# Measurement of the CP violating phase, $\phi_s$ , in Run 2 using $B^0_s \to J\!/\psi\,K^+K^-$

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#### **CP** Violation

- CP Violation is a necessary condition for baryon asymmetry in the Universe [A. D. Sakharov, JETP Lett. 5, 24-27 (1967)]
- Present in the Standard Model, but too small by  $10^{10}$  to explain asymmetry
- Heavy-quark hadrons are excellent place to search for new sources of CPV



#### $\phi_s$ (CPV in interference of mixing and decay)

 $\mathcal{A}_{CP}(t) = \frac{\Gamma(\bar{\mathrm{B}}^{0}_{\mathrm{s}} \rightarrow f) - \Gamma(\mathrm{B}^{0}_{\mathrm{s}} \rightarrow f)}{\Gamma(\bar{\mathrm{B}}^{0}_{\mathrm{s}} \rightarrow f) + \Gamma(\mathrm{B}^{0}_{\mathrm{s}} \rightarrow f)} \approx \eta_{f} \sin \phi_{s} \sin(\Delta m_{s} t)$ 



φ<sub>s</sub> : phase difference between amplitudes w/ and w/o oscillation in b → cc̄s decays
 Sensitive probe of NP in B<sup>0</sup><sub>s</sub> mixing and decay

- $\phi_s^{SM} = -0.0365 \pm 0.0013$  rad [CKMFitter]
- $\phi_s^{AVG} = -0.021 \pm 0.031 \text{ rad} [\text{HFLAV Summer 2017}]$







LHCb Detector Layout

- Interaction VErtex LOcator ( $\varepsilon_{track} \approx 96\%$ )
- Ring-Imaging Cherenkov ( $\varepsilon_{PID}(K) \approx 95\%$ ) (*MisID*(K  $\rightarrow \pi$ )  $\approx 5\%$ )
- High-granularity Muon ( $\varepsilon_{PID}(\mu) \approx 97\%$ ) (*MisID*( $\mu \rightarrow \pi$ )  $\approx 3\%$ )
- 4% of solid angle = 40% of heavy quark cross-section
- Decay time resolution: 45 fs
- Run 1:  $\sim 3 \, \text{fb}^{-1}$
- Run 2 (2015 & 2016): ~ 2 fb



## Analysis strategy For Run 2 $\phi_s$ with $B_s^0 \rightarrow J/\psi \phi$



- ${\rm B}^0_{\rm s} 
  ightarrow {\rm J}/\psi\,\phi$  is the golden mode for measuring  $\phi_{\rm s}$
- Measure  $\phi_s$ ,  $\Delta \Gamma_s$ ,  $\Gamma_s \Gamma_d$
- Final state is a mixture of CP-even/CP-odd, requires angular analysis to disentangle  $CP|J/\psi\phi\rangle_{\ell} = (-1)^{\ell}|J/\psi\phi\rangle_{\ell}$
- $\bullet~\mbox{Good tagging performance to resolve } {\rm B}_{\rm s}^0$  flavour at production
- $\bullet~$  High decay-time resolution to see fast  ${\rm B}^0_{\rm s}$  oscillation and determine  $\Delta {\it m}_{\it s}$
- Flavour-tagged time-dependent angular fit
- Robust understanding and modelling of background and acceptance effects



#### Run 2:

- Higher centre-of-mass energy means 2x the heavy quark cross-section
- More statistical power

#### MC Corrections:

- Data-driven calibration of final-state Particle ID
- Multidimensional Gradient-Boosted Reweighting for Data-MC agreement

#### Selection:

- Multivariate analysis using Boosted Decision Tree

- Avoid variables that can bias angular or decay time distributions

#### New Invariant Mass Model:

- Double-sided Crystal Ball function with per-event mass

#### error as a conditional observable





#### Peaking Backgrounds



Negative weighted  $\Lambda_b^0$  Monte-Carlo embedded to subtract remaining background

 $\begin{array}{l} \Lambda_0^b \to J/\psi \, p \mathrm{K}: \\ & - \text{Veto event when } P(\mathrm{K} \to \mathrm{p})_{max} > 0.7 \text{ and consistent with} \\ \Lambda_0^b \pm 15 \mathrm{MeV} \\ \mathrm{B}^0 \to J/\psi \, \mathrm{K}^*(\mathrm{K}\pi): \\ & - \mathrm{Veto event when } P(\mathrm{K} \to \mathrm{K})_{max} < 0.35 \ P(\mathrm{K} \to \pi)_{max} > 0. \end{array}$ 

