Table of Contents

The Bender tutorials	1.1
Getting started	2.1
The first two (almost useless) examples	2.1.1
The first algorithms	2.1.2
Loops & compounds	2.1.3
Advanced Bender	3.1
Fill-family of methods	3.1.1
TisTos-family of methods	3.1.2
Using tools	3.1.3
Bender & simulation	4.1
MC truth	4.1.1
MC match	4.1.2
processing of HepMC	4.1.3
Bender & Ganga	5.1
BenderScript	6.1
Contributing	7.1
Examples of formatting	7.2
Download PDF	8.1

The Bender tutorials build passing

These are tutorials for Bender application: "User-friendly python analysis environment for LHCb".

It is the first attempt to convert existing TWiki-based tutorials to GitHub platform, inspired by the great success of LHCb StarterKit lessons.

Bender is LHCb Python-based Physics Analysis Environment. It combines the physics content of DaVinci-project with the interactive python abilities provided by GaudiPython. It also could be considered as "Interactive LoKi". The major functionality comes from ROOT/Reflex dictionaries for the basic C++ classes and the interfaces.

These dictionaries are used primary for POOL persistency and effectively reused for interactivity. The main purpose of top-level scripts is the coherent orchestration of the Reflex dictionaries and the proper decoration of the available interfaces.



Bender dependencies are sketched here:

Doxigen documentation for Bender is accessible here.

It is assumed that users are already has some knowledge of LHCb software, in particular DaVinci and are familiar with LHCb Starterkit.

You can also add relative links within the website like this one to the first section!

Getting started

Click on the "Examples of formatting" section on the left

The first two almost useless, but very important examples

Learning objectives

• Understand the overall structure of Bender *module* and the configuration of the application

Do-nothing

Learning objectives

• Understand the overall structure of Bender *module* using the oversimplified example

Any valid Bender module must have two essential parts

- function run with the predefined signature
- function configure with the predefined dignature

For the most trivial ("do-nothing") scenario function run is

```
def run ( nEvents ) :
    # some fictive event loop
    for i in range( 0 , min( nEvents , 10 ) ) : print ' I run event %i ' % i
    return 0
```

In a similar way, the simplest "do-nothing"-version of configure -function is

```
def configure ( datafiles , catalogs = [] , castor = False , params = {} ) :
    print 'I am configuration step!'
    return 0
```

As one clearly sees, these lines do nothing useful, but they are perfectly enough to be classified as the first Bender code. Moreover, the python module with these two function can already be submitted to Ganga/Grid, and Ganga will classify it as valid Bender code. Therefore this code is already "ready-for-Ganga/Grid"!

The details for the curious students: how Ganga/Grid treat Bender modules?

Actually Ganga executes at the remote node the following wrapper code

Thats all! From this snippet you see:

- the code must have the structure of python module, namely no executable lines should appear in the main body of the file
 - (note the difference with respect to the *script*)
- it must have two functions run and configure
 - (everythnig else is not used)

The whole module is here:

```
## 1) some user code :
1
   def run ( nEvents ) :
        for i in range( 0 , min( nEvents , 10 ) ) : print ' I run event %i ' % i
4
        return 0
5
   ## 2) configuration step
6
    def configure ( datafiles , catalogs = [] , castor = False , params = {} ) :
7
        print 'I am configuration step!'
8
9
        return 0
DoNothing.py hosted with ♥ by GitHub
                                                                                        view raw
```

In practice, before the submission the jobs to Ganga/Grid, the code needs to be tested using some test-data. This, formally unnesessary, but very important step can be easily embedded into your module using python's __main__ clause:

```
if '__main__' == __name__ :
    print 'This runs only if module is used as the script! '
    configure ( [] , catalogs = [] , params = {} )
    run ( 10 )
```

Note that these lines effectively convert the *module* into *script*, and finally one gets:

```
## 1) some user code :
1
 2
    def run ( nEvents ) :
         for i in range( 0 , min( nEvents , 10 ) ) : print ' I run event %i ' % '
3
         return 0
 4
5
    ## 2) configuration step
6
    def configure ( datafiles , catalogs = [] , castor = False , params = {} ) :
 7
         print 'I am configuration step!'
8
        return 0
9
    ## 3) steer the job
    if '__main__' == __name__ :
         print 'This runs only if module is used as the script!'
11
         run ( 10 )
DoNothing.py hosted with ♥ by GitHub
                                                                                       view raw
```

How to run it interactively?

The answer is trivial:

lb-run Bender/prod python DoNothing.py

That's all. Make a try and see what you get!

Unnesessary but very useful decorations:

It is highly desirable and recommended to put some "decorations" a top of this minimalistic lines:

- add magic #!/usr/bin/env python line as the top line of the module/script
- make the script executable: chmod +x ./DoNothing.py
- add a python documentation close to the begin of the script
 - fill some useful python attributes with the proper informaton
 - __author__
 - __date__
 - __version__
 - o do not forget to add documenation in Doxygen-style and use in comments following tags
 - @file
 - @author
 - **.**..

With all these decorations the complete module is here

For all subsequent lessons we'll gradually extend this script with the additional functionality, step-by-step converting it to something much more useful.

In practice, ...

In practice, the prepared and *ready-to-use* function run is imported from some of the main Bender module Bender Main , and the only one really important task for the user is to code the function configure .

