



# Early Measurements- Beauty Hadron Cross-sections

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**BEAUTY2CHARM**  
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Broad aim: We want to measure beauty production cross-sections with hadronic final states, e.g.:

$$\sigma(pp \rightarrow B^{\pm} X) = \frac{N^{\pm}}{\mathcal{L} \times \mathcal{B}(B^{\pm} \rightarrow D^0 \pi^{\pm}) \times \mathcal{B}(D^0 \rightarrow K^{\pm} \pi^{\mp}) \times \epsilon_{tot}}$$

- ▶  $N^{\pm}$  is the total number of signal events
- ▶  $\mathcal{L}$  is the integrated luminosity
- ▶  $\mathcal{B}(B^{\pm} \rightarrow D^0 \pi^{\pm})$  and  $\mathcal{B}(D^0 \rightarrow K^{\pm} \pi^{\mp})$  are branching ratios
- ▶  $\epsilon_{tot}$  is the total efficiency

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The total efficiency has a number of inputs:

$$\epsilon_{tot} = \epsilon_{acc} \times R_{PID} \times R_{track+reco} \times \epsilon_{HLT1} \times \epsilon_{HLT2} \times \epsilon_{sel} \times \epsilon_{GEC}$$

- ▶  $\epsilon_{acc}$  is the fraction of signal with all final-state particles within the fiducial region of the detector acceptance
- ▶  $R_{PID}$  is ratio of PID efficiencies of data compared to MC
- ▶  $R_{track+reco}$  is the ratio of tracking efficiencies of data compared to MC
- ▶  $\epsilon_{HLT1}$  is the efficiency of HLT1
- ▶  $\epsilon_{HLT2}$  is the efficiency of HLT2
- ▶  $\epsilon_{sel}$  is the efficiency of the offline selection
- ▶  $\epsilon_{GEC}$  is the Global Event Cut efficiency

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# Calculating Acceptance (Generator) Efficiencies

- ▶ Acceptance efficiency is the proportion of events that lie within LHCb's detector acceptance.
- ▶ These requirements, known as *Daughters in LHCb* or (*DecProdCut*) are:
  - ▶ Daughter particles in range  $0.01 \text{ (0.05)} < \theta < 0.4$  for charged (neutral) particles
  - ▶ All daughters moving in the same direction (product of PZ of pairs of daughters is positive)
- ▶ To do this, we want to simulate a load of particle events (not applying this cut) and then count how many obey the requirements
- ▶ LHCb simulation software is known as *Gauss*
- ▶ Example options found [here](#)



# Removing DecProdCut

- ▶ Tutorials on general use of Gauss can be found [here](#)
- ▶ Go to Gen/DecFile/dkfiles and copy appropriate .dec file

```
## EventType: 11264011
#
# Descriptor: {[[B0]nos -> (D- => K+ pi- pi-) K+]cc, [[B0]os -> (D+ => K- pi+ pi+) K-]cc}
#
# NickName: Bd_D-K+
##
# Cuts: None
##
# Documentation: Includes resonances in D- decay
# EndDocumentation
#
# PhysicsWG: B20C
# Tested: Yes
# Responsible: Nicola Serra
# Email: Nicola.Serra@cern.ch
# Date: 20100507
```

Decfiles package explained further [here](#)





- Go to Gen/DecFile/options and copy {PARTICLE ID}.py

```
# Event Type: 11264011
#
# ASCII decay Descriptor: [[[B0]nos -> (D- => K+ pi- pi-) K+]]cc, [[[B0]os -> (D+ => K- pi+ pi+) K-]]cc]
#
from Configurables import Generation
Generation().EventType = 11264011
Generation().SampleGenerationTool = "SignalRepeatedHadronization"
from Configurables import SignalRepeatedHadronization
Generation().addTool( SignalRepeatedHadronization )
Generation().SignalRepeatedHadronization.ProductionTool = "Pythia8Production"
from Configurables import ToolSvc
from Configurables import EvtGenDecay
ToolSvc().addTool( EvtGenDecay )
#ToolSvc().EvtGenDecay.UserDecayFile = "$DECFILESROOT/dkfiles/Bd_D-K+=DecProdCut.dec"
ToolSvc().EvtGenDecay.UserDecayFile = "$DECFILESROOT/dkfiles/Bd_D-K+.dec"
#Generation().SignalRepeatedHadronization.CutTool = "DaughtersInLHCb"
Generation().SignalRepeatedHadronization.SignalPIDList = [ 511,-511 ]
```