- Veto event when  $P({\rm K}\to{\rm K})_{max}<$  0.35,  $P({\rm K}\to\pi)_{max}>$  0.7 and consistent with  ${\rm B}^0\pm 15 MeV$ 

Year	$\Lambda_b^0 \rightarrow J/\psi  pK$		${ m B^0}  ightarrow { m J}\!/\!\psi{ m K^*}$	
	before veto	after veto	before veto	after veto
2015	4.3%	1.3%	0.2%	0.1%
2016	4.3%	1.2%	0.3%	0.1%

Remaining  $\Lambda^0_{\rm b}$  background is statistically subtracted to avoid biasing effect in angular distributions



#### Decay Time Resolution

Extract the detector resolution from a sample of promptly produced J/ $\psi$  mesons from the PV Resolution Model: Dirac-delta function and two exponentials convolved with a triple Gaussian with common mean + additional component to account for events reconstructed from wrong primary vertex

Calibration: Using single effective Gaussian computed from the dilution of the triple Gaussian

Dilution in bins of 
$$\delta_{t}$$
:  $D = \sum_{i=1}^{3} f_{i}e^{-\sigma_{i}^{2}\Delta m_{s}^{2}/2}$  Effective Gaussian width  $\sigma_{eff} = \sqrt{(-2/\Delta m_{s}^{2}) \ln D}$   

$$\int_{i=1}^{9} \int_{i=1}^{9} \int_{i=1$$

### Angular Acceptance

Angular acceptance effect is modelled with normalisation weights in the resultant PDF for each individual polarisation state

Angles are computed from a re-fit of fully reconstructed events - resolution improved by 40% Procedure:

- Normalisation weights from fully simulated signal events
- Iterative procedure to correct MC/Data kinematic difference



Projections of angular acceptances for 3 helicity angles



#### Decay Time Acceptance

#### New Strategy for Acceptance

Method: Using a high-yield channel with well-know lifetime for data control sample

- Select  $B^0 \rightarrow J/\psi K^*(\rightarrow K^+\pi^-)$  control the same way as signal channel  $B^0_s \rightarrow J/\psi \phi$
- Obtain acceptance of B<sup>0</sup> from data using known lifetime
- $\bullet~$  Correct for difference in  ${\rm B}^0_{\rm s}$  and  ${\rm B}^0$  using MC ratio:

$$\epsilon^{ ext{B}^{0}_{ ext{s}}}_{ ext{data}}(t) = \epsilon^{ ext{B}^{0}}_{ ext{data}}(t) imes rac{\epsilon^{ ext{B}^{0}_{ ext{s}}}_{ ext{sim}}(t)}{\epsilon^{ ext{B}^{0}}_{ ext{sim}}(t)}$$



## Flavour Tagging



- $\bullet~$  Crucial to tag  $\rm B^0_s$  flavour at production
- Taggers are Neural Nets optimized for Run 1
- Calibration works well with Run 2 data



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#### Results Data Fit Projections



### Summary

To do:

- Finalize systematics
- Unblind results

Expected statistical precision:

• 
$$\sigma(\phi_{\rm s}) = 0.042 \, \, {
m rad} \, ({
m Run} \, \, 1: \, 0.049 \, \, {
m rad})$$

• 
$$\sigma(\Delta\Gamma_s) = 0.008 \, {\rm ps}^{-1}$$
 (Run 1: 0.0091  ${\rm ps}^{-1}$ )

• 
$$\sigma(\Gamma_{\rm s} - \Gamma_{\rm d}) = 0.005 \, {\rm ps}^{-1}$$
 (HFLAV: 0.004)





## Thank you



## Backup



## Event Selection

Correct MC before training with:

- PIDCalib
- Gradient Boosted Reweighting to sWeighted data
- Avoid vars that introduce effects difficult to correct:
  - $IP\chi^2(K, \mu)$ : angular/time correlation, could impact uniformity of decay time efficiency
  - DIRA(B<sub>s</sub><sup>0</sup>): large decay time acceptance effect

