery



# Analysis strategy

- Measure  $\mathcal{B}(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)$  with respect to  $B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+$

$$\begin{aligned} \mathcal{B}(B_{(s)}^0 \rightarrow e^\pm \mu^\mp) &= \frac{f_u}{f_{d(s)}} \times \mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+) \\ &\times \frac{\mathcal{N}(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)}{\mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)} \\ &\times \frac{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)}{\varepsilon(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)} \end{aligned}$$

- Validation of the efficiency corrections checking  $r_{J/\psi} = 1$  in bremsstrahlungs categories

$$\begin{aligned} r_{J/\psi} &= \frac{\mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)}{\mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)} \\ &= \frac{\mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)}{\mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)} \times \frac{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)}{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)} \\ &= 1 \end{aligned}$$

# Analysis workflow

## 1. Selection

- Stripping, offline and trigger selection
- PID to remove physics background
- MVA to remove combinatorial background

## 2. Determine and correct for selection efficiency

- Correct for tracking, PID, L0 and  $B$  kinematics
- Use  $B^+ \rightarrow J/\psi(\mu^+\mu^-)K^+$  and  $B^+ \rightarrow J/\psi(e^+e^-)K^+$  as control and calibration modes

## 3. Determine $\mathcal{B}(B^0 \rightarrow e^\pm\mu^\mp)$ and $\mathcal{B}(B_s^0 \rightarrow e^\pm\mu^\mp)$

- Simultaneous fit of the  $e\mu$  mass split by years and brem categories
- Use  $B^+ \rightarrow J/\psi(\mu^+\mu^-)K^+$  as normalisation channel

## 4. Derive the limits for the branching fractions

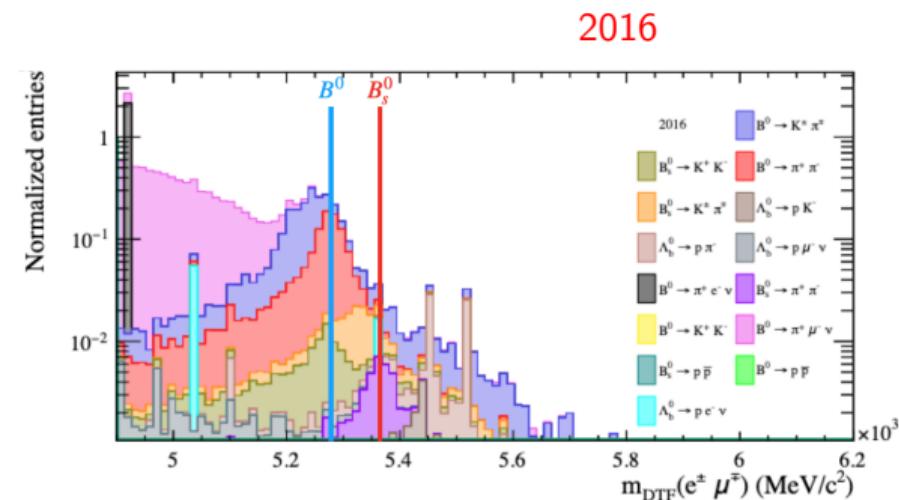
# Selection

## Pre-selection

- use dedicated pre-selection selections for our signal and normalisation channels
- Fiducial cuts, chosen to align with calibration samples acceptance, few examples:  
 $\rightarrow p_T(\mu) > 0.8 \text{ GeV}$ ,  $p_T(e) > 0.5 \text{ GeV}$ ,  $IP\chi^2(e) > 25$ ,  $p(e, \mu) < 200 \text{ GeV}$ .
- trigger on single electron or signal muon candidate, require good tracks and use decay topology

