

Axion-Like Particle Search in

$$B \rightarrow K^{(*)} + a, a \rightarrow \gamma\gamma$$

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ALP in FCNC

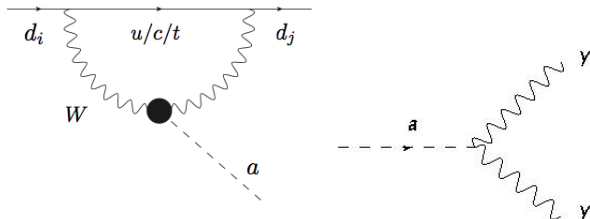
Introduction

ALPs (Axion-Like Particles: pseudoscalar neutral massive particles) could couple to gauge bosons and induce FCNC.

$$\text{Coupling: } \mathcal{L} \supset -\frac{g_a V}{4} a V_{\mu\nu} \tilde{V}^{\mu\nu}$$

Where:

$\tilde{V}^{\mu\nu} = \epsilon^{\mu\nu\rho\sigma} V_{\rho\sigma}/2$, a and V are ALP and gauge boson fields.



See Brian's paper: <https://arxiv.org/pdf/1611.09355.pdf>

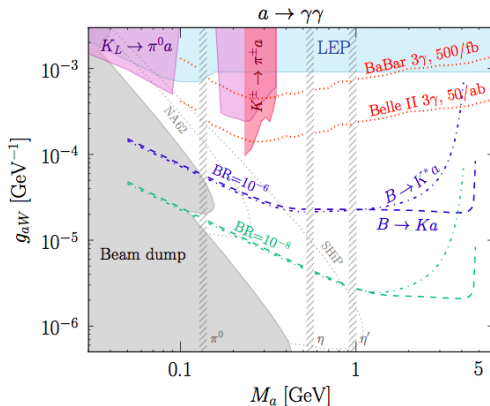
ALP in FCNC

Estimated sensitivity

Searched channel (diphoton resonance):

$$B \rightarrow K^{(*)} a, a \rightarrow \gamma\gamma$$

Exploits (often-overlooked) coupling of ALPs to $W^\pm \Rightarrow$ FCNC production of ALPs (direct quark coupling unnecessary).



What was done till last time (June 2017):

- General overview of the ntuples;
- Characterize signal properties and resolution;
- Compare signal to bkg (continuum and B^+B^-);
- Explore first discriminant variables;
- First steps into MVA.

What's **new**:

- Finalization of exploration of discriminant variables;
- Optimization of MVA method and strategy;
- Check stability of MVA over ALP masses and cuts;
- Validation MC bkg with Run3 data;
- Fit and characterization of signal.

Backgrounds

Sources of bkg: $B\bar{B}$, Bhabha, continuum, $\tau^+\tau^-$, $\mu^+\mu^-$, ...

Main sources of bkg are continuum and $B\bar{B}$:

- Continuum ($e^+e^- \rightarrow q\bar{q}$; $q = uds + c$) generally really different behavior than signal but a random K and 2 γ can be reconstructed as signal $\rightarrow$ combinatorial bkg.
Shape variables are powerful to discriminate: signal is uniform (Bs from Υ are slow), bkg is highly directional (jets).
- $B\bar{B}$ can look like signal, but $B \rightarrow K + X$, $X = \pi^0, \eta, \eta' \rightarrow \gamma\gamma$ have low BR and we veto over them; remains the combinatorial.

- Lot of variables $\Rightarrow$ **MVA**;
- Training MVA for the 2 main bkg, check a posteriori that is good for the secondary bkg too;
- Need to choose a training strategy:
 - Strategy 1step: train MVA for signal vs (B^+B^- and continuum) together, weighted, including m_{ES} and ΔE ;
 - Strategy 2steps: train MVA for signal vs continuum, then separately for B^+B^- , including m_{ES} and ΔE ;
 - Strategy 3steps: train MVA for signal vs continuum without m_{ES} nor ΔE , cut over m_{ES} and ΔE , train for signal vs B^+B^- .
- Also need to choose a MVA method (neural network, boosted decision tree, ...).

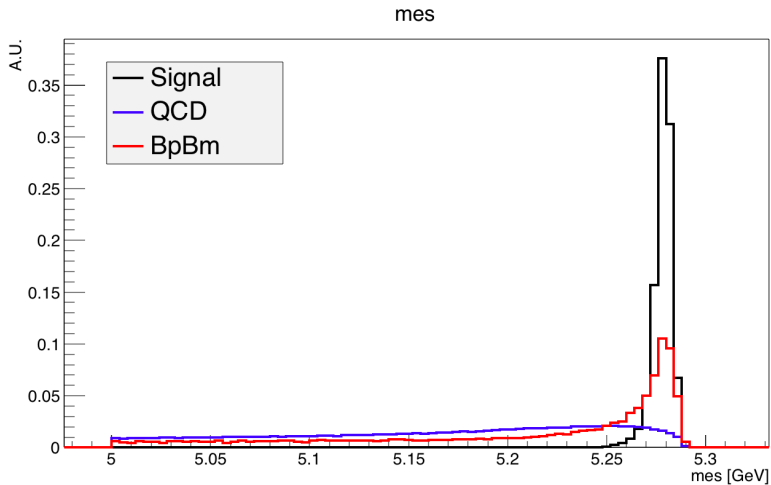
Discriminant variables

Variables used for selection:

- Δ_E ;
- m_{ES} ;
- Legendre0;
- Legendre2;
- $\theta_{sphericity}$;
- θ_{thrust} ;
- γ energy;
- γ helicity angle;
- K helicity angle;
- K PID selector;
- Pi0Veto.