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•  $P_T(K, \mu)$ : large angular acceptance effect

- max Tr  $\chi^2/\mathrm{ndf}(\mathrm{K})$
- min Log ProbNNk(K)
- max Tr  $\chi^2/\mathrm{ndf}(\mu)$
- min Log ProbNNmu( $\mu$ )
- Log ENDVERTEX  $\chi^2/\mathrm{ndf}(J/\psi)$
- *P*<sub>T</sub>(φ)
- ENDVERTEX  $\chi^2/\mathrm{ndf}(\mathrm{B_s^0})$
- Log DTF  $\chi^2/\mathrm{ndf}(\mathrm{B}^0_\mathrm{s})$
- Log IP $\chi^2/\mathrm{ndf}(\mathrm{B^0_s})$
- $P_T(B_s^0)$



ProbNNk resampled and compared to sWeighted data

Resample MC PID distributions (ProbNN) with PIDCalib:

- ProbNNk/mu (Κ, μ)
- Also correlate ProbNNk(K) to ProbNNpi/p(K) to later use for vetos
- Resample based on *P*, *P*<sub>T</sub> and nTracks.



### MC Corrections

#### GB Reweighting





Gradient Boosted Reweighting:

- Uses iterative chain of decision trees to equalise unbinned multi-dimensional distributions
- Reweight on  $P_T(B_s^0)$ ,  $\eta(B_s^0)$ , GhostProb(K,  $\mu$ ), nLongTracks
- Binned reweighter struggles to match all inputs



### Selection

Optimizing MVA cut





$$FOM = \frac{(\sum_{i} sw_{i})^{2}}{\sum_{i} (sw_{i}^{2})}$$

• 
$$F_{max} = 89206$$

- $F_{max}$  at BDTG3 = 0.78
- Sig 102822, Bkg 26419



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## Peaking backgrounds

 $\Lambda_b^0$  background

After veto we expect 1192.0  $\pm$  82.8  $\Lambda_b^0$  in  $m(J/\psi K^+K^-) \in [5200, 5550]$ MeV

Procedure to subtract with negative weights:

- Reweight  $\Lambda^0_b$  MC phase space to match the resonant structure in  $\Lambda^0_b\to J/\psi\, p{\rm K}$  pentaquark analysis
- Resample tagging information from data, missing in  $\Lambda_b^0$  MC
- Normalize negative weights to  $-1181 \Lambda_b^0$
- Merge with Data



#### Invariant Mass Model

Signal Model: Double-sided Crystal Ball function (CB2) with per-event mass error as a conditional observable, quadratic dependence in mass error:  $\sigma = s_1 \sigma_i + s_2 \sigma_i^2$ Background Model: Exponential for combinatorial and Gaussian distribution for  $B^0 \rightarrow J/\psi K^+K^-$  contribution

Procedure to statistically subtract background with negative weights:

- Fix tails of CB2 with Gradient Boosted reweighted MC
- Fit to  $m(J/\psi K^+K^-)$  with a Primary Vertex constraint
- Fit in 6 m(KK) bins [990, 1008, 1016, 1020, 1024, 1032, 1050] MeV



#### Invariant Mass Model

New Model takes care of the  $m(B_s^0)$  and helicity angle correlations Projections of the mass fit in 3 bins of  $\cos(\theta_{\mu})$ 





- calculate normalisation weights
- perform the final fit to data to obtain parameter estimates
- reweigh the simulated sample, with event weights defined by  $\epsilon = \textit{PDF}(\Omega)/\textit{PDF}_{gen}$
- reweigh simulated events such that the distributions of  $p(K^+)$  and  $p(K^-)$  match those in background subtracted data
- stop if  $\Delta_p/\sigma_p < 0.01$  for all physics parameters, p, continue with step 1., otherwise



#### Table: Overall tagging performance.

Category	Faction(%)	$\varepsilon_{ m tag}(\%)$	$\mathcal{D}^2$	$arepsilon_{ ext{tag}}\mathcal{D}^2(\%)$
OS-only	14.31	10.19	0.086	0.87±0.03
SSK-only	59.60	42.41	0.031	$1.32{\pm}0.37$
OS&SSK	26.09	18.57	0.099	$1.85{\pm}0.14$
Total	100.00	71.17	0.057	4.04±0.39