# Inputs: Acceptance (Generator) Efficiency

- ▶ Run with options file like this:

```
from Gauss.Configuration import *

#importOptions('$APPCONFIGOPTS/Gauss/Beam6500GeV-md100-nu1.6.py')
#importOptions('$APPCONFIGOPTS/Gauss/Beam7000GeV-md100-nu7.6-HorExtAngle.py')
#importOptions('$GAUSSOPTS/GenStandAlone.py')
#importOptions('$DECFILESROOT/options/11264001.py')
#importOptions('$LBPYTHIA8ROOT/options/Pythia8.py')

importOptions('$APPCONFIGOPTS/Gauss/Beam7000GeV-md100-nu7.6-HorExtAngle.py')
importOptions('$GAUSSOPTS/GenStandAlone.py')
#importOptions('$APPCONFIGOPTS/Gauss/EnableSpillover-25ns.ov')
importOptions('$DECFILESROOT/options/11264011_noDecProdCut.py')
importOptions('$LBPYTHIA8ROOT/options/Pythia8.py')
importOptions('$APPCONFIGOPTS/Gauss/Gauss-Upgrade-Baseline-20150522.py')
importOptions('$APPCONFIGOPTS/Gauss/G4PL_FTFP_BERT_EmNoCuts.py')

GaussGen = GenInit("GaussGen")
GaussGen.FirstEventNumber = 1
GaussGen.RunNumber = 231

Generation().SignalPlain.CutTool = ""

from Configurables import LHCbApp
#LHCbApp().DDDBtag = 'dddb-20170721-3'
#LHCbApp().CondDBtag = 'sim-20170721-2-vc-md100'
LHCbApp().DDDBtag = 'dddb-20201211'
LHCbApp().CondDBtag = 'sim-20201218-vc-md100'
LHCbApp().EvtMax = 100
```

# Calculating Generator Efficiencies

- ▶ Run Gauss with the previous options
- ▶ Convert output xgen to ntuple (using options file such as [this](#))
- ▶ Apply DecProdCut selections in analysis script

```
import ROOT
f= ROOT.TFile.Open({ROOT_FILE}, "read")
t = f.Get("{}MCDecayTree".format({TUPLE_NAME}))

num = t.GetEntries('Kplus_THETA > 0.01 && Kplus_THETA < 0.4 && '/
'piminus_THETA > 0.01 && piminus_THETA < 0.4 && piminus0_THETA > 0.01 '/
'&& piminus0_THETA < 0.4 && Kplus0_THETA > 0.01 && Kplus0_THETA < 0.4 '/
'&& Kplus_PZ * piminus_PZ > 0 && Kplus_PZ * piminus0_PZ > 0 && '/
'Kplus_PZ * Kplus0_PZ > 0')

denom = t.GetEntries('')
```





Our procedure:

- ▶ MC simulation of our signal modes have been made and can be found on the bookkeeping at '/MC/Upgrade/Beam7000GeV-Upgrade-MagDown-Nu7.6-25ns-Pythia8/Sim10-Up08/11264001,11264011,12163001,13264021,15364010/XDIGI'
- ▶ Obtain "raw" sample by ntupling straight from xdigi. Example DaVinci options [here](#)
- ▶ Run HLT1 in *passthrough* followed by HLT2 (both in [Moore](#)) and make ntuple [DaVinci](#)
- ▶ Can apply HLT1 decisions in analysis script after ntupling
- ▶ We find HLT1 efficiency *relative to HLT2*- TOS efficiencies for HLT1 and HLT2 now possible with HltEfficiencyChecker [link](#)

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# HLT1 in Passthrough

HLT1 Passthrough line exists in default Allen configuration and HLT1 will run in passthrough by default.