Discriminant variables

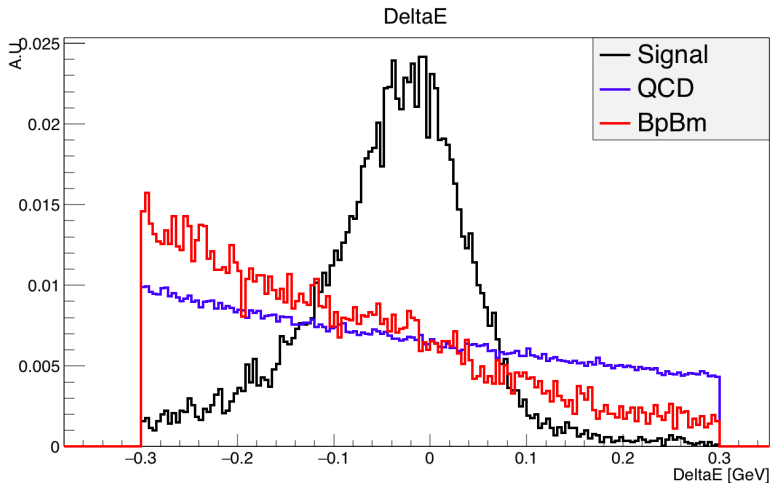
m_{ES}



Discriminant variables

Δ_E

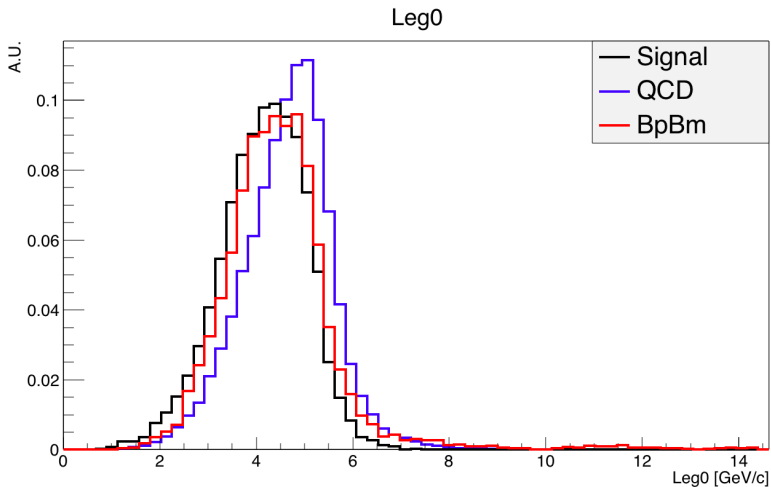
Pre-selections applied at the time of ntuple generation.



Discriminant variables

Leg0

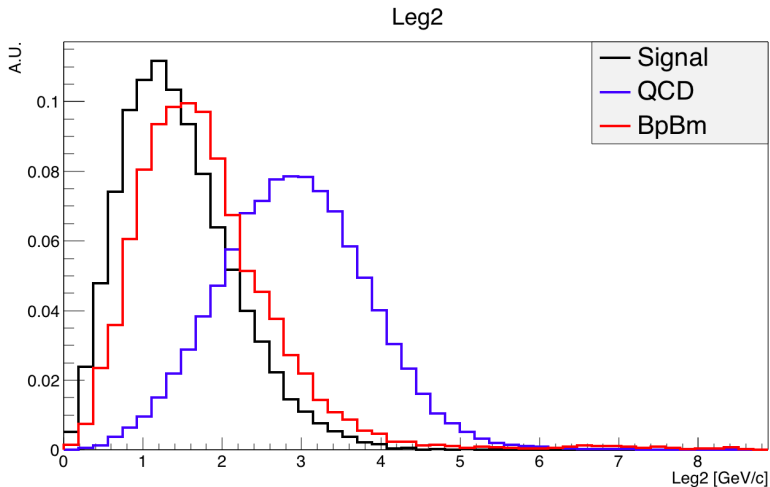
Legendre moment 0.



Discriminant variables

Leg2

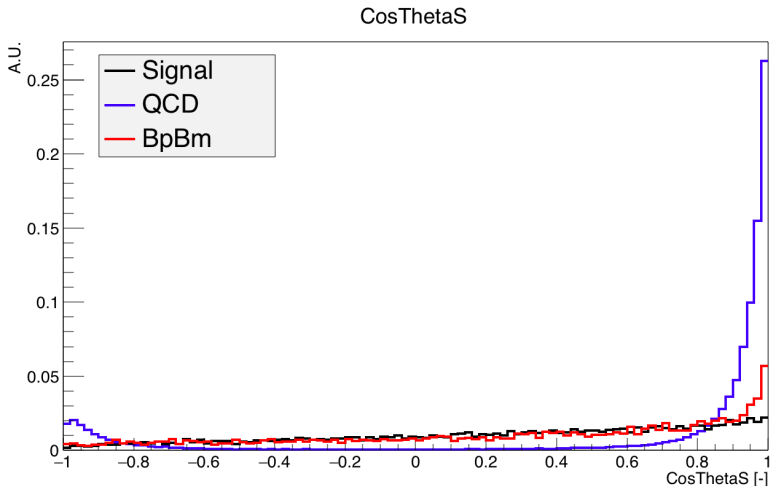
Legendre moment 2.



Discriminant variables

$\cos(\theta_S)$

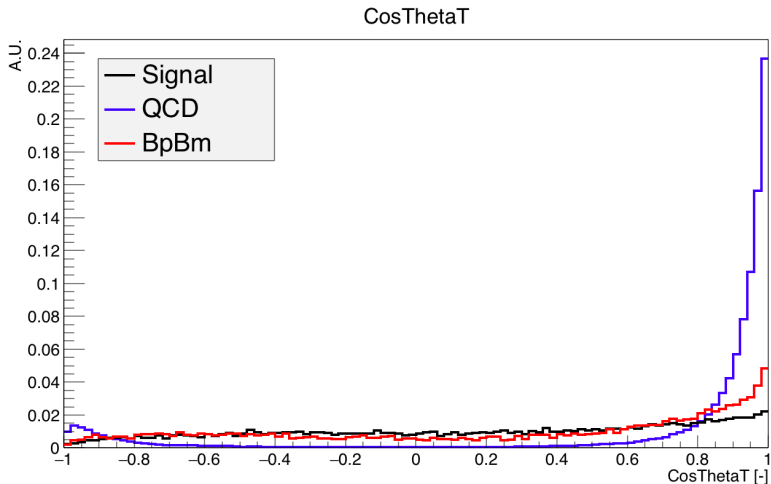
θ_S = sphericity angle (between sphericity axis of B and ROE).



Discriminant variables

$\cos(\theta_T)$

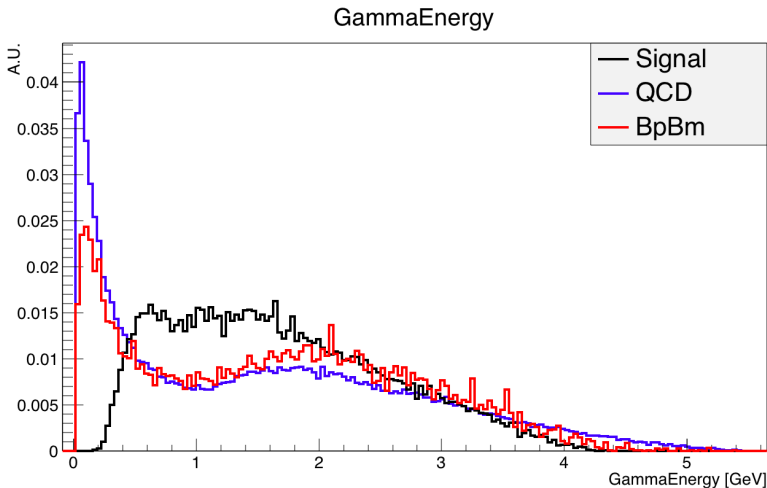
θ_T = thrust angle (between thrust axis of B and ROE).