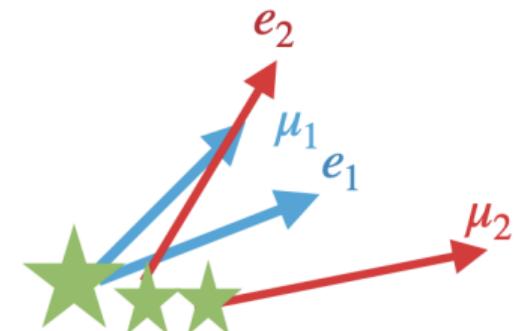
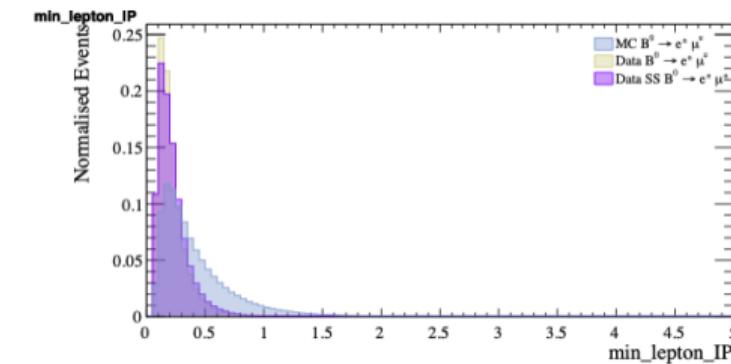
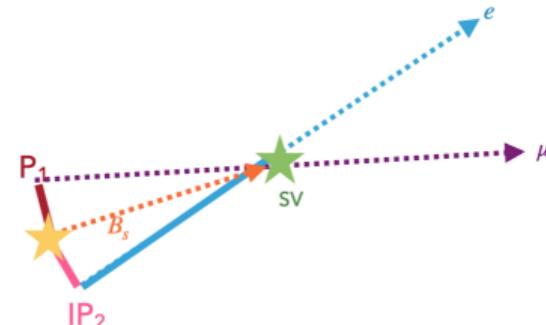
## Particle Identification PID

- Criteria to reduce and remove physics background ( $e \rightarrow K/\pi$ ,  $\mu \rightarrow K/\pi$ )
- Main peaking backgrounds considered: 2-body hadronic decays ( $B_{(s)}^0 \rightarrow h^+ h^-$  with  $h^{(')} = \pi, K, p$ )
- Also considered:  $B^0 \rightarrow \pi \ell \nu$ ,  $\Lambda_b \rightarrow p \ell \nu$ ,  $\Lambda_b \rightarrow p K$  and  $\Lambda_b \rightarrow p \pi$   
 $\rightarrow e : \text{PIDe} > -2$  and  $\text{MC15TuneV1ProbNN} > 0.8$   
 $\rightarrow \mu : \text{MC15TuneV1ProbNN} > 0.4$



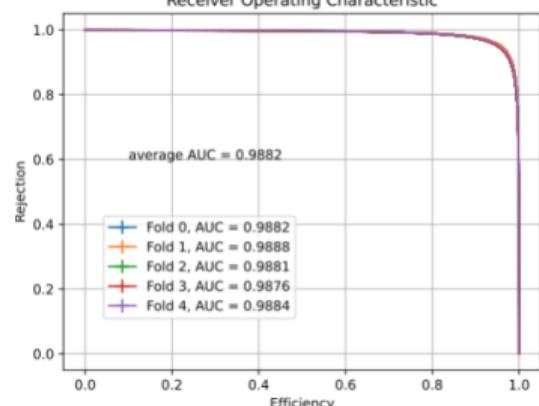
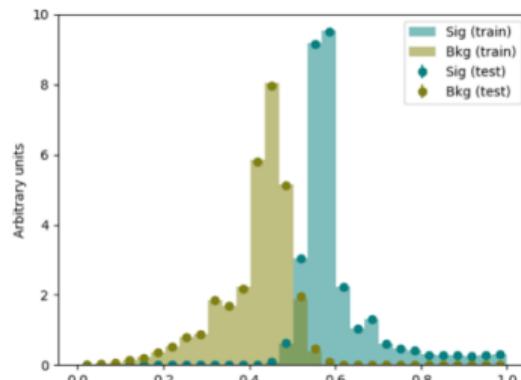
# Multivariate Analysis

- Combinatorial background: two tracks associated to a common vertex
- Train a Boosted Decision Tree (BDT) to remove combinatorial background
- Choose discriminating variables that contain information of the topology, vertex quality and track isolation  
→ e.g. the smallest of the lepton IPs with respect to the PV



# BDT training

- Signal: efficiency corrected  $B_{(s)}^0 \rightarrow e^\pm \mu^\mp$  simulation
- Background: sideband data
- Tested different sets of discriminating variables and algorithms
- Use a total of 14 discriminating variables
- Use cross validation (with k=5 folds) for training