Moore tutorial found [here](#)

Example options files and instructions on how to run can be found [here](#)



# Persisting Trigger Decisions

Moore/Hlt/Hlt2Conf/options/hlt2\_integration\_test.py:

```
with reconstruction.bind(from_file=False):  
    config = run_moore(options, make_lines, public_tools, allen_hlt1=True)  
    dump_hlt2_configuration(config, "hlt2_integration.tck.json")
```

ntuple\_options\_<mode>\_HLT1\_passthrough.py:

```
def get_hlt2_unpackers(is_simulation):  
    """Configures algorithms for reading HLT2 output.  
  
    This is a temporary measure until support for Run 3 HLT2 output is added to  
    an LHCb application.  
    """  
    unpacker = LHCb__UnpackRawEvent(  
        'UnpackRawEvent',  
        BankTypes=['ODIN'],  
        RawBankLocations=['/Event/DAQ/RawBanks/ODIN'])  
  
    dec_reports_hlt1 = HltDecReportsDecoder("Hlt1",SourceID="Hlt1",OutputHltDecReportsLocation="/Event/Hlt1/DecReports",OutputLevel=1)  
    dec_reports = HltDecReportsDecoder(  
        SourceID="Hlt2", OutputHltDecReportsLocation="/Event/Hlt2/DecReports")  
    reading_algs = ([reading.decoder()] + reading.unpackers() + [dec_reports_hlt1] + [dec_reports] + [unpacker,createODIN()])  
    if is_simulation:  
        reading_algs = reading.mc_unpackers() + reading_algs  
    return reading_algs
```

# Persisting Trigger Decisions

ntuple\_options\_<mode>\_HLT1\_passthrough.py::

```
def configure_packed_locations(tck_location):
    """Configures HltANNSvc to know about packed locations and hlt2 decision names used in Moore.

    tck_location (string): Location of json file containing trigger configuration.

    """

    with open(tck_location) as f:
        tck = json.load(f)
    ann_config = tck["HltANNSvc/HltANNSvc"]
    HltANNSvc(PackedObjectLocations={
        str(k): v
        for k, v in ann_config["PackedObjectLocations"].items()
    })
    HltANNSvc(Hlt1SelectionID={
        str(k): v
        for k, v in ann_config["Hlt1SelectionID"].items()
    })
    HltANNSvc(Hlt2SelectionID={
        str(k): v
        for k, v in ann_config["Hlt2SelectionID"].items()
    })
```



# Persisting Trigger Decisions

tuple\_options\_<mode>\_HLT1\_passthrough.py::

```
TriggerTool = dtt.addTupleTool("TupleToolTrigger/TriggerTool")
TriggerTool.VerboseHlt1 = True
TriggerTool.VerboseHlt2 = True
TriggerTool.FillHlt1=True
TriggerTool.FillHlt2=True
TriggerTool.Verbose=True
TriggerTool.Hlt1DecReports = "/Event/Hlt1/DecReports"
TriggerTool.Hlt2DecReports = "/Event/Hlt2/DecReports"
TriggerTool.TriggerList = [{list_of_trigger_lines_to_be_persisted}]
```

```
# Load the 'TCK's dumped from the HLT2 Moore job
configure_packed_locations("hlt2_integration_minbias.tck.json")

# Configure the unpacking of data (we assume we want MC information)
# and the running of the user algorithms. The order is important.
ApplicationMgr().TopAlg = get_hlt2_unpackers(is_simulation=True) + user_algs
```



- ▶ Recently held up by a segmentation violation in Allen (see [here](#))
- ▶ This was eventually fixed by [Allen!775](#)
- ▶ Inclusion of RetinaClusters will becoming new default for xdigi files in order to run successfully with Moore.
  - ▶ RetinaClusters can be added to xdigi files following instructions [here](#).
  - ▶ Explanation of how to run without adding RetinaClusters to be added to Allen documentation.

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# Estimating Signal Yields Example

- ▶ Binned differential cross-sections for  $pp \rightarrow B^+ X$  taken from [LHCb-ANA-2016-005](#)
- ▶ Assume  $\mathcal{L} = 100 \text{ pb}^{-1}$  and multiply by this
- ▶ Multiply by  $\mathcal{B}(B^\pm \rightarrow D^0 \pi^\pm) = (4.68 \pm 0.13) \times 10^{-3}$  and  $\mathcal{B}(D^0 \rightarrow K^\pm \pi^\mp) = (3.95 \pm 0.031) \times 10^{-2}$
- ▶ Multiply by binned acceptance and HLT efficiencies