Discriminant variables

γ energy

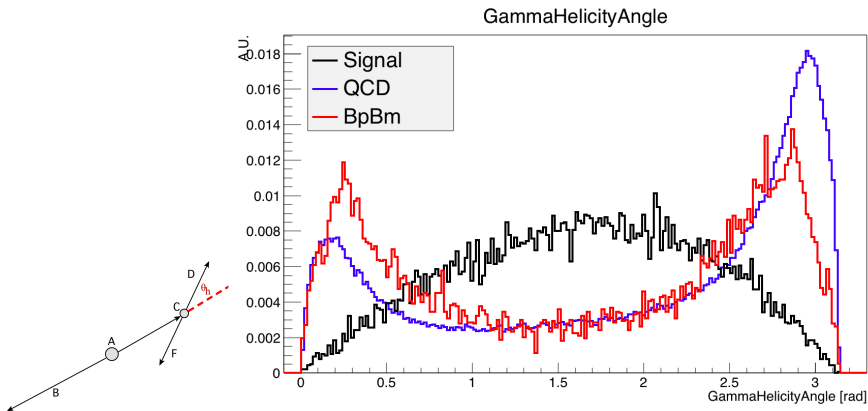
Energy of γ from ALP; will lead to a preselection.



Discriminant variables

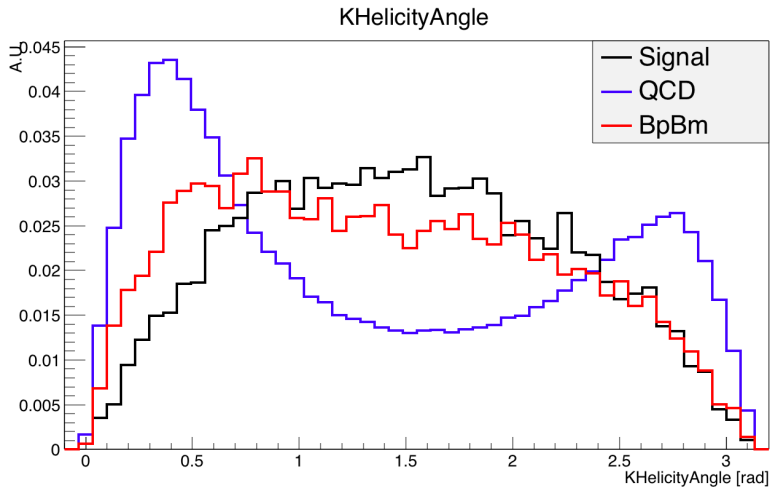
γ helicity angle

Helicity angle of γ s from ALP.



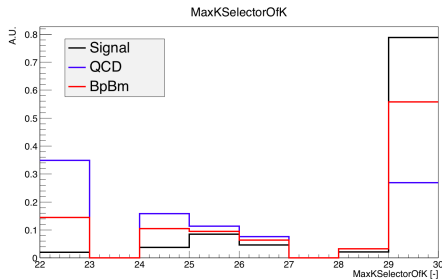
Discriminant variables

K helicity angle



Discriminant variables

K PID selector



"K Selector" stands for "PID Selector".

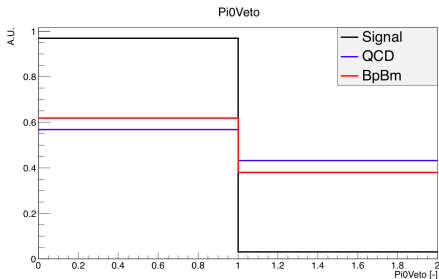
A PID selector is passed if some quality requirements on the track are satisfied.

Highest the number (24-29), highest the probability that the K candidate is really a K.

- 29: SuperTightKMKaonMicroSelection
- 28: VeryTightKMKaonMicroSelection
- 27: TightKMKaonMicroSelection
- 26: LooseKMKaonMicroSelection
- 25: VeryLooseKMKaonMicroSelection
- 24: SuperLooseKMKaonMicroSelection
- 22: None of the above selectors has been passed

Discriminant variables

Pi0Veto



Pi0Veto is to eliminate contamination from true $X = \pi^0, \eta, \eta'$ (i.e. ALP candidates with mass close $\sim$ to X 's).

True $X \rightarrow \gamma_A \gamma_B$: if in an event an ALP candidate is made with either γ_A or γ_B the Pi0Veto is 1.

Powerful to recognize bkg
(Pi0Veto = 1 $\rightsquigarrow$ event is bkg).

Strategy	Method	FoM
1step	MLP	139
	BDT	132
2steps	MLP	130
	BDT	169
3steps	MLP	152
	BDT	133

Table: MVA: comparison strategies and methods. 2 GeV signal used.

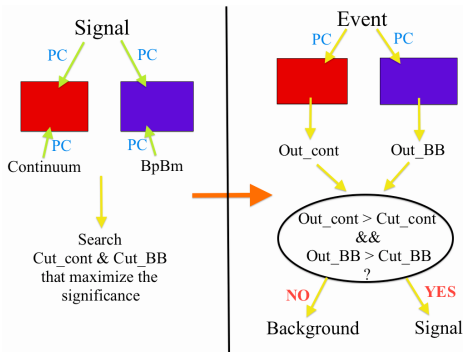
The FoM (Figure of Merit) is directly proportional to the actual significance, computed with continuum and B^+B^- :

$$\text{FoM} = \frac{N_{\text{signal}}^{\text{passed}}}{\sqrt{\sigma_{B^+B^-} \cdot \frac{N_{B^+B^-}^{\text{passed}}}{N_{B^+B^-}^{\text{generated}}} + \sigma_{\text{continuum}} \cdot \frac{N_{\text{continuum}}^{\text{passed}}}{N_{\text{continuum}}^{\text{generated}}}}} \quad (1)$$

MVA

Winner strategy and method - details

- 1 Preselections: ~ 30 MeV away from $X = \pi^0, \eta, \eta'$ masses; threshold for E_γ of 250 MeV;
- 2 $MVA_{\text{continuum}}$: train with Leg0, Leg2, $\cos(\theta_T)$, $\cos(\theta_S)$, MaxKSelectorOfK, m_{ES} , ΔE , γ helicity angle;
- 3 MVA_{B+B^-} : train with $\cos(\theta_T)$, $\cos(\theta_S)$, Pi0Veto, m_{ES} , ΔE , K helicity angle, γ helicity angle.