DaVinci

Learning objectives

• Understand the internal structure of the configure function

For the *typical* case in practice, the function *configure* (as the name suggests) contains three parts

1. static configuration: the configuration of Davinci configurable (almost unavoidable)

- 2. input data and application manager: define the input data and instantiate Gaudi's application manager (mandatory)
- 3. *dynamic configuration*: the configuration of GaudiPython components (optional)

Static configuration

For the first part, the instantiation of DaVinci configurable is almost unavoidable step:

Here we are preparing application to read PSIX.MDST - uDST with few useful selections for B&Q Working Group. Note that in this part one can use all power of DaVinci/Gaudi Congifurables . In practice, for physics analyses, it is veyr convinient to use here Selection framework, that allows to configure DaVinci in a very compact, safe, robust and nicely readable way, e.g. let's get from Transient Store some selection and print its content

```
from PhysConf.Selections import AutomaticData, PrintSelection
particles = AutomaticData ( 'Phys/SelPsi2KForPsiX/Particles' )
particle = PrintSelection ( particles )
```

As the last sub-step of (1), one needs to pass the selection object to $\mbox{DaVinci}$

```
dv.UserAlgorithms.append ( particles )
```

Where is `SelectionSequence`?

The underlying SelectionSequence object will be created automatically. You should not worry about it.

Input data and application manager

This part is rather trivial and almost always standard:

```
from Bender.Main import setData, appMgr
## define input data
setData ( inputdata , catalogs , castor )
## instantiate the application manager
gaudi = appMgr() ## NOTE THIS LINE!
```

while setData can appear anywhere inside configure function, the line with appMgr() is very special. After this line, no static configuration can be used anymore. Therefore all the code dealing with Configurables and Selections must be placed above this line.

Dynamic configuration

For this particular example, it is not used, but will be discussed further in conjunction with other lessons.

The complete configure function is:

```
= False ,
                                            ## use the direct access to castor/EOS ?
5
                     castor
 6
                     params
                              = {}
                                       ) :
 7
         ## import DaVinci & configure it!
         from Configurables import DaVinci
         ## delegate the actual configuration to DaVinci
         rootInTES = '/Event/PSIX'
         dv = DaVinci ( DataType
12
                                   = '2012'
                        InputType = 'MDST'
14
                        RootInTES = rootInTES )
         from PhysConf.Selections import AutomaticData, PrintSelection
16
         particles = AutomaticData( 'Phys/SelPsi2KForPsiX/Particles' )
17
         particles = PrintSelection ( particles )
         dv.UserAlgorithms.append ( particles )
         ## define the input data
         setData ( inputdata , catalogs , castor )
24
         ## get/create application manager
25
         gaudi = appMgr()
         return SUCCESS
DaVinciEx.py hosted with ♥ by GitHub
                                                                                      view raw
```

The prepared and ready-to-use function run is imported Bender. Main:

```
from Bender.Main import run
```

Now our Bender module (well, it is actually pure <code>DaVinci</code> , no real Bender here!) is ready to be used with Ganga/Grid. For local interactive tests we can use the trick with <code>__main__</code> clause: The <code>__main__</code> clause in our case contains some input data for local tests:

The complete moodule can be accessed here

How to run it?

Again, the answer is trivial (and universal):

```
lb-run Bender/prod python DoNothing.py
```

That's all. Make a try and see what you get!

Challenge

Try to convert any of your existing DaVinci simple *script* into Bender *module* and run it interactively. You can use the result of this excersize for subsequent lessons.

What is `castor`? Why `LFN` is used as input file name?

Bender is smart enough, and for many cases it can efficiently convert input LFN into the real file name.

1. First, if you have Grid proxy enabled (lncb-proxy-init) is uses internally LHCbDirac to locate and access the file. This way is not very fast, but for all practial cases this look-up is almost always successful, however for some cases certain hints could be very useful. In particular, you can specify the list of Grid sites to look for data files:

```
## define input data
setData ( inputdata , catalogs , castor = castor , grid = ['RAL', 'CERN', 'GRIDKA'] )
```

- 2. Second, for CERN, one can use option castor = True , that activates the local look-up on input files at CERN-CASTOR and CERN-EOS storages (root://castorlhcb.cern.ch and root://eoslhcb.cern.ch). This look-up is much faster than the first option, but here the success is not guaranteed, since not all files have their replicas at CERN.
- 3. Third, for access to special locations, e.g. some local files, Bender also makes a try to look into directories specified via the environment variable BENDERDATAPATH (column separated list of paths) and also try to contruct the file names using the content of environment variable BENDERDATAPREFIX (semicolumn separated list of prefixes used for construction the final file name). The file name is constructed using all (n+1)*(m+1) variants, where n is number of items in BENDERDATAPATH and m is number of items in BENDERDATAPREFIX. Using the combination of BENDERDATAPATH and BENDERDATAPREFIX variables one can make very powerful matching of *short* file names (e.g. LFN) to the actual file. Using these variables one can easily perform a local and efficient access to Grid files from some *close* Tier-1/2 center.

Keypoints

Wth these two examples, you should aready be able to

- code sove valid (but useless) Bender modules
- run them interactively

The first Bender algorithms

Prequisites

- One needs to undertand the stucture of Bender module: run , configure functions and the __main__ clause
- One needs to know the sctructure and the content of configure function

Learning objectives

- Understand Bender algorithms
 - How to code them?
 - How to embedd them into the application?

Hello, world!

Traditionally for tutorials, the first algorithm prints Hello, world . The Bender algorithm inherits from the class Algo , imported from Bender.Main module. This python base is indeed a C++ -class, that inherits from LoKi::Algo class, that in turn inherits from DaVinciTupleAlgorithm . The simplest algorithm is rather trivial:

```
from Bender.Main iport Algo, SUCCESS
class HelloWorld(Algo):
    """The most trivial algorithm to print 'Hello,world!'""

def analyse( self ) : ## IMPORTANT!
    """The main 'analysis' method"""
    print 'Hello, world! (using native Python)'
    self.Print( 'Hello, World! (using Gaudi)')
    return SUCCESS ## IMPORTANT!!!
```

Important note:

• one *must* implement the method analyse that gets no argument and returns StatusCode

Optionally one can (re)implement other important methods, like __init__ , initialize , finalize , etc... In particular initialize could be used to locate some *tools* and or pre-define some useful code fragments, e.g. some *expensive* or non-trivial LoKi-functors.