# Estimated $B^+ \rightarrow (D^0 \rightarrow K^+ \pi^-) \pi^+$ Signal Yields

$p_T$ (GeV/c)	$2.0 < y < 2.5$	$2.5 < y < 3.0$	$3.0 < y < 3.5$	$3.5 < y < 4.0$	$4.0 < y < 4.5$
0.0 - 0.5	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan
0.5 - 1.0	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan
1.0 - 1.5	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan	nan $\pm$ nan
1.5 - 2.0	nan $\pm$ nan	nan $\pm$ nan	0.0 $\pm$ 0.0	164.0 $\pm$ 83.0	66.0 $\pm$ 47.0
2.0 - 2.5	nan $\pm$ nan	169.0 $\pm$ 99.0	640.0 $\pm$ 173.0	833.0 $\pm$ 186.0	818.0 $\pm$ 179.0
2.5 - 3.0	nan $\pm$ nan	152.0 $\pm$ 88.0	1386.0 $\pm$ 261.0	2351.0 $\pm$ 354.0	1312.0 $\pm$ 235.0
3.0 - 3.5	nan $\pm$ nan	450.0 $\pm$ 151.0	2307.0 $\pm$ 355.0	3616.0 $\pm$ 453.0	1907.0 $\pm$ 303.0
3.5 - 4.0	nan $\pm$ nan	980.0 $\pm$ 235.0	3196.0 $\pm$ 426.0	3066.0 $\pm$ 400.0	1817.0 $\pm$ 295.0
4.0 - 4.5	nan $\pm$ nan	1430.0 $\pm$ 266.0	3088.0 $\pm$ 408.0	3767.0 $\pm$ 463.0	2132.0 $\pm$ 324.0
4.5 - 5.0	0.0 $\pm$ 0.0	2362.0 $\pm$ 346.0	3444.0 $\pm$ 429.0	2818.0 $\pm$ 380.0	1909.0 $\pm$ 306.0
5.0 - 5.5	51.0 $\pm$ 51.0	2132.0 $\pm$ 327.0	4486.0 $\pm$ 483.0	3518.0 $\pm$ 423.0	1855.0 $\pm$ 291.0
5.5 - 6.0	341.0 $\pm$ 123.0	2144.0 $\pm$ 326.0	5097.0 $\pm$ 521.0	2940.0 $\pm$ 389.0	1883.0 $\pm$ 299.0
6.0 - 6.5	124.0 $\pm$ 72.0	2805.0 $\pm$ 371.0	3807.0 $\pm$ 430.0	2283.0 $\pm$ 317.0	1246.0 $\pm$ 233.0
6.5 - 7.0	512.0 $\pm$ 142.0	2563.0 $\pm$ 346.0	3444.0 $\pm$ 402.0	1568.0 $\pm$ 250.0	1165.0 $\pm$ 234.0
7.0 - 7.5	368.0 $\pm$ 125.0	2632.0 $\pm$ 343.0	2720.0 $\pm$ 334.0	2832.0 $\pm$ 349.0	947.0 $\pm$ 192.0
7.5 - 8.0	386.0 $\pm$ 120.0	2095.0 $\pm$ 312.0	1770.0 $\pm$ 263.0	1557.0 $\pm$ 238.0	785.0 $\pm$ 160.0
8.0 - 8.5	605.0 $\pm$ 172.0	2258.0 $\pm$ 310.0	2412.0 $\pm$ 322.0	1516.0 $\pm$ 244.0	852.0 $\pm$ 186.0
8.5 - 9.0	488.0 $\pm$ 150.0	1781.0 $\pm$ 264.0	1967.0 $\pm$ 274.0	1315.0 $\pm$ 225.0	553.0 $\pm$ 141.0
9.0 - 9.5	281.0 $\pm$ 101.0	1499.0 $\pm$ 236.0	1477.0 $\pm$ 254.0	1839.0 $\pm$ 270.0	811.0 $\pm$ 178.0
9.5 - 10.0	401.0 $\pm$ 123.0	1092.0 $\pm$ 203.0	1211.0 $\pm$ 218.0	1072.0 $\pm$ 200.0	558.0 $\pm$ 148.0
10.0 - 10.5	659.0 $\pm$ 169.0	1678.0 $\pm$ 265.0	874.0 $\pm$ 187.0	1213.0 $\pm$ 212.0	366.0 $\pm$ 113.0
10.5 - 11.5	383.0 $\pm$ 81.0	1288.0 $\pm$ 165.0	920.0 $\pm$ 130.0	556.0 $\pm$ 100.0	504.0 $\pm$ 97.0
11.5 - 12.5	388.0 $\pm$ 88.0	686.0 $\pm$ 119.0	856.0 $\pm$ 128.0	459.0 $\pm$ 86.0	246.0 $\pm$ 72.0
12.5 - 14.0	287.0 $\pm$ 60.0	637.0 $\pm$ 89.0	476.0 $\pm$ 79.0	407.0 $\pm$ 69.0	185.0 $\pm$ 46.0
14.0 - 16.5	226.0 $\pm$ 43.0	403.0 $\pm$ 54.0	416.0 $\pm$ 58.0	295.0 $\pm$ 45.0	133.0 $\pm$ 29.0
16.5 - 23.5	129.0 $\pm$ 20.0	165.0 $\pm$ 21.0	127.0 $\pm$ 18.0	85.0 $\pm$ 15.0	40.0 $\pm$ 11.0
23.5 - 40.0	22.0 $\pm$ 5.0	25.0 $\pm$ 5.0	15.0 $\pm$ 4.0	13.0 $\pm$ 4.0	3.0 $\pm$ 2.0