MVA

Winner strategy and method - MVA outs

Example of the distribution of the MVAs outs, for 1.5 GeV ALP mass training.

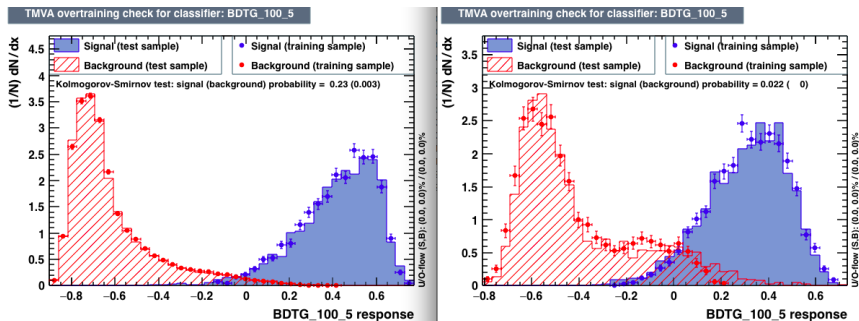
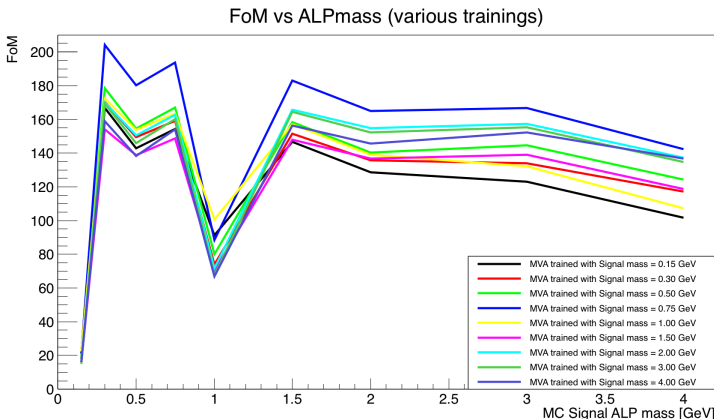


Figure: Left: continuum MVA out. Right: B^+B^- MVA out.

Intermediate value of cuts, interpolated between different masses (see next slide).

Significance slightly decreasing as ALP mass increases.

Drops near π^0 and η' masses.

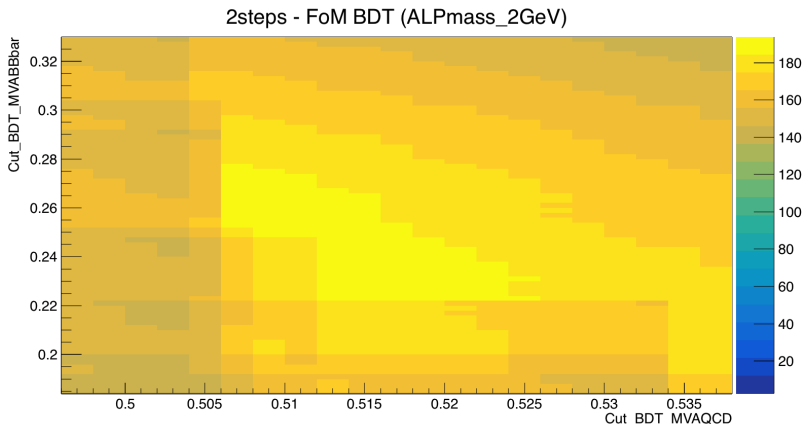


MVA

Significance stability in cuts space

Searched not only point with highest significance but also stable in cuts variations.

Intersection between different MC Signal ALP masses' plots $\Rightarrow$ average cuts ($cut_{continuum} = 0.45$, $cut_{B^+B^-} = 0.16$).

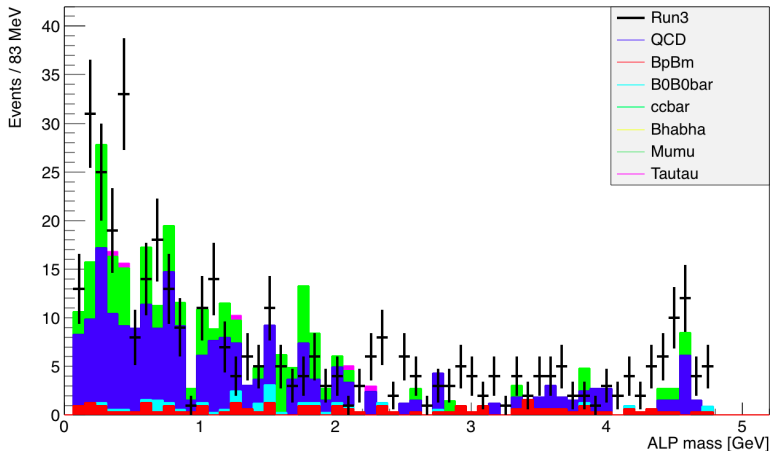


Run3 vs MC bkg

Unblinding Run3

Good agreement between Run3 and MC bkg.

ALP mass spectrum [Run3 vs MC bkg]

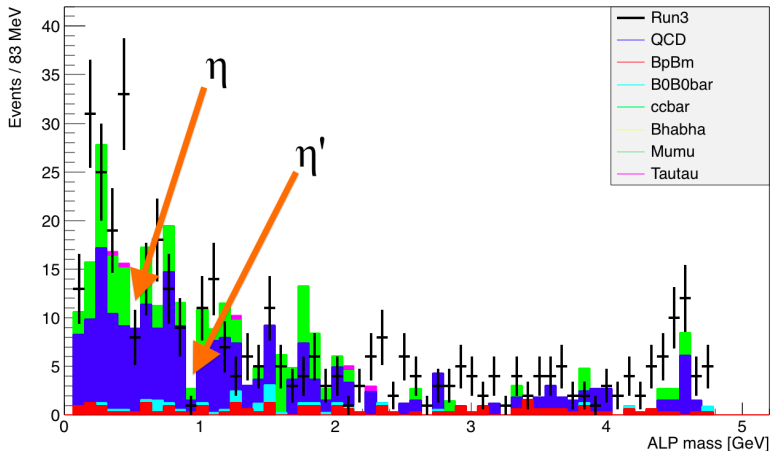


Run3 vs MC bkg

Cut around π^0 , η and η'

We can appreciate the cuts around the vetoed masses.

