

# Zimányi School (4-8 December, 2023)

## XY Factorization Bias in Luminosity Measurements

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HUN  
REN

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# What Is Luminosity?

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Introduction

2D Fits

Simulation

Choosing the Right  
Model(s)

Final Result

Conclusion

$$\mathcal{L}_{inst} = \frac{R}{\sigma_{vis}} = \frac{fN_1N_2}{2\pi\Sigma_x\Sigma_y}$$

- $\mathcal{L}_{inst}$ : instantaneous luminosity
- $R$ : measured rate of the quantity
- $\sigma_{vis}$ : visible cross-section
- Basically the rate of any physical process can be given as  
$$\frac{dN}{dt} = \mathcal{L}_{inst} \cdot \sigma_{vis}$$
- $f$ : machine revolution frequency (11245 Hz)
- $N_{1,2}$ : bunch populations
- $\Sigma_{x,y}$ : horizontal & vertical beam overlap sizes (widths),  
measured by vdM scans (next slide)

# About vdM Scans in a Nutshell

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Introduction

2D Fits

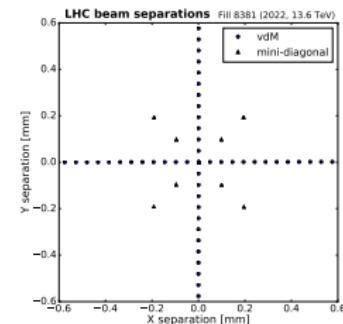
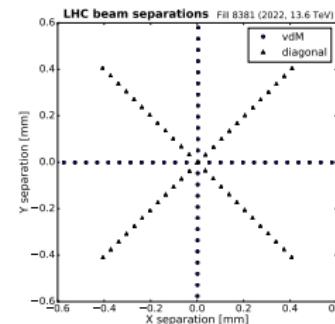
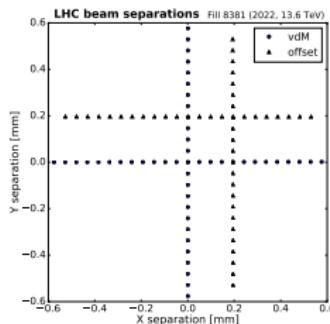
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- 2 bunches with  $\varrho_1(x, y)$  &  $\varrho_2(x, y)$  proton density functions collide
- vdM scans sample the convolution function along the axes:  
 $(\varrho_1 \cdot \varrho_2)(0, y)$  &  $(\varrho_1 \cdot \varrho_2)(x, 0)$



- vdM data separately fitted in the 2 directions with a Double Gaussian
- Width parameter calculated from the fits:  
$$\sigma_{vis} = \int \int (\varrho_1 \cdot \varrho_2)(x, y) dx dy \sim 2 \cdot \max(\varrho_1 \cdot \varrho_2) \cdot \Sigma_x \Sigma_y$$

# The XY Factorisation Bias

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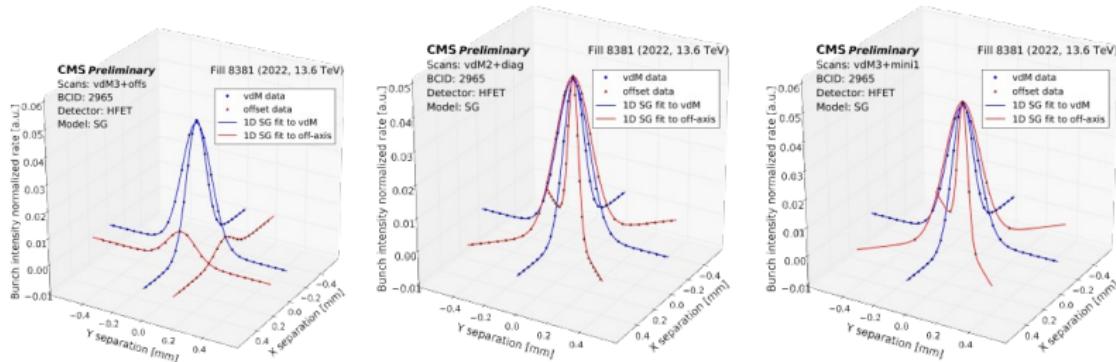
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BUT!  $\sigma_{vis} = \int \int (\varrho_1 \cdot \varrho_2)(x, y) dx dy \sim 2 \cdot \max(\varrho_1 \cdot \varrho_2) \cdot \Sigma_x \Sigma_y$   
approximation is exact only if the convolution shape is  
factorizable:  $(\varrho_1 \cdot \varrho_2)(x, y) = f_1(x) \cdot f_2(y) \rightarrow$  we have to test it!