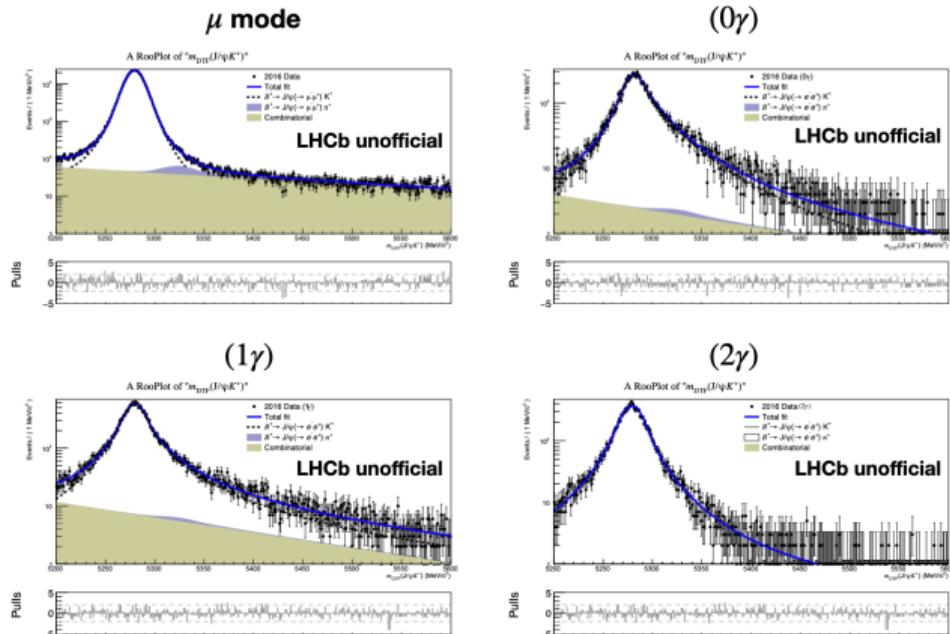


# Fits to $B^+ \rightarrow J/\psi(\rightarrow \ell^+\ell^-)K^+$

- Simultaneous fit in  $0\gamma$ ,  $1\gamma$  and  $2\gamma$  categories in electron modes and muon mode

$1\gamma$ : only one electron has brem added  
 $2\gamma$ : both electrons has brem added

- Floating  $\pi \rightarrow K$  mis-ID rate shared between brems and  $e/\mu$  mode.
- Fits used to validate  $w_{PID}$  &  $w_{L0}$  corrections measuring  $r(J/\psi)$



# Corrections

## Efficiency corrections

- Selection efficiencies,  $\varepsilon$ , are taken from simulation
- Well known that the tracking, the PID and the L0 trigger response is badly modeled in simulation  
→ derive corrections with data driven methods

## Kinematic corrections

- Observe discrepancies in the modelling of the B kinematics in simulation
- Train a BDT with the GBreweighter package to obtain corrections
- Corrections are obtained from  $B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+$  MC and sWeighted data
- Port corrections to  $B \rightarrow e^\pm \mu^\mp$  and  $B \rightarrow J/\psi(e^+ e^-) K^+$

→ validate our corrections, by measuring  $r_{J/\psi}$

# $r_{J/\psi}$ cross check

- Validate corrections ( $w_{TRK} \times w_{PID} \times w_{wL0} \times w_{wBKIN}$ )
- Calculate  $r_{J/\psi}$  for two bremsstrahlungs categories:  $0\gamma$   $1\gamma$

$$\begin{aligned} r_{J/\psi} &= \frac{\mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)}{\mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)} \\ &= \frac{\mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)}{\mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)} \times \frac{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow e^+ e^-) K^+)}{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)} \\ &= 1 \end{aligned}$$

year	Correction	$r_{J/\psi}(0\gamma)$	$r_{J/\psi}(1\gamma)$
2016	no corrections	1.14	1.29
	fully corrected	1.03	1.12
2017	no corrections	1.12	1.21
	fully corrected	1.01	1.06
2018	no corrections	1.20	1.31
	fully corrected	1.00	1.06