ALP mass spectrum [Run3 vs MC bkg]

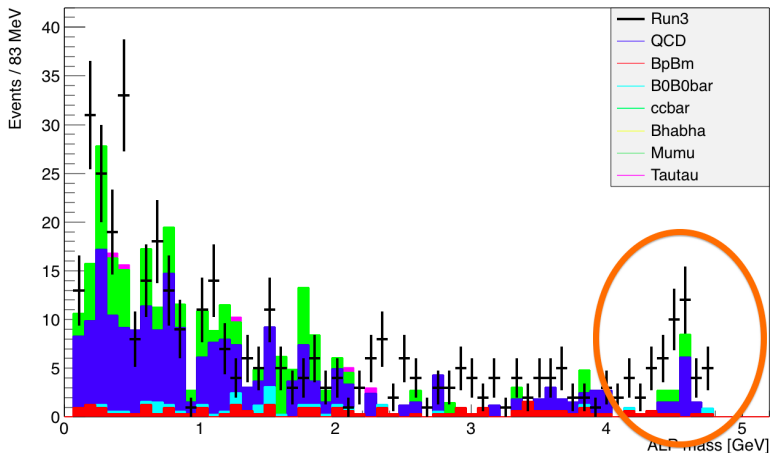


Run3 vs MC bkg

Bump

Bump at end of spectrum. Partially modeled by MC but worth investigation.

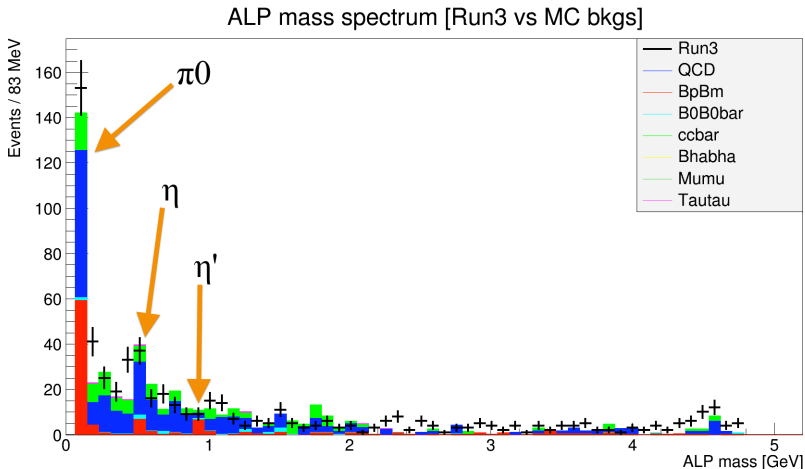
ALP mass spectrum [Run3 vs MC bkg]



Run3 vs MC bkg

Without vetoing the critical masses

Same plot without vetoing the π^0 , η , η' masses.



Run3 vs MC bkg

Causes of the bump

Bump at ~ 4.5 GeV could be caused by a combination of:

- geometrical acceptance;
- m_{ES} requirement.

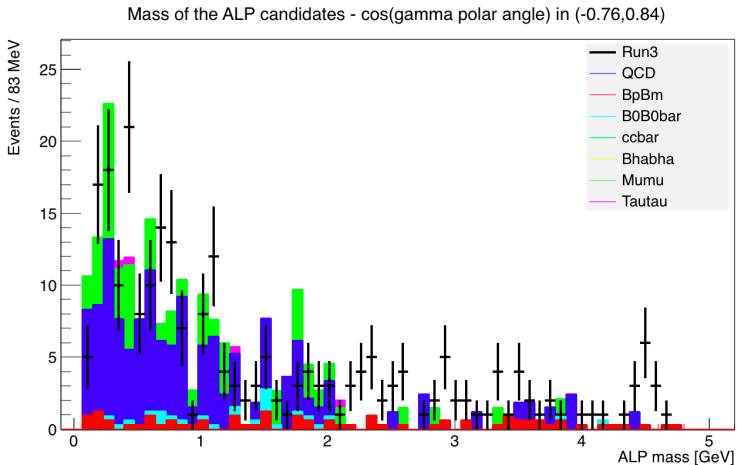
or by some not-well modeled bkg. So we tried the following:

- cut on polar angle of γ from ALP;
- modifying m_{ES} window.

Run3 Bump

γ polar angle cut

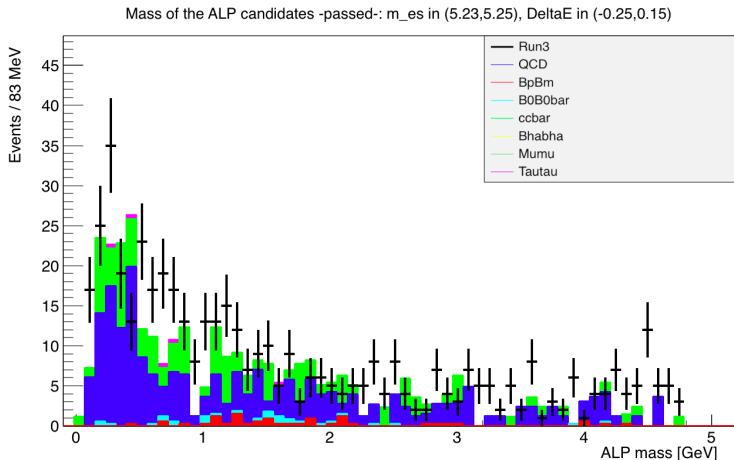
Putting a cut on γ polar angle was not decisive: heavy loss in statistic and bump still present.



Run3 Bump

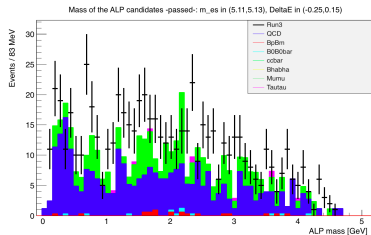
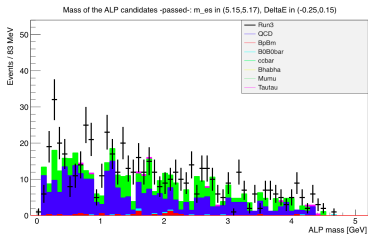
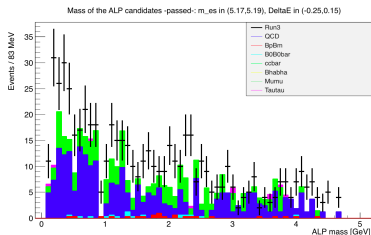
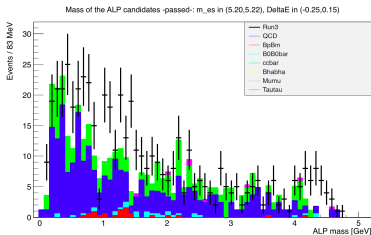
m_{ES} window modification

Here we trained the MVA with the same variables except m_{ES} and ΔE and cut over them at the end.
Bump is reduced.