- Measurements: convolution shape not factorizable → **XY factorization bias**
- 2D fit using multiple functions: **Single Gaussian, super Gaussian, Double Gaussian** (with various constraints), **q-Gaussian, polynomial Gaussian**

# Input Data

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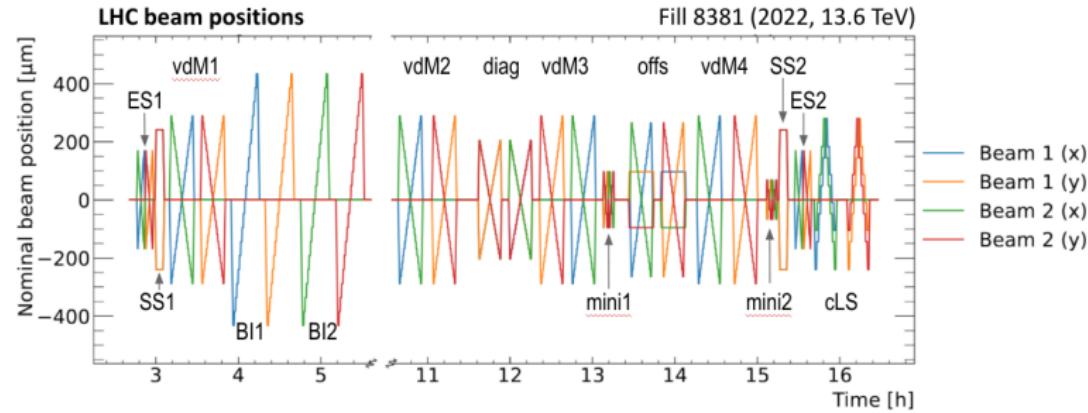
2D Fits

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vdM scans used in this analysis:

- vdM(2-4)
- offset
- diagonal
- mini(1-2)

# 2D Fits: Visualization

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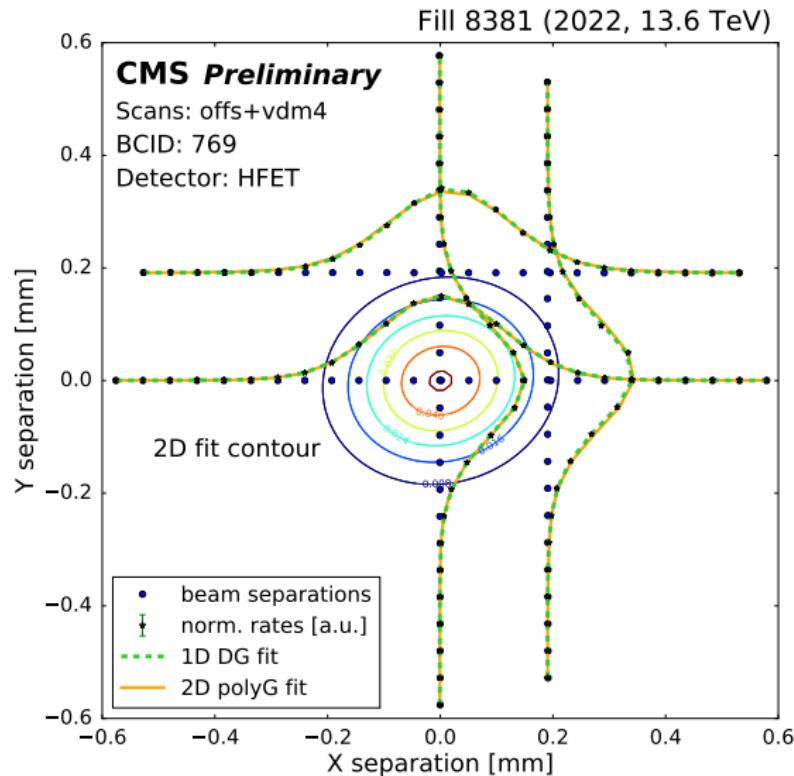
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# Simulation

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- Given: 2D fit function parameters with uncertainties from measured data (from vdM+diag / vdM+mini / vdM+offs) → sample parameter space to get vdM “data”
- Vary these parameters according to the uncertainty → random 2D shape (this also lets the uncertainty of the 2D fit propagate onto the correction)
- Calculate rate at pre-defined (nominal) separation points using the randomized shape
- Do (hypothetical) vdM fit (1D) on this artificial data
- Compare  $\sigma_{vis}$  from the simulated vdM with the exact  $\sigma_{vis}$  of the randomized 2D shape → get a factorisation correction
- Standard deviation over many randomized shapes → fit uncertainty
- Method cross-checked with closure test