Where to put the algorithm code?

It is recommended to put the algorithm code directly in the main body of your module, outside of configure function. It allows to have visual separation of the algorithmic and configuration parts. Also it helps for independent reuse of both parts.

How to embedd the algorithm into the application?

There are two approaches brute-force, that works nicely with such primitive code as Helloworld algorithm above and the

intelligent/recommended approach, that smoothly insert the algorithm into the overall flow of algorithms, provided by Davinci

Brute-force

One can instantiate the algorithm in **configure** method **after** the instantiation of application manager, and add the algorithm, into the list of top-level algorithms, known to Gaudi:

```
gaudi = appMgr()
alg = HelloWorld('Hello')
gaudi.addAlgorithm( alg )
```

For this particular simple case one can also just replace the list of top-level Gaudi algorithms with a single Helloworld algorithm:

```
gaudi = appMgr()
alg = HelloWorld('Hello')
gaudi.setAlgorithms( [ alg ] )
```

More on an optional _dynamic configuration_

As it has been said earlier, the part of configure function, placed after gaudi=appMgr() line corresponds to *dynamic* configuration, and here one can continue the further configuration of the algorithm, e.g.

```
gaudi = appMgr()
alg = HelloWorld('Hello')
alg.QUQU = 'qu-qu!' ## define and set some "parameter"
gaudi.setAlgorithms( [ alg ] )
```

Later, this new *parameter* can be accessed e.g. in analyse function:

```
class HelloWorld(Algo):
    """The most trivial algorithm to print 'Hello,world!'"""

def analyse( self ) : ## IMPORTANT!
    """The main 'analysis' method"""
    print 'Hello, world! (using native Python)', self.QUQU ## use "parameter"
    self.Print( 'Hello, World! (using Gaudi)')
    return SUCCESS ## IMPORTANT!!!
```

Such trick is in general a bit fragile, but it is often useful if one has several instances of the algorithm that differ only by some configuration parameter.

```
alg1 = MyALG ( ... )
alg2 = MyALG ( ... )
alg3 = MyALG ( ... )
alg1.decay_mode = '[D0 -> K- pi+]CC'
alg2.decay_mode = '[D0 -> K- K+ ]CC'
alg3.decay_mode = '[D0 -> pi- pi+]CC'
```

This approach is very easy and rather intuitive, but is not so easy to insert the algorithm into existing non-trivial flow of algorithms without a danger to destroy the flow. In this way one destroys various standard actions, like (pre)filtering, luminosity calculation etc., therfore it could not be recommended for the real physics analyses, but it could be used for some simple special cases.

Intelligent approach

For intelligent approach one uses Selection wrapper for Bender algorithm, BenderSelection. This wrapper behaves as any other

selection-objects, and it lives in static configuration part of configure function:

As the next step in *dynamic configuration* part of configure function one instantiates the algoritm taking all the configuration from the selection-object:

```
gaudi = appMgr()
alg = HelloWorld( hello )
```

To complete the module one (as usual) need to combine in the file

- 1. implementation of Helloworld algorithm
- 2. configure function with proper static and dynamic configurations
- 3. __main__ clause
- 4. (run function is imported from Bender.Main module)

The complete module can be accessed here

Get data, fill histos & n-tuples

Well, now your Bender algorithm knows how to print Hello, world! . Note that it also gets some data: in the previous example we fed it with particles -selection. Now try to get this data inside the algorithm and make first simpel manipulations with data.

select method

The method select is a heart of Bender algorithm. It allows to select/filter the particles that satisfies some criteria from the input particles. The basic usage is:

```
myB = self.select ( 'myB' , ('B0' == ABSID ) | ('B0' == ABSID ) )
```

The method returns collection filtered particles

The first argument is the tag, that will be associated with selected particles, the second argument is the selection criteria. The tag *must* be unique, and the selection criteria could be in a form of

- LoKi predicate: LoKi-functor that get the particle as argument and return the boolean value
- *decay descriptor*, e.g. 'Beauty --> J/psi(1S) K+ K-' . Some components of the decay descriptor can be *marked*, and in this case, only the *marked* partcles will be selected:

```
myB = self.select ( 'beauty' , 'Beauty --> J/psi(1S) K+ K-') ## get the heads of the decay <math>myK = self.select ( 'kaons' , 'Beauty --> J/psi(1S) ^K+ ^K-') ## get only kaons
```

As soon as one gets some good, filtered particles there are many possible actions

• print it!

```
myB = self.select ( 'myB' , ('B0' == ABSID ) | ('B0' == ABSID ) )
print myB
```

loop

```
myB = self.select ( 'myB' , ('B0' == ABSID ) | ('B0' == ABSID ) )
for b in myB :
print 'My Particle:', p
print 'some quantities: ', M(p) , PT(p) , P(p)
```

• fill histograms

• fill n-tuple:

```
myB = self.select ( 'myB' , ('B0' == ABSID ) | ('B0' == ABSID ) )
t = self.nTuple('TupleName')
for b in myB :
t.column_float ( 'pt' , PT (p)/GeV)
t.column_float ( 'm' , M (p)/GeV)
t.column_float ( 'm_psi' , M1 (p)/GeV)
t.column_float ( 'm_kk' , M23(p)/GeV)
t.write()
```

For n-tuples...