Will have multiple PID (DLL) cuts on each final state track

- ▶ Multiple options explored to determine efficiencies of these cuts
  - ▶ PIDGen2/PIDCorr not suitable as control of systematics associated with kernel width/templating less reliable
  - ▶ Cross-sections will be systematically limited
- ▶ Move forward with PIDCalib2

Investigated several options within PIDCalib2

- ▶ Would rather not have to re-create HLT2 lines and selections with PID cuts removed
  - ▶ **Correction histograms** that give the ratio between data and MC PID response to correct MC after PID cuts have been made
- ▶ Cross-section measurements are performed in bins of  $p_T$  and  $\eta$ 
  - ▶ Instead of using sWeighted calibration data (nominal PIDCalib2 method) do separate fits in bins of  $p_T$  and  $\eta$  - simple **cut and count method**
  - ▶ Removes problem that signal and background shapes in the calibration samples are correlated with kinematics of the track = no systematic for sWeighting

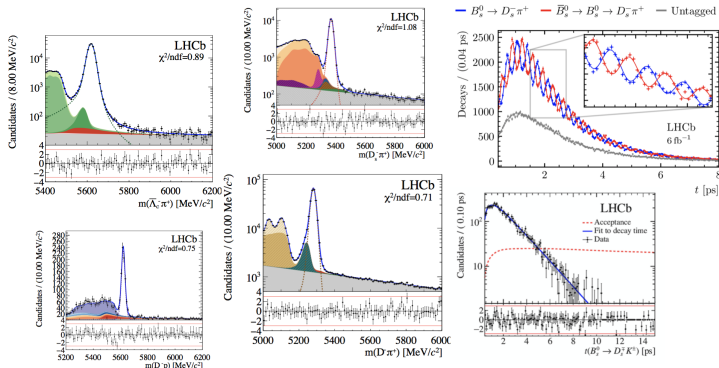
# HLT2 Line Update

- Bs2DsPi
  - LAZY\_AND: Hlt2B2OC\_BdToDsmPi\_DsmToKpKmPim\_Line #=150000 Sum=8
  - $1e+6 * 8./150000. = 53.33 \text{ Hz}$
- Bd2DPi
  - LAZY\_AND: Hlt2B2OC\_BdToDmPi\_DmToPimPimKp\_Line #=150000 Sum=11
  - $1e+6 * 11./150000. = 73.33 \text{ Hz}$
- Bd2DK
  - LAZY\_AND: Hlt2B2OC\_BdToDmK\_DmToPimPimKp\_Line #=150000 Sum=10
  - $1e+6 * 10./150000. = 66.66 \text{ Hz}$
- Lb2LcPi
  - LAZY\_AND: Hlt2B2OC\_LbToLcpPi\_LcpToPKPi\_Line #=150000 Sum=6
  - $1e+6 * 6./150000. = 40.00 \text{ Hz}$
- Bu2D0Pi
  - LAZY\_AND: Hlt2B2OC\_BuToD0Pi\_D0ToHH\_Line #=150000 Sum=11
  - $1e+6 * 11./150000. = 73.33 \text{ Hz}$