# Correlation Between Detectors

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Introduction

2D Fits

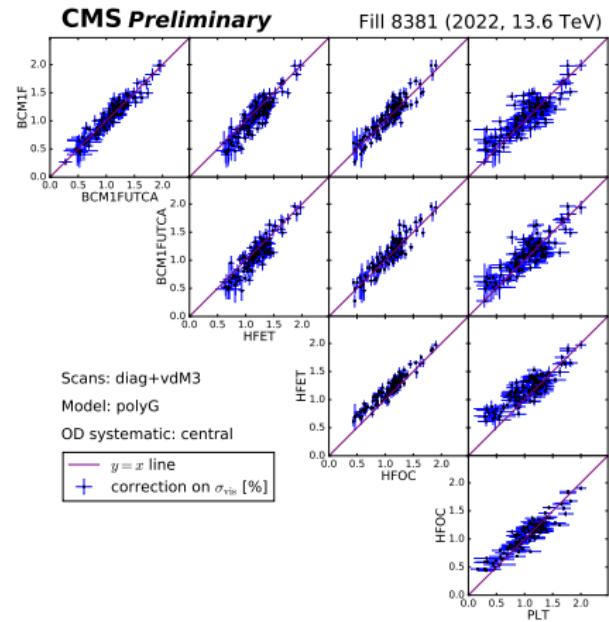
Simulation

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Conclusion

- Strong correlations in case of a stable shape model and well-understood luminometers
- These luminometers are used to define the final correction and its uncertainty



# Correlation Between Models

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2D Fits

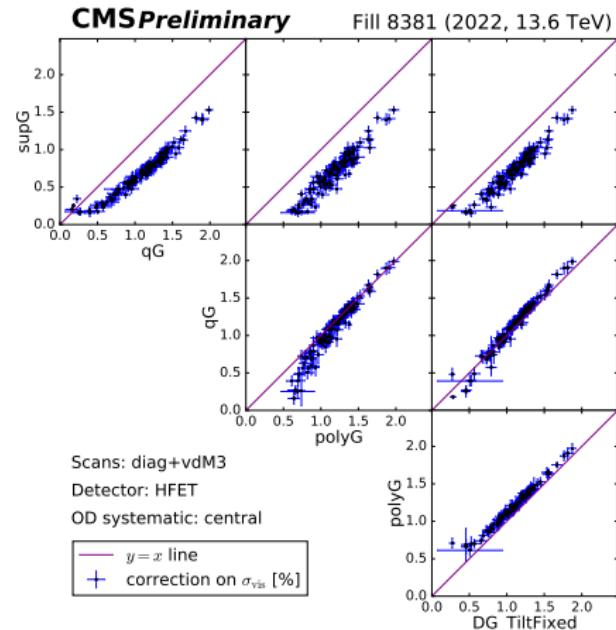
Simulation

Choosing the Right  
Model(s)

Final Result

Conclusion

- Strong correlations
- supG: smallest corrections
- These models are used to define the final correction and its uncertainty



# Model Dependence of Results

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2D Fits

Simulation

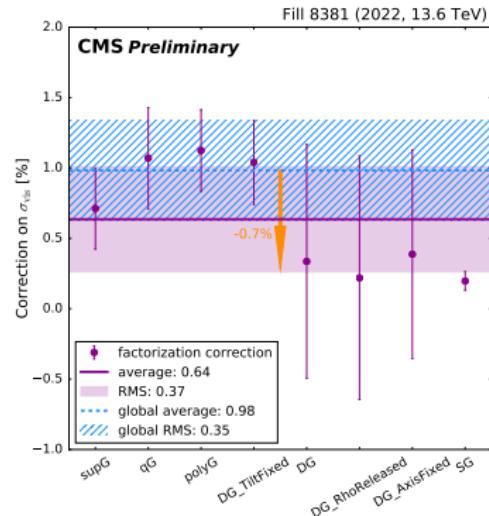
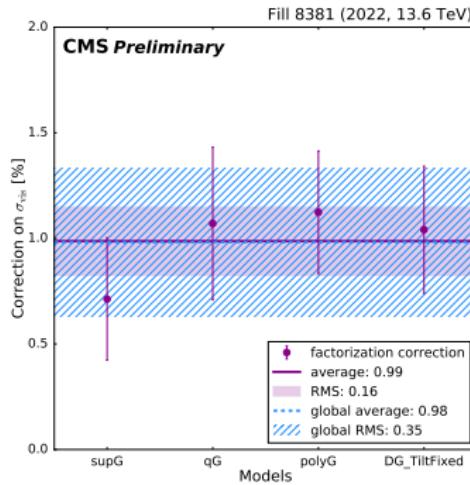
Choosing the Right  
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Comparison of the results from the chosen vs other models:

- Global average out of the results of the stable models
- Additional model systematics of 0.7% accounts for the results of the 3 unstable DG models



# Checking Time Dependence

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2D Fits

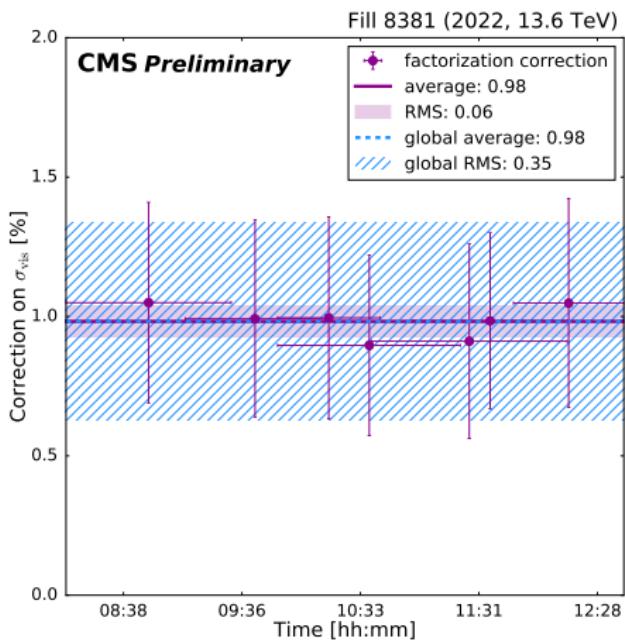
Simulation

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- No time dependence during these measurements
- Points refer to:  
vdM2+diag,  
diag+vdM3,  
vdM3+mini1,  
vdM3+offs,  
mini1+vdM4,  
offs+vdM4,  
vdM4+mini2  
combinations



# BCID Dependence of Corrections

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2D Fits

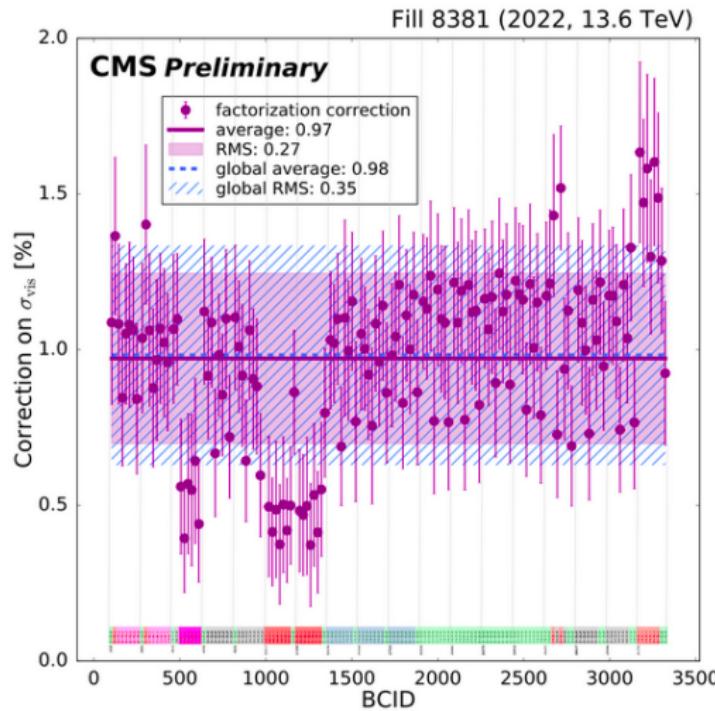
Simulation

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Conclusion

- BCID dependence gives the vast majority of the RMS
- Pattern corresponds to the collision scheme in LHC (i.e. in which experiments the participating bunches collide)



# Summary and Outlook

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The final result:  $(0.98 \pm 0.35 \pm 0.7)\% \rightarrow (1.0 \pm 0.8)\%$