Run3 Bump

m_{ES} window modification - overview



Signal fit

Gaussian vs Crystal Ball

Fit signal with Gaussian and with **Crystal Ball** (CB), the second being significantly better (see χ^2 and plots).

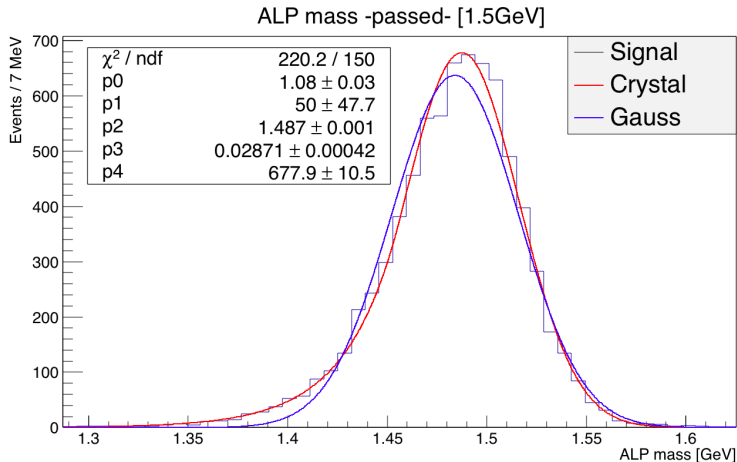
Problem with parameter n of the CB, went too high and fitter crashed (n is an exponent); limited at max 50, has to be fixed.

ALPmass	Gauss χ^2/DoF	CB χ^2/DoF (n max = 50)
0.3	296/132 = 2.24	242/130 = 1.86
0.5	421/125 = 2.27	231/123 = 1.87
0.75	432/140 = 3.08	255/138 = 1.85
1.5	437/152 = 2.88	220/150 = 1.47
2	369/110 = 3.60	151/108 = 1.39
3	228/64 = 3.58	83/62 = 1.34
4	286/66 = 4.34	95/64 = 1.49

Table: Signal fit: Gaus vs CB χ^2_{red} .

Signal fit

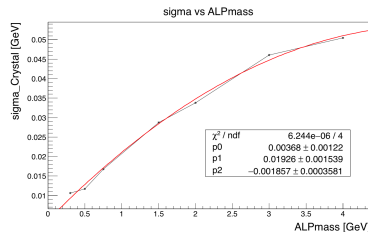
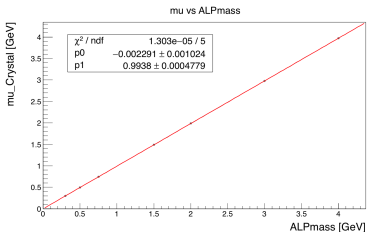
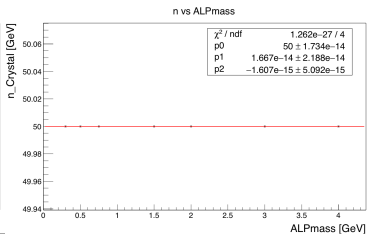
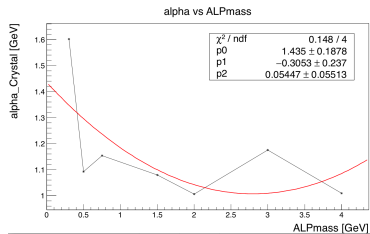
Crystal Ball fit



$$p0 = \alpha, p1 = n, p2 = \mu, p3 = \sigma, p4 = N.$$

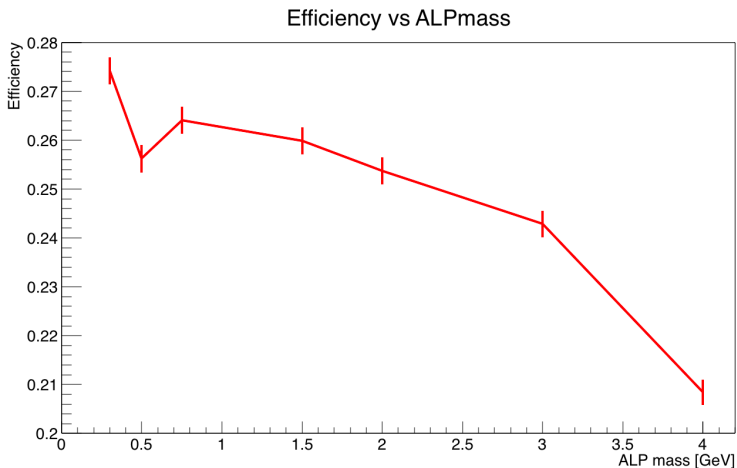
Signal fit

CB parameters fits



Signal efficiency

Using the average MVA cuts we obtain high efficiency:



What's next?

- Finalize the investigation about the bump;
- Fix the CB fitting procedure;
- Use of flattened mass signal to investigate mass-dependent variables (as γ energy or γ polar angle);
- Simulate signal together with bkg to compute yield extraction efficiency;
- Systematic errors.

Thank you!

Backup slides

$$m_{ES} = \sqrt{\left(\frac{\frac{s}{2} + \vec{p}_i \cdot \vec{p}_B}{E_i}\right)^2 - \vec{p}_B^2}$$

With E_i , $\vec{p}_i$ referred to initial four-momentum and $\vec{p}_B$ is the B candidate momentum, both evaluated in the laboratory frame.

$$\Delta_E = E_B - \frac{\sqrt{s}}{2}$$

Where E_B is the energy of the B cand in the CM frame.