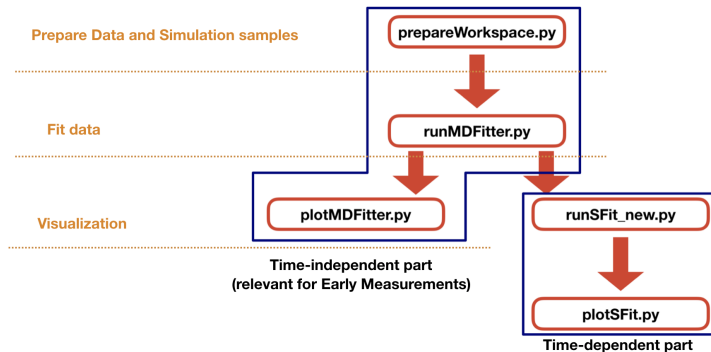
Rates calculated using 150,000 input HLT1 filtered minbias events.

All lines well below 100 Hz.

- ▶ We are using **PhysFit/B2DXFitters** package (a part of Urania), a common ROOT based tool to study  $B \rightarrow DX$  transitions.
- ▶ Verified during Run1 and Run2 in +10 physics analyses.
- ▶ Setup instructions: [link](#), MDFit tutorial: [link](#).



- ▶ Small extensions needed for effective fitting in  $\eta$ ,  $p_T$  bins.
- ▶ All modes:  $B^+ \rightarrow D^0 \pi$ ,  $B^0 \rightarrow D^- \pi^+ / K$ ,  $B_s^0 \rightarrow D_s^- \pi^+$ ,  $\Lambda_b^0 \rightarrow \Lambda_c^+ \pi^-$  available within one package sharing most of the code.
- ▶ User side: setting up configuration files.
- ▶ Basic workflow:

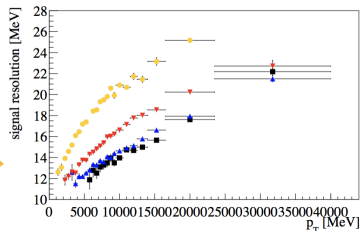
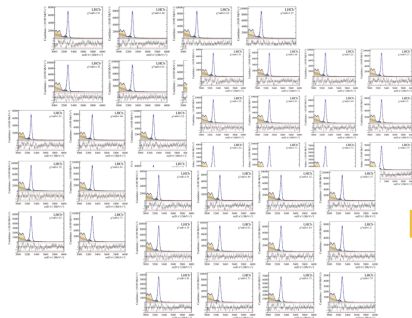


# What are we doing now?

Two ways for Run3 data preparation:

- ▶ study Run2 data,
- ▶ perform large scale toys.

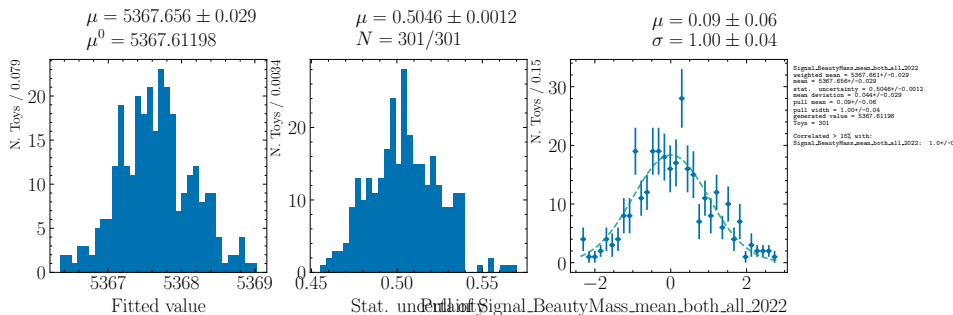
Study Run2 data: signal resolution dependence in  $\eta$ ,  $p_T$ .



# What are we doing now?

Large scale toys studies.

- ▶ Generate and fit back a large scale of samples.
- ▶ Comment: no code sharing between generation and fitting, which increase robustness of studies.
- ▶ Check if model unbiased with limited statistic of 1000 signal candidates.



See [here](#) for example snakefile and instructions on pull studies

Presented the different pieces of software and techniques used so far in our analysis

- ▶ Simulation using Gauss with DecProdCut removed
- ▶ Trigger selection using Moore, with HLT1 in passthrough and trigger decisions persisted
- ▶ Fitting using Urania's PhysFit/B2DXFitter