- Global RMS: 0.35% (from all  $144 \times 4 \times (9 \text{ or } 10) \times 7 \times 5$  measurements)
- Shape uncertainty: 0.7%
- Closure test: negligible

Uncertainties in more detail:

- BCID (144): 0.27% (true dependence: collision pattern at LHC)
- Model (only the 4 well-behaving): 0.16%
  - For all 8: 0.37%
  - Additional model uncertainty to cover the lack of knowledge of true shape: 0.7%
- OD systematics (10 for offset + 9 for diag): 0.05%
- Time (aka. scan pairing, 7 combinations): 0.05%
- Luminometer (5): 0.02%

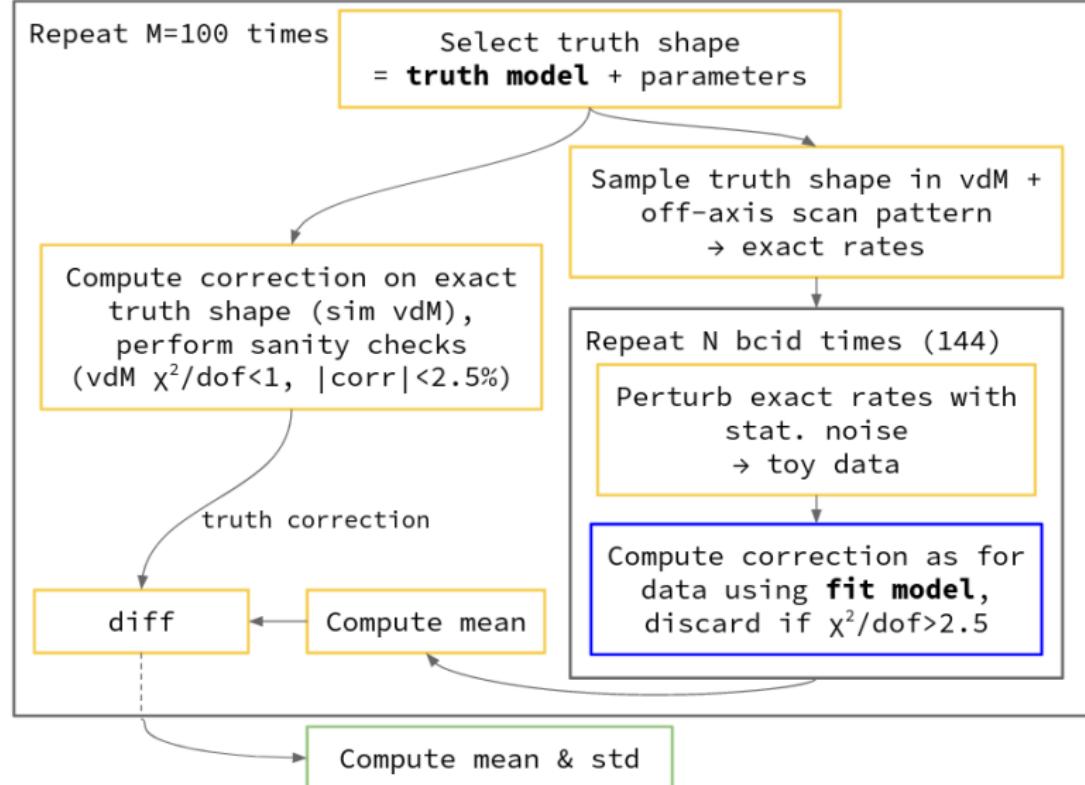
# Thank You for Your Time!

[Backups »](#)

# Workline

- 1D fits
- Rate matching for offset → more precise OD correction
- $\langle \text{rate matching vs lin OD correction separation} \rangle = 0.003 \text{ mm}$   
(used as systematic variation)
- 2D fits - further tunes of DG shape considered:
  - RhoFixed: fitting with  $\varrho = 0$  requirement
  - RhoReleased: on the basis of RhoFixed results, refit releasing  $\varrho$
  - Reparametrizations: aiming to avoid oscillation between 2 minima in the prediction of the correction in DG case
    - TiltFixed: fixing the sigmas in the ≈same directions between the 2 DGs (in 2D) to be the same (X and Y direction still can be different), also fixing  $d\varrho = 0$
    - AxisFixed: fixing the angle between the axes of the DG's SG components to 0 (requiring having the same orientation)
- Simulation
- Closure test (next slide)