Sicne n-tuples (ROOT's TTree objects) resides in ROOT-file, to use n-tuples, one also need to declare the output file for TTree s: The easiest way is to rely on TupleFile property of Davinci:

Challenge

Add select statements, histos and tuples to your Helloworld algorithm, created earlier, and run it.

Solution

The complete module is accesible here

Keypoints

Wth these two examples, you should aready be able to

- code Bender *algorithms* and insetr them into overall algorithm flow
- loop over data, fill historgrams and n-tuples

Create the compound particles in Bender

Prequisites

• One needs to understand the way how Bender accesses the data

Learning objectives

• Understand how Bender algorithm combines the particles and creates the compound particles

Make-B

The next example illustrates how one combines the particles and create the compound particles inside the Bender algorithm. Let's consider a simple case of creation of B+ -> J/psi(1S) K+ decays.

The first step is rather obvious: before getting the combinations, we need to get the individual components. Here select function does the job:

```
## get J/psi mesons from the input
psis = self.select ( 'psi' , 'J/psi(1S) -> mu+ mu-' )
## get energetic kaons from the input:
kaons = self.select ( 'k' , ( 'K+' == ID ) & ( PT > 1 * GeV ) & ( PROBNNK > 0.2 ) )
```

The loop over psi k combinations is ratehr trivial:

```
## make a loop over J/psi K combinations :
for b in self.loop( 'psi k' , 'B+' ) :
    ## fast evaluation of mass (no fit here!)
    m12 = b.mass ( 1 , 2 ) / GeV
    print 'J/psiK mass is %s[GeV]' % m12
    p1 = b.momentum ( 1 ) / GeV
    p2 = b.momentum ( 2 ) / GeV
    p12 = b.momentum ( 1 , 2 ) / GeV
    print 'J/psiK momentum is is %s[GeV]' % p12
```

Looping object (b here), as the name, suggests, make a loop over all psi k combinations. The loop is done in CPU efficient way, and no expensive vertex fitting is performed. One can estimate various raw (no fit) kinematical quantities using functions momentum, mass, etc... (Note that indices starts from 1. For all LoKi-based functions the index 0 is reserved for self-reference, the mother particle itself). These raw quantities can be used for quick reject of bad combinations before making CPU-expensive vertex fit. If/when combination satisfies certain criteria, the vertex

fit and creatino of the compound particle is triggered automatically if any of particle/vertex information is retrieved (either directy via particle/vertex method, or indirectly, e.g. via call to any particle/vertex LoKi-functor. The good created mother particles can be saved for subsequent steps under some unique tag:

```
for b in self.loop( 'psi k' , 'B+' ) :
    ## fast evaluation of mass (no fit here!)
    m12 = b.mass ( 1 , 2 ) / GeV
    if not 5 < m12 < 6 : continue
    chi2vx = VCHI2 ( b )  ## indirect call for vertex fitr and creation of B+ meson
    if not 0<= chi2v < 20 : continue
    m = M ( p ) / GeV
    if not 5 < m < 5.6 : continue
    m.save('MyB')</pre>
```

Obviously the looping can be combnied with filling of historgams and n-tuples.

How to deal with charge conjugation?

```
One can make two loops:
```

```
psis = self.select ( 'psi' , 'J/psi(1S) -> mu+ mu-' )
kplus = self.select ( 'k+' , ( 'K+' == ID ) & ( PT > 1 * GeV ) & ( PROBNNK > 0.2 ) )
kminus = self.select ( 'k-' , ( 'K-' == ID ) & ( PT > 1 * GeV ) & ( PROBNNK > 0.2 ) )

bplus = self.loop( 'psi k+', 'B+' ) ## the first loop object
bminus = self.loop( 'psi k-', 'B-' ) ## the second loop object
for cc in ( bplus , bminus ) :
    for b in cc :
        m12 = b.mass(1,2) / GeV
        ...
        b.save('MyB')
```

The popular alternative is ${\it charge-blind}$ loop, that is a bit simpler, but it requires some accuracy:

```
psis = self.select ( 'psi' , 'J/psi(1S) -> mu+ mu-' )
## ATTENTION: select both K+ and K-, note ABSPID here
k = self.select ( 'k' , ( 'K+' == ABSID ) & ( PT > 1 * GeV ) & ( PROBNNK > 0.2 ) )

for b in self.loop ( 'psi k' , 'B+' )
    m12 = b.mass(1,2) / GeV
    ...
    psi = b(1) ## get the first daughetr
    k = b(2) ## get the second daughter

## ATTENTION: redefine PID on-flight
    if Q( k ) > 0 : b.setPID( 'B+' )
    else : b.setPID( 'B-' )
b.save('MyB')
```

The saved particles can be extracted back using the method selected:

```
myB = self.selected('MyB')
for b in myB :
    print M(b)/GeV
```

Configuration

It is clear that to build $\,$ B+ -> J/psi(1S) K+ $\,$ decays, one needs to

feed the algorithm with J/psi(1S)-mesond and kaons. Using selection machinery is the most efficient ansd transaprent way to do it.

```
from PhysConf.Selections import AutomaticData
jpsi = AutomaticData( '/Event/Dimuon/Phys/FullDSTDiMuonJpsi2MuMuDetachedLine/Particles' )

from StandardParticles import StdLooseKaons as kaons
bsel = BenderSelection ( 'MakeB' , [ jpsi , kaons ] )
```

The complete example of creation of B+ -> J/psi(1S) K+ decays starting from DIMUON.DST is accessible from here

Keypoints

Wth these example, you should be able to do

• code Bender *algorithm* that perofem loopint, combianig anc creation of compound particles

Advanced Bender

Here we discuss some advanced features of Bender, namely

- ullet the powefull fill of n-tuples provied by BenderTools.Fill module
- ullet the treatment of $\tilde{ tilde{ t$