# Invariant mass fit

## Simultaneous fit of $m_{e\mu}^{DTF}$

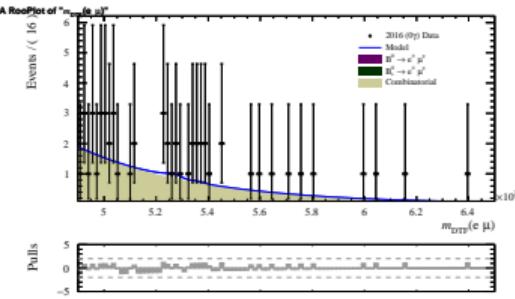
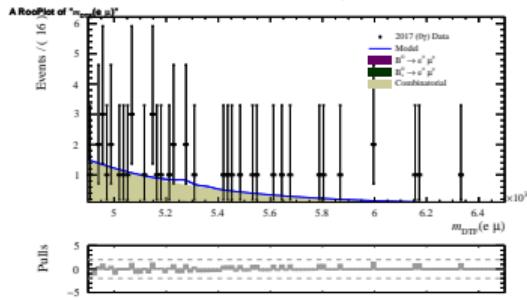
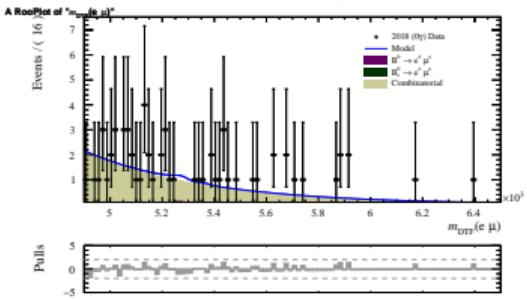
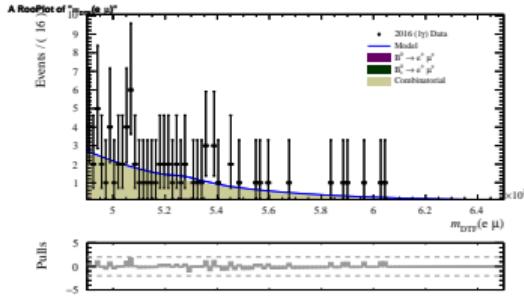
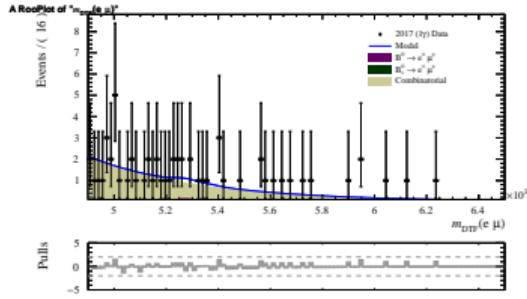
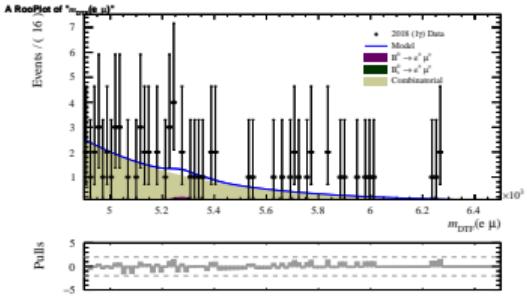
- Model:
  - $B^0 \rightarrow e^\pm \mu^\mp$ : bifurcated DSCB
  - $B_s^0 \rightarrow e^\pm \mu^\mp$ : bifurcated DSCB
  - Combinatorial: exponential
- Fit 6 datasets: 2 brem categories  $\times$  3 years
- Branching fractions are shared between all categories

$$\mathcal{N}(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)_{year,brem} = \frac{f_{d(s)}}{f_u} \times \frac{\varepsilon(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)_{year,brem}}{\varepsilon(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)_{year}} \times \frac{\mathcal{B}(B_{(s)}^0 \rightarrow e^\pm \mu^\mp)}{\mathcal{B}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)} \times \mathcal{N}(B^+ \rightarrow J/\psi(\rightarrow \mu^+ \mu^-) K^+)_{year}$$

## Fit under bkg only hypothesis

Toys: generate from fits to data sidebands, and extrapolation over full fit range

# Fit toy datasets

2016, 0 $\gamma$ 2017, 0 $\gamma$ 2018, 0 $\gamma$ 2016, 1 $\gamma$ 2017, 1 $\gamma$ 2018, 1 $\gamma$ 

# Summary

So far:

- Full selection in place
  - offline, trigger and alignment selection finalised
  - BDT trained and optimised to suppress combinatorial background
- Implemented full correction to simulation (tracking, PID, L0 and B kinematics)
- $r_{J/\psi}$  determined applying corrections
- Simultaneous fits to data and toy datasets

On-going:

- Sensitivity studies
- Background studies and validation using  $B \rightarrow hh$  stripping output
- Systematics from  $\varepsilon$  and mass fits