# Closure Test

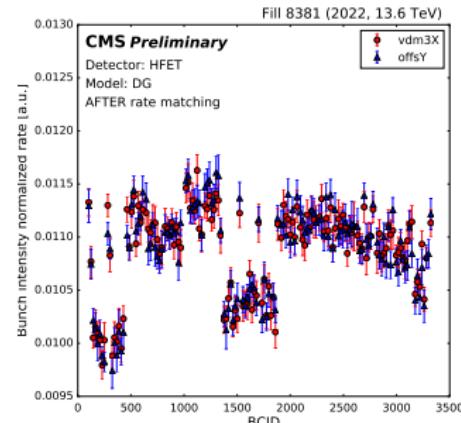
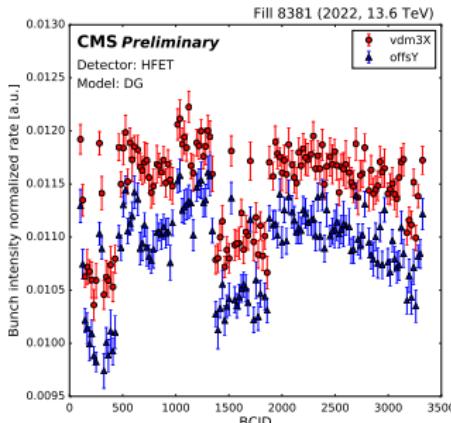
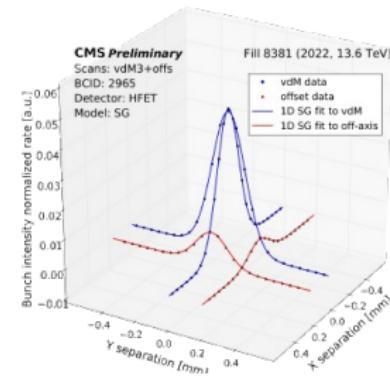


# Rate Matching

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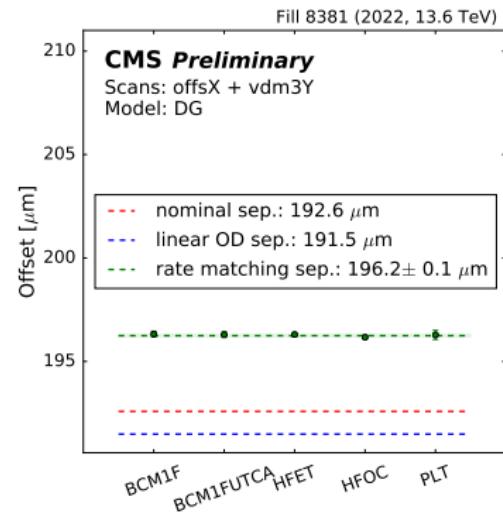
Additional Orbit  
Drift Correction

Fast change of the rate with separation → offset scans: even a small change in the separation can have a significant effect → additional correction on offset data ensuring that rates match at the “common” scan point:



# Additional Orbit Drift Correction

- Linear orbit-drift (OD) correction applied beforehand on input data (based on beam position measurements by DOROS: *Diode ORbit and OScillation system*)
- Rate matching  $\approx$  additional OD correction in the non-scanning direction of the offset scan
  - Largest shift shown
  - Luminometers agree well
  - Value also used as a  $\pm$  variation estimating possible unknown OD



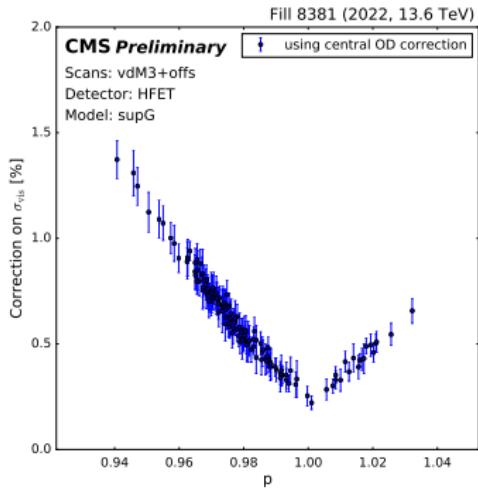
# Parameter Dependence

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Additional Orbit  
Drift Correction

$$supG_{2D}(x, y; V, \mu_x, \mu_y, \sigma_x, \sigma_y, \varrho, p) = \frac{V}{N} \exp\left(-\frac{r^{2p}}{2}\right)$$