Advanced fill of n-tuples

The n-tuple filling functionality, described above is drastically extended using the functions from BenderTools.Fill module. The import of this module add following functions to the base class Algo:

Method	Short description		
treatPions	add information about pions		
treatKaons	add information about kaons		
treatProtons	add information about protons		
treatMuons	add information about muons		
treatPhotons	add information about photons		
treatDiGammas	add information about di-photons (pi0, eta,)		
treatTracks	add information about the tracks		
treatKine	add detailed kinematic information for the particle		
fillMasses	masses of sub-combinations		
addRecSummary	add rec-summary information		
addGecInfo	add some GEC-info		

These methods can be considered as a kind of very light *tuple-tools*. All of them are (well) documented and one can easily inspect them:

```
import BenderTools.Fill
from Bender.Main import Algo
help(Algo.treatPions)
```

Also all these methods print detailed how-to infomratino in log-file at the moment of the first invoke, and it vasn be very helpful to understand the branches in n-tuple/tree, e.g.

```
# BenderTools.Fill
                            INFO
                                  treatTracks: The method adds track-specific information into n-tuple
      tup = ... ## n-tuple
    b = ... ## the particle (or looping object)
     self.treatTracks ( tup , b , '_B' ) ## suffix is optional
    Following variables are added into n-tuple:
#
     - deltaM2_min_track_ss/os[+suffix]:
    Minimal value of delta_m2(track1, track2) for all pairs of same-sign (``_ss'')
     and opposite sign ``_os'' tracks, where function minm2 is
     delta_M2(p1,p2) = (m^2(p1+p2) - 2*m^2(p1)-2*m^2(p2))/m^2(p1+p2)
     see LoKi::Kinematics::deltaM2
     - deltaAlpha_min_track_ss/os[+suffix]:
    Minimal value of the angle between two momenta for all pairs of same-sign (``_ss'')
     and opposite sign ``_os'' tracks
     see LoKi::Kinematics::deltaAlpha
     - overlap_max_track_ss/os[+suffix]:
    Maximal value ``overlap'' for all pairs of same-sign (``_ss'')
    and opposite sign ``_os'' tracks
       ``Overlap'' is defined as fraction of common/shared hits between two tracks
    see LHCb::HasIDs::overlap
      - minPt_track[+suffix]
    Minimal pT of the tracks
#
     - min/maxEta_track[+suffix]
     Minimal/maximal eta/pseudorapidity of the tracks
     - maxChi2_track[+suffix]
   Maximal chi2/ndf for the track
      - minKL_track[+suffix]
    Minimal value of Kullback-Leibler divergency for the tracks
       - maxTrGh_track[+suffix]
    Maximal value of Track Ghost probability for the tracks (track-based)
      - maxAnnGh_track[+suffix]
    Maximal value of
                           Ghost probability for the tracks (PID-based)
     - n track[+suffix]
#
     Number of tracks in the decay
    And then for each track in the decay:
     - p_track[+suffux] momentum of the track
     - pt_track[+suffux] transverse momentum of the track

- eta_track[+suffux] eta/pseudorapidity of the track

- phi_track[+suffux] phi (azimuth angle) of the track
#
     - chi2_track[+suffux] chi2/ndf of the track
     - PChi2_track[+suffux] fit probability calculated from chi2/ndf of the track
     - ann_track[+suffix] Ghost probability (PID-based)
#
      - trgh_track[+suffix] Track Ghost probability (Track-based)
```

The typical usage of these methods is:

```
tup = self.nTuple('MyTuple')
for p in particles :

psi = p(1) ## the first daughter: J/psi

## fill few kinematic variables for the particles:
self.treatKine ( tup , p , '_b' ) ## use the suffix to mark variables
self.treatKine ( tup , psi , '_psi' ) ## use the suffix to mark variables

self.treatKaons ( tup , p ) ## fill some basic information for all kaons
self.treatMuons ( tup , p ) ## fill some basic information for all muons
self.treatTracks ( tup , p ) ## fill some basic information for all charged tracks

tup.write()
```

Challenge

- 1. Add (some of) these functions into your previous Bender module with n-tuples.
- 2. Run it and observe the detailed printout in log-file
- 3. Observe new variables in your n-tuple/tree and find their description in the log-file or via help(Algo.<THEMETHOD>)
 - Is the description for all new varibales clear enough?

Solution

The complete module is accessible here and the corresponsing log-file is here

Handling of TisTos -information in Bender

Bender offers set of methods to handle TisTos -information in (relatively) easy way. This functionality comes from BenderTools.TisTos module. In short, it adds three relates method in the base class Algo:

Method	Short description
decisions collect the trigger decisions for the given particle	
trgDecs print the collected trigger statistics in readable way	
tisTos	fill N-tuple with tistos -information

All of them are (relatively well) documented and one can easily inspect them:

```
import BenderTools.TisTos
from Bender.Main import Algo
help(Algo.decisions)
```

How to know what trigger lines are relevant or the given decay/particle?