Legendre moments

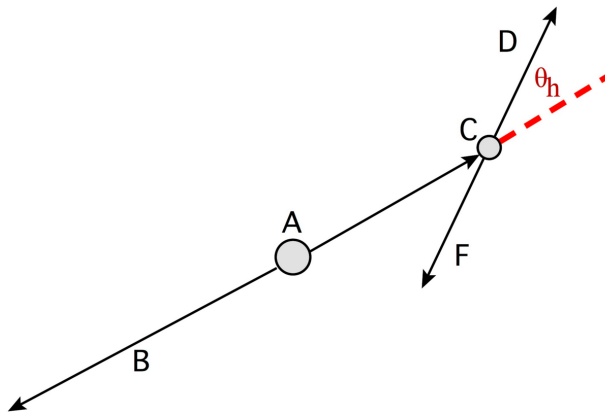
Definition

$$L_k = \sum_{j \in ROE} p_j \cos^k(\theta_j)$$

where θ_j is the angle between B thrust axis and the particle in the sum (ROE = Rest Of the Event).

Helicity angle

Definition



Thrust and Sphericity angles

Definition

The **thrust axis** $\vec{T}$ is the unit vector along which the projection of the total momentum of group of particle is maximized.

θ_T is the angle between the thrust axis of the B candidate and the one of the ROE.

The **sphericity tensor** is given by: $S^{\alpha,\beta} = \frac{\sum_{i=1}^N p_i^\alpha p_i^\beta}{\sum_{i=1}^N |\vec{p}|^2}$ and has 3

eigenvalues and eigenvectors.

Sphericity axis is the versor of the eigenvector with the highest eigenvalue.

θ_S is the angle between the sphericity axis of the B candidate and the one of the ROE.

Crystal Ball

Definition

$$f(x; \alpha, n, \bar{x}, \sigma) = N \cdot \begin{cases} \exp\left(-\frac{(x-\bar{x})^2}{2\sigma^2}\right), & \text{for } \frac{x-\bar{x}}{\sigma} > -\alpha \\ A \cdot \left(B - \frac{x-\bar{x}}{\sigma}\right)^{-n}, & \text{for } \frac{x-\bar{x}}{\sigma} \leq -\alpha \end{cases}$$

where

$$A = \left(\frac{n}{|\alpha|}\right)^n \cdot \exp\left(-\frac{|\alpha|^2}{2}\right),$$

$$B = \frac{n}{|\alpha|} - |\alpha|,$$

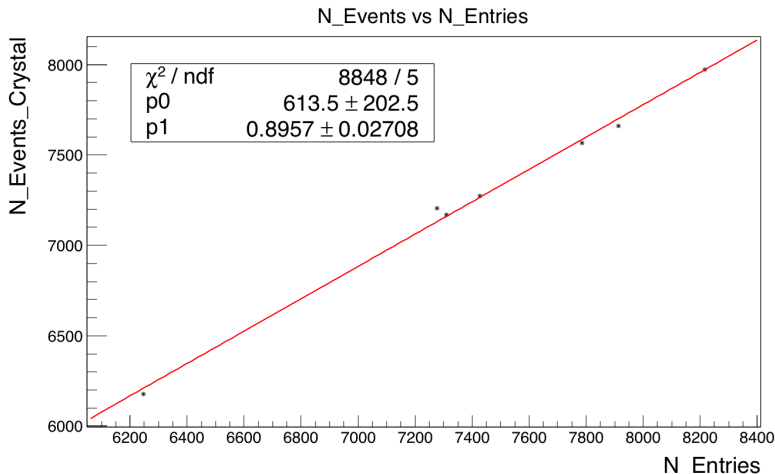
$$N = \frac{1}{\sigma(C+D)},$$

$$C = \frac{n}{|\alpha|} \cdot \frac{1}{n-1} \cdot \exp\left(-\frac{|\alpha|^2}{2}\right),$$

$$D = \sqrt{\frac{\pi}{2}} \left(1 + \operatorname{erf}\left(\frac{|\alpha|}{\sqrt{2}}\right)\right).$$

Crystal Ball

N_{events} vs N_{entries}



The following pre-preselections are applied at the time of ntuple generation:

- K candidate has to be at least a "KCombinedSuperLoose";
- γ candidate has to be at least "GoodPhotonsLoose";
- $\Delta E \in (-0.3; 0.3)$ GeV;
- $m_{ES} \in (5.0; 5.4)$ GeV;
- $m_B \in (4.5; 6.0)$ GeV.

Away from $X = \pi^0, \eta, \eta'$ masses:

$$|m_{ALP} - m_{\pi^0}| < 25 \text{ MeV} \wedge |m_{ALP} - m_{\eta}| < 30 \text{ MeV} \wedge \quad (2)$$

$$\wedge |m_{ALP} - m_{\eta'}| < 35 \text{ MeV} \quad (3)$$

Threshold for E_{γ} :

$$E_{\gamma} > 250 \text{ MeV} \quad (4)$$

MLP stands for Multilayer perceptron, a kind of neural network.
BDT stands for Boosted Decision Tree.

$MVA_{\text{continuum}}$:

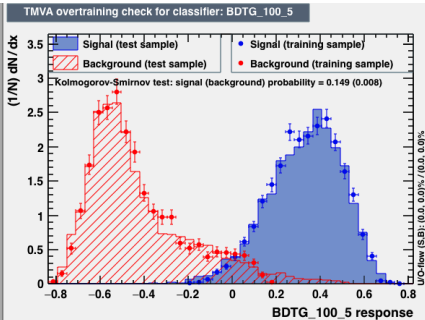
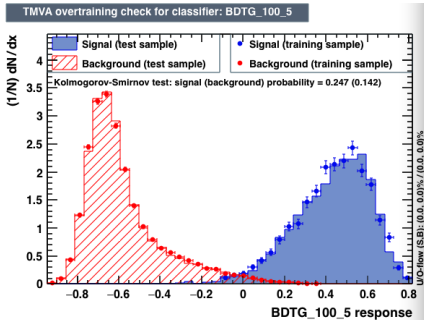
$$MVA_{\text{out}} > 0.45 \quad (5)$$

$MVA_{B^+B^-}$:

$$MVA_{\text{out}} > 0.16 \quad (6)$$

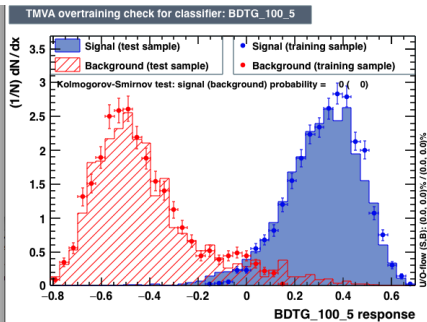
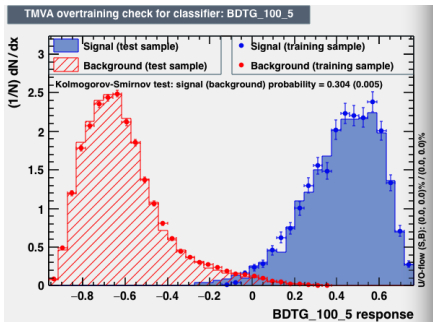
MVA out

0.3 GeV



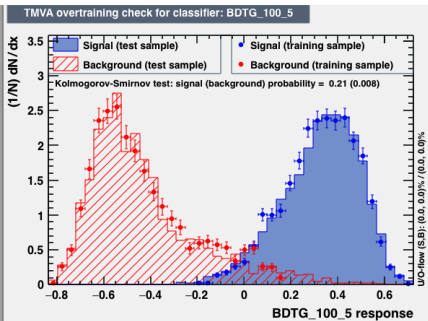
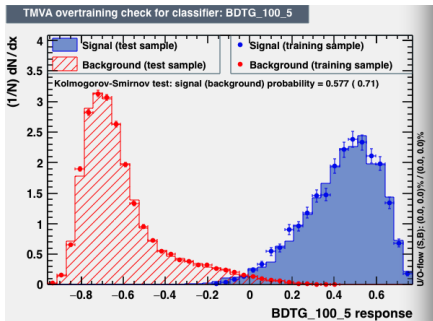
MVA out

0.5 GeV



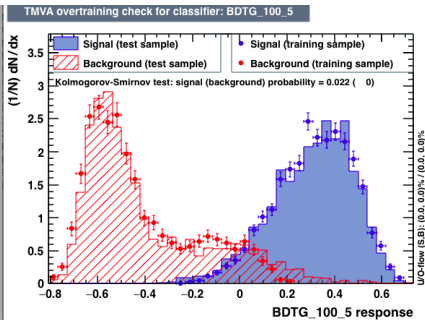
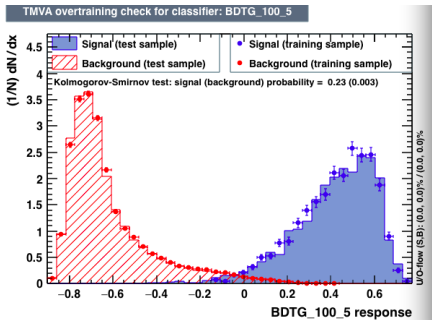
MVA out

0.75 GeV



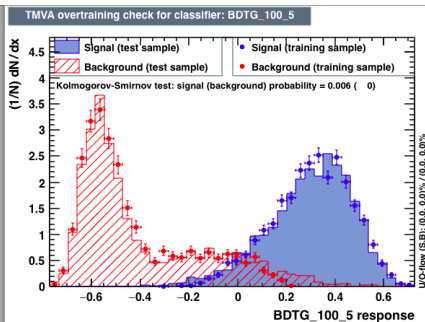
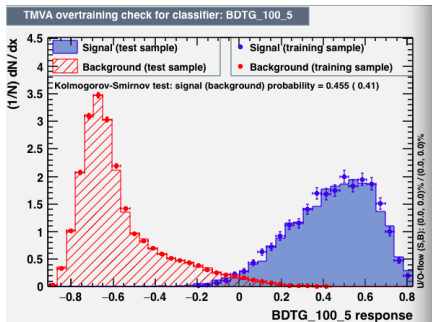
MVA out

1.5 GeV



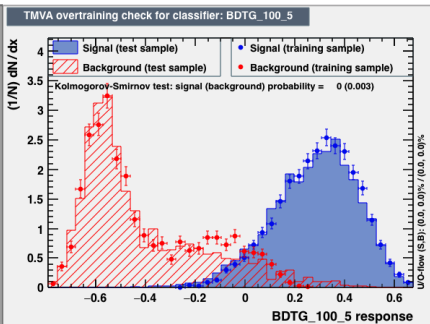
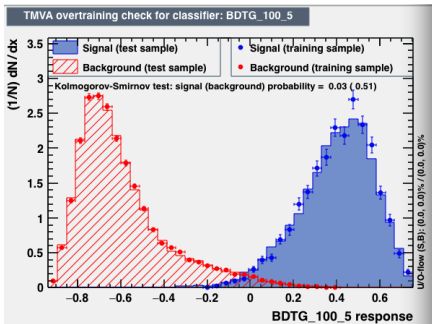
MVA out

2 GeV



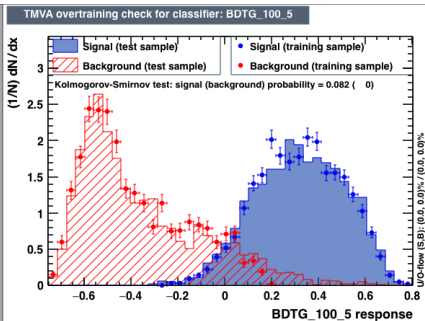
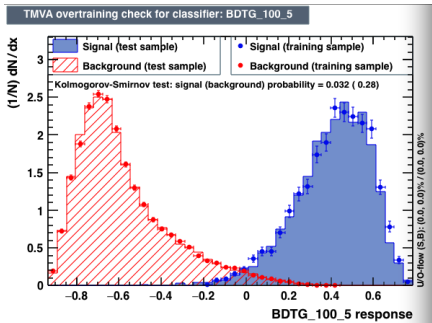
MVA out

3 GeV



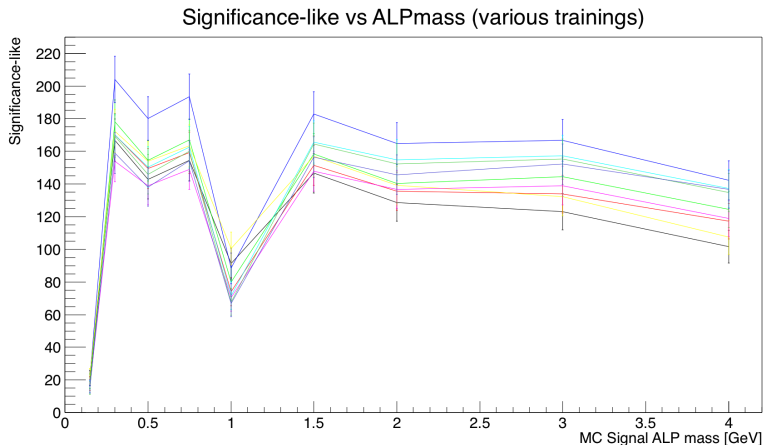
MVA out

4 GeV