- $p$  parameter of supG  
(responsible for the  
"peakiness" of the supG  
shape:  $p > 1$  means flatter  
than Gaussian shape)  
divides obtained correction  
value to 2 trends
- Phenomena present in  
other models as well (with  
the corresponding  
parameters)

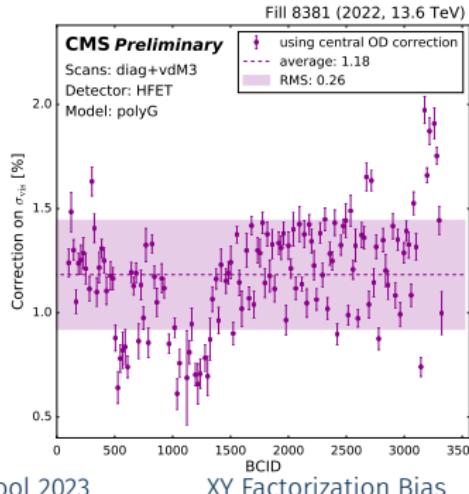
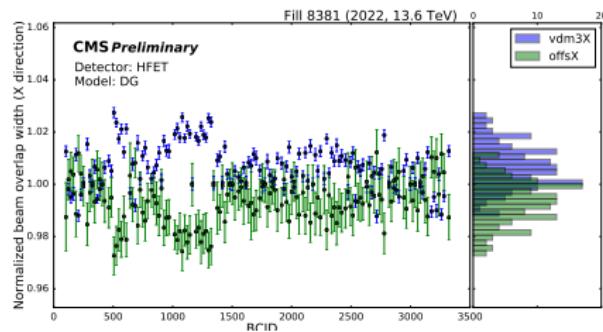


# BCID-Dependence Pattern

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Additional Orbit  
Drift Correction

- Plotting the correction values wrt. the BCID: same pattern present as in 1D
- Pattern corresponds to the collision scheme in LHC (i.e. in which experiments the participating bunches collide)



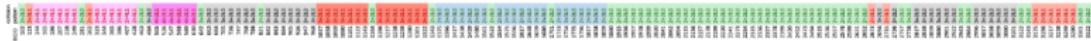
# LHC Collision Pattern

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Additional Orbit  
Drift Correction

- 2+2L0:** bunch in beam 1 and bunch beam 2 collides only at IP1 and IP5
- 3+2L0:** bunch in beam 1 also collides at IP2 (ALICE)
- 3+2L1:** bunch in beam 1 also collides at IP8 (LHCb)
- 2+3L0:** bunch in beam 2 also collides at IP2 (ALICE)
- 2+3L1:** bunch in beam 2 also collides at IP8 (LHCb)
- 4+2L1:** bunch in beam 1 collides at all for IPs
- 2+4L1:** bunch in beam 2 collides at all for IPs

LHC filling scheme: [525ns\\_146b\\_144\\_35\\_22\\_8bpl\\_20inj\\_nocloseLR.csv](#)



# 2D Functions in Use I/IV.

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Additional Orbit  
Drift Correction

Short-hand notation:  $r^2 = \frac{\left(\frac{x-\mu_x}{\sigma_x}\right)^2 + \left(\frac{y-\mu_y}{\sigma_y}\right)^2 - 2\rho \frac{x-\mu_x}{\sigma_x} \frac{y-\mu_y}{\sigma_y}}{1-\rho^2}$

2D fitting functions used in the analyses:

- Single Gaussian:

$$SG_{2D}(x, y; V, \mu_x, \mu_y, \sigma_x, \sigma_y, \rho) = \frac{V}{2\pi\sigma_x\sigma_y\sqrt{1-\rho^2}} \exp\left(-\frac{r^2}{2}\right)$$

- Super Gaussian:

$$supG_{2D}(x, y; V, \mu_x, \mu_y, \sigma_x, \sigma_y, \rho, p) = \frac{V}{N} \exp\left(-\frac{r^{2p}}{2}\right)$$

• Where the normalization:  $N = 2\pi\sigma_x\sigma_y\sqrt{1-\rho^2} \cdot \frac{\Gamma(1/p)}{p} 2^{1/p}$

- ... (next slides)