Often information about the releavant trigger lines are spread in corridors in a form of myths, general beliefs or, in the best case, references to ANA-notes for some similar analysis. However it is very simple to collect this infomation using Bender. The method decisions is our friend here, the usage of this method require some preparatory work for the algorithm, namely one needs to instrument initialize - method:

```
class TrgLines(Algo):
    """Collect infomation about the trigger lines relevant for certain decays/particles
    """
    def initialize ( self ) :
        sc = Algo.initialize ( self ) ## initialize the base class
        if sc.isFailure() : return sc

#
        ## container to collect trigger information, e.g. list of fired lines
#
        triggers = {}
        triggers ['psi'] = {} ## slot to keep the information for J/psi
        triggers ['K'] = {} ## slot to keep the information for kaons
        triggers ['B'] = {} ## slot to keep the information for B-mesons

        sc = self.tisTos_initialize ( triggers , lines = {} )
        if sc.isFailure() : return sc
        return SUCCESS
```

Here in this example we want to collect TisTos -information for J/psi , K and the whole B+ -meson. Then in the main analyse -method one just needs to invoke the method decisions for each particles in interest:

```
def analyse( self ): ## IMPORTANT!
"""The main 'analysis' method """
...
for b in particles:

    psi =b(1) ## the first daughter: J/psi
    k =b(2) ## the second daughter: K

    ## collect trigger information for J/psi
    self.decisions ( psi , self.triggers['psi'] )

    ## collect trigger information for kaons
    self.decisions ( k , self.triggers['K'] )

    ## collect trigger information for B-mesons
    self.decisions ( b , self.triggers['B'] )
...
return SUCCESS
```

Thats all. Then when jobs runs it dumps to the log-file the running trigger statistics, and the statistics is dumped into the file TrgLines_tistos.txt (<ALGNAME>_tistos.txt in general). The summary table looks like:

```
Triggers for psi
*************************
Hlt1_TIS psi #lines: 7 #events 321
( 0.62 +- 0.44 ) Hlt1DiMuonHighMassDecision
 ( 0.62 +- 0.44 )
                    Hlt1DiMuonLowMassDecision
( 19.00 +- 2.19 ) Hlt1TrackAllL0Decision
( 3.74 +- 1.06 ) Hlt1TrackAllL0TightDecision
( 4.36 +- 1.14 ) Hlt1TrackMuonDecision
                    Hlt1TrackPhotonDecision
 ( 0.93 +- 0.54 )
( 0.31 +- 0.31 )
                    Hlt1VertexDisplVertexDecision
                     TOTAL
(100.00 +- 0.31 )
Hlt1_TOS psi #lines:
                      9 #events 321
( 72.90 +- 2.48 )
                     Hlt1DiMuonHighMassDecision
( 59.81 +- 2.74 )
                    Hlt1DiMuonLowMassDecision
( 0.31 +- 0.31 )
                    Hlt1DiProtonDecision
                    Hlt1SingleMuonHighPTDecision
( 9.03 +- 1.60 )
 ( 0.31 +- 0.31 )
                     Hlt1SingleMuonNoIPDecision
                    Hlt1TrackAllL0Decision
( 44.24 +- 2.77 )
( 10.90 +- 1.74 )
                    Hlt1TrackAllL0TightDecision
 ( 74.77 +- 2.42 )
                    Hlt1TrackMuonDecision
( 0.31 +- 0.31 )
                     Hlt1TrackPhotonDecision
(100.00 +- 0.31 )
                      T0TAL
Hlt1_TPS psi #lines: 6 #events 321
( 13.71 +- 1.92 ) Hlt1DiMuonHighMassDecision
( 10.90 +- 1.74 )
                    Hlt1DiMuonLowMassDecision
( 0.31 +- 0.31 )
                     Hlt1DiProtonDecision
 (3.12 + - 0.97)
                      Hlt1TrackAllL0Decision
                    Hlt1TrackAllL0TightDecision
 ( 0.93 +- 0.54 )
 ( 4.67 +- 1.18 )
                    Hlt1TrackMuonDecision
 (100.00 +- 0.31 )
                      TOTAL
```

Only a short fragment is shown here, one gets similar fragments for all declared particles (psi , K and B) and for all trigger levels (LO , Hlt1 and Hlt2). The full table is accessible here The content of the summaty table is rather intuitive: it summarizes the fire frequences for varios trugegr lines for three regimes TIS , TOS and TPS . Inspecting such table, one immediately concludes that the most relevan Hlt1-TOS -line is Hlt1DimuonHighMassDecision . Other Hlt1-TOS -lines are less relevant here. But please note that here only very small statistics is used (321 event), and with larger statistics sthe conclusions could be corrected. E.g. due to small statistics here, for Hlt1-TIS -lines the choice is not evident: one clearly see that Hlt1TrackAllLODecision line is important, but for importance of other lines one csn judge only after the significant increase of the statistics.

In practice to make a decision, large statistics is required (for real data and/or for simulated samples). And here these files are very useful. The trigger statistics is saved not only in <algnosure color by the samples of the color by the samples of the color by the color by

Challenge

- 1. Add decisions -function for your previous Bender module with n-tuples.
 - (Do not forget to instrument the initialize method)
- 2. Run it and observe the output summary table
- 3. Identify the relevant LO-TOS , Hlt1-TOS and Hlt2-TOS lines for your decay
 - Does it correspond to your expectations?

Solution

The complete module is accessible here and the corresponsing summary table is here

How to add the TisTos -information to n-tuple/tree?

Now, when we have the lists of the relevant lines, one wants to add infrommation about the decisions of these lines into n-tuple/tree. The method tistos is out friend here. To use this method one needs to instrument initialize -method:

```
class TisTosTuple(Algo):
    """Enhanced functionality for n-tuples
   def initialize ( self ) :
        sc = Algo.initialize ( self ) ## initialize the base class
        if sc.isFailure() : return sc
        ## container to collect trigger information, e.g. list of fired lines
        triggers ['psi'] = {} ## slot to keep information for J/psi
        ## the lines to be investigated in details
        lines = \{\}
        lines [ "psi" ] = {} ## trigger lines for J/psi
        ## six mandatory keys:
        lines [ "psi" ][ 'LOTOS' ] = 'LO(DiMuon|Muon)Decision'
        lines [ "psi" ][ 'LOTIS' ] = 'LO(Hadron|DiMuon|Muon|Electron|Photon)Decision'
        lines [ "psi" ][ 'Hlt1TOS' ] = 'Hlt1(DiMuon|TrackMuon).*Decision'
        lines [ "psi" ][ 'Hlt1TIS' ] = 'Hlt1(DiMuon|SingleMuon|Track).*Decision'
lines [ "psi" ][ 'Hlt2TOS' ] = 'Hlt2DiMuon.*Decision'
        lines [ "psi" ][ 'Hlt2TIS' ] = 'Hlt2(Charm|Topo|DiMuon|Single).*Decision'
        sc = self.tisTos_initialize ( triggers , lines )
        if sc.isFailure() : return sc
        return SUCCESS
```

Here one defines six regex -patterns that describe the six sets of triggers lines: L0 , Hlt1 , Hlt2 vs TIS , TOS . These expressions are coded according to the information, obtainer earlier. The next step is rather trivial: in analyse method one need to invoke the method tistos for the given particle, J/psi in our case:

This code adds several variables into n-tuple/tree, see log-file or use help(Algo.tistos) . Also TisTos -information for *global* and *physics* triggers is added.

```
In details,

The fragment from log-file:
```

```
# BenderTools.TisTos
                        INFO tisTos: Fill TisTos information into n-tuple
   # for d in particles :
   # self.tisTos ( d  ,
#
#
                     tup
#
    #
                     'd0_' ,
                   self.lines ['D0'] ,
#
   #
                   self.l0tistos
#
                   self.l1tistos
                    self.l2tistos
#
   ``lines'' here is a dictionary of lines (or regex-patterns) with
   - LOTOS
#
#
    - LOTIS
   - Hlt1TOS
#
   - Hlt1TIS
   - Hlt2TOS
   - Hlt2TIS
    e.g.
   # lines = {}
   # lines ['LOTOS' ] = 'LOHadronDecision'
    # lines ['LOTIS' ] = 'LO(Hadron|Muon|DiMuon)Decision'
   # lines ['Hlt1TOS'] = ...
   # lines ['Hlt1TIS'] = 'Hlt1Topo.*Decision'
   # lines ['Hlt2TOS'] = ...
   # lines ['Hlt2TIS'] = ...
    Technically it is useful to keep it as ``per-particle-type'' dictionary
#
   # def initialize ( self ) :
#
   # ...
         self.lines
                        = {}
         self.lines ['B' ] = {}
#
       self.lines ['B' ]['L0TOS'] = ...
       self.lines ['B' ]['LOTOS'] = ...
    #
#
    #
         self.lines ['psi'] = {}
       self.lines ['psi']['L0TOS'] = ...
    #
#
   # return SUCCESS
#
    # def analyse ( self ) :
         particles = ...
#
#
    #
        for B in particles :
#
          self.tisTos ( B
                        tup ,
#
                          'B0_' ,
#
    #
                          self.lines['B'] ,
                          self.l0tistos ,
#
    #
#
                          self.l1tistos
                          self.12tistos
#
#
   #
#
   # return SUCCESS
   The function adds few columns to n-tuple, the most important are
   - ``<label>l0tos'' that corresponds to 'L0-TOS'
    - ``<label>l0tis'' that corresponds to 'L0-TIS'
#
    - ``<label>l1tos'' that corresponds to 'Hlt1-TOS'
#
    - ``<label>l1tis'' that corresponds to 'Hlt1-TIS'
   - ``<label>l2tos'' that corresponds to 'Hlt2-TOS'
   - ``<label>l2tis'' that corresponds to 'Hlt2-TIS'
   Additionally information for five predefined lists is stored:
    - ``<label>l0phys'' : L0-physics lines,
   - ``<label>l1phys'' : Hlt1-physics lines (routing bit #46)
#
  - ``<label>l2phys'' : Hlt2-physics lines (routing bit #77)
```

```
# - ``<label>l1glob'' : Hlt1Global decision
      - ``<label>12glob'' : Hlt2Global decision
  #
      The stored value is unsigned short, a bit-representaion of ITisTos::TisTosTob object
      - see ITisTos
  #
      - see ITisTos::TisTosTob
  #
     Later, in processing of TTree one can use these flags as :
     >>> tree = ...
     >>> tree.Draw('M'
                                    ) ## no trigger requirements
      >>> tree.Draw('M', '(10tos&2)==2') ## require L0-tos with respect to the list of 'L0TOS'-lines
      >>> tree.Draw('M', '(l1tos&2)==2') ## require Hlt1-tos with respect to the list of 'Hlt1TOS'-lines
      >>> tree.Draw('M', '(12tos&2)==2') ## require Hlt2-tos with respect to the list of 'Hlt2TOS'-lines
     >>> tree.Draw('M', '(10tis&4)==4') ## require LO-tis with respect to the list of 'LOTIS'-lines
      >>> tree.Draw('M', '(l1tis&4)==4') ## require Hlt1-tis with respect to the list of 'Hlt1TIS'-lines
  #
      >>> tree.Draw('M', '(12tis&4)==4') ## require Hlt2-tis with respect to the list of 'Hlt2TIS'-lines
     >>> tree.Draw('M', '(10tos&3)==3') ## require L0-tus with respect to the list of 'L0TOS'-lines
     >>> tree.Draw('M', '(l1tos&3)==3') ## require Hlt1-tus with respect to the list of 'Hlt1TOS'-lines
      >>> tree.Draw('M', '(12tos&3)==3') ## require Hlt2-tus with respect to the list of 'Hlt2TOS'-lines
     >>> tree.Draw('M', '(10tos&1)==1') ## require L0-tps with respect to the list of 'L0TOS'-lines
     >>> tree.Draw('M', '(l1tos&1)==1') ## require Hlt1-tps with respect to the list of 'Hlt1TOS'-lines
     >>> tree.Draw('M', '(12tos&1)==1') ## require Hlt2-tps with respect to the list of 'Hlt2TOS'-lines
      >>> tree.Draw('M', '(10phys&2)==2') ## require LO-tos with respect to LO-physics lines
      >>> tree.Draw('M', '(10phys&4)==4') ## require L0-tis with respect to L0-physics lines
     One can avoid bit-wise operations using ``verbose=True'' flag.
      In this case one gets following boolean variables in n-tuple:
       - ``<labeltag>_tos'' Trigger On Signal
     - ``<labeltag>_tis'' Trigger Independently on Signal
  #
     - ``<labeltag>_tps'' Trigger Partially on Signal
      - ``<labeltag>_tus'' Trigger Used Signal (== TOS || TPS )
      - ``<labeltag>_dec'' Trigger decision
     where ``<labeltag>'' is
     - ``<label>l0tos''
     - ``<label>10tis''
      - ``<label>l0phys'
  #
      - ``<label>l1tos''
  #
     - ``<label>l1tis''
  #
     - ``<label>l1phys'
     - ``<label>l1glob'
  #
      - ``<label>l2tos''
     - ``<label>l2tis''
 #
     - ``<label>l2phys'
     - ``<label>l2glob'
```

Challenge

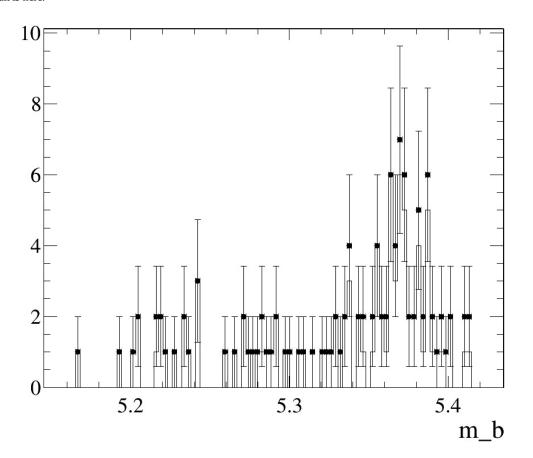
- 1. Add tistos -function for your previous Bender module with n-tuples.
 - (Do not forget to instrument the initialize method)
- 2. Run it and observe new variables in n-tuple/tree
- 3. Make a plot of B-mass for all candidates, and for candiated that are LO-TOS, H1t-TOS and H1t2-TOS with respect to the set of selected trigger lines.

Solution

The complete module is accessible here and the corresponsing log-file is here To make the corresponsing plot, e.g. start (i)python:

```
import ROOT
f = ROOT.TFile('TisTosTuples.root', 'READ')
t = f.Get('TisTos/MyTuple')
t.Draw ( 'm_b' ,'' , 'e1' )
t.Draw ( 'm_b' , ' psi_lotos_tos && psi_l1tos_tos && psi_l2tos_tos' , 'same hist' )
```

The result is here:



Contributing

bender-tutorials is an open source project, and we welcome contributions of all kinds:

- New lessons;
- Fixes to existing material;
- Bug reports; and
- Reviews of proposed changes.

By contributing, you are agreeing that we may redistribute your work under these licenses. You also agree to abide by our contributor code of conduct.

Getting Started

- 1. We use the fork and pull model to manage changes. More information about forking a repository and making a Pull Request.
- 2. To build the lessons please install the dependencies.
- 3. For our lessons, you should branch from and submit pull requests against the master branch.
- 4. When editing lesson pages, you need only commit changes to the Markdown source files.
- 5. If you're looking for things to work on, please see the list of issues for this repository. Comments on issues and reviews of pull requests are equally welcome.

Dependencies

To build the lessons locally, install the following:

1. Gitbook

Install the Gitbook plugins:

\$ gitbook install

Then (from the bender-tutorials directory) build the pages and start a web server to host them:

\$ gitbook serve

You can see your local version by using a web-browser to navigate to http://localhost:4000 or wherever it says it's serving the book.

The title

Learning Objectives

- The starterkit lessons all start with objectives about the lesson
- Objective 2 with some *formatted* **text** *like* this

Basic formatting

You can make **bold**, *italic* and strikethrough text. Add relative links like this one and absolute links in a couple of different ways.

Have bulleted lists:

- Point 1
- Point 2
 - o Sub point
 - Sub point
 - Sub point
- Point 2

Use numbered lists:

- 1. First
- 2. Second
 - i. Second first
 - i. Second first first
 - ii. Second second
- 3. Third

LaTeX

You can use inline LaTeX maths such as talking about the decay $\$D^{*+} \subset D^0 \subset K^{-}$ yiéhtarrow $K^{-} \in K^{-}$

Code highlighting

And have small lines of code inline like saying print("Hello world") or have multiple lines with syntax highlighting for python:

```
import sys

def stderr_print(string):
    sys.stderr.write(string)

stderr_print("Hello world")
```

bash:

```
lb-run Bender/latest $SHELL
dst_dump -f -n 100 my_file.dst 2>&1 | tee log.log
```

Callouts

• Summary point 1

Prequisites	
 Prequisite 1 Prequisite 2 	
Objectives	
 Objective 1 Objective 2 	
Challenge	
Set a challenge here, and the solution will remain hidden until it's clicked • How to print?	
Solution	
The answer is: print("Hello world")	
Extra details that are hidden by default	
Some extra details	
Keypoints	

Quotes

This was said by someone

Tables

Simple tables are possible

First Header	Second Header
Content from cell 1	Content from cell 2
Content in the first column	Content in the second column

Images



Section types

This is a section

Subsections

And a subsection

Subsubsections

And a subsubsection