## 2D Functions in Use II/IV.

- ... (previous slide)
- q-Gaussian:  $qG_{2D}(x, y) = \frac{V}{C(q) \cdot \sigma_x \sigma_y \sqrt{1-\varrho^2}} \cdot e_q\left(-\frac{r^2}{(4-2q)}\right)$ 
  - Where:  $e_q(x) = [1 + (1 - q)x]_+^{\frac{1}{1-q}}$
  - And:  $C(q) = \begin{cases} \frac{4-2q}{1-q} \pi \frac{\Gamma(\frac{2-q}{1-q})}{\Gamma(\frac{2-q}{1-q} + 1)} & q < 1 \\ \frac{4-2q}{q-1} \pi \frac{\Gamma(\frac{1}{q-1}-1)}{\Gamma(\frac{1}{q-1})} & \text{otherwise.} \end{cases}$
- Polynomial Gaussian:  $\text{poly}G_{2D}(x, y; V, \mu_x, \mu_y, \sigma_x, \sigma_y, \varrho, a_2, \bar{a}_4) = \frac{V}{N} (1 + a_2 r^2 + a_4 r^4) \cdot \exp(-0.5r^2)$ 
  - Where the normalization:  $N = 2\pi\sigma_x\sigma_y\sqrt{1-\varrho^2} \cdot (1 + 2a_2 + 8a_4)$
  - And the other parameters:  $a_4 = \begin{cases} \bar{a}_4 & \text{if } a_2 > 0 \\ \bar{a}_4 + a_2^2/4 & \text{otherwise.} \end{cases}$
  - For positivity:  $\bar{a}_4 > 0$  is required
- ... (next slides)

## 2D Functions in Use III/IV.

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Additional Orbit  
Drift Correction

- ... (previous slides)

- Double Gaussian:

$$DG_{2D}(x, y) = V \cdot \left( v_R \cdot SG_{2D}\left(x, y; 1, \mu_x, \mu_y, \sigma_x^{(1)}, \sigma_y^{(1)}, \varrho^{(1)}\right) + (1 - v_R) \cdot SG_{2D}\left(x, y; 1, \mu_x, \mu_y, \sigma_x^{(2)}, \sigma_y^{(2)}, \varrho^{(2)}\right) \right)$$

- Reparametrizations of the DG 2D fitting function:

$$\begin{aligned} \cdot \quad & DG\_TiltFixed(x, y) = V \cdot \left( v_R \cdot SG_{2D}\left(x, y; 1, \mu_x, \mu_y, \sigma_x^{(1)}, \sigma_y^{(1)}, \varrho + d\varrho\right) + (1 - v_R) \cdot SG_{2D}\left(x, y; 1, \mu_x, \mu_y, \sigma_x^{(2)}, \sigma_y^{(2)}, \varrho - d\varrho\right) \right) \end{aligned}$$

$$\begin{aligned} \cdot \quad & \text{Where: } \sigma_x^{(1)} = \sigma_x \sqrt{s_R} \sqrt{s_A}, \quad \sigma_y^{(1)} = \sigma_y \sqrt{s_R} / \sqrt{s_A}, \\ & \sigma_x^{(2)} = \sigma_x / \sqrt{s_R} / \sqrt{s_A}, \quad \sigma_y^{(2)} = \sigma_y / \sqrt{s_R} \sqrt{s_A} \end{aligned}$$

- ... (next slide)

## 2D Functions in Use IV/IV.

- ... (previous slides)
- Another reparametrization of the DG 2D fitting function also needs for reparametrization of the 2D single Gaussian function:

$$SG_{2D}^{(\alpha)}(x, y; V, \mu_x, \mu_y, \sigma_x, \sigma_y, \alpha) = \frac{V}{2\pi\sigma_x\sigma_y\sqrt{1-\varrho^2}} \cdot \exp\left(-\frac{r_\alpha^2}{2}\right)$$

- Where:  $r_\alpha^2 = \begin{pmatrix} x - \mu_x \\ y - \mu_y \end{pmatrix}^T R(-\alpha) \begin{pmatrix} \sigma_x^{-2} & 0 \\ 0 & \sigma_y^{-2} \end{pmatrix} R(\alpha) \begin{pmatrix} x - \mu_x \\ y - \mu_y \end{pmatrix}$ 
  - Where:  $R(\alpha)$  is a  $2 \times 2$  rotation matrix

- With these the reparametrized DG:

$$DG\_AxisFixed(x, y) = V_R \cdot SG_{2D}^{(\alpha)}(x, y; 1, \mu_x, \mu_y, \sigma_x^{(1)}, \sigma_y^{(1)}, \alpha + d\alpha) + (1 - V_R) \cdot SG_{2D}^{(\alpha)}(x, y; 1, \mu_x, \mu_y, \sigma_x^{(2)}, \sigma_y^{(2)}, \alpha - d\alpha)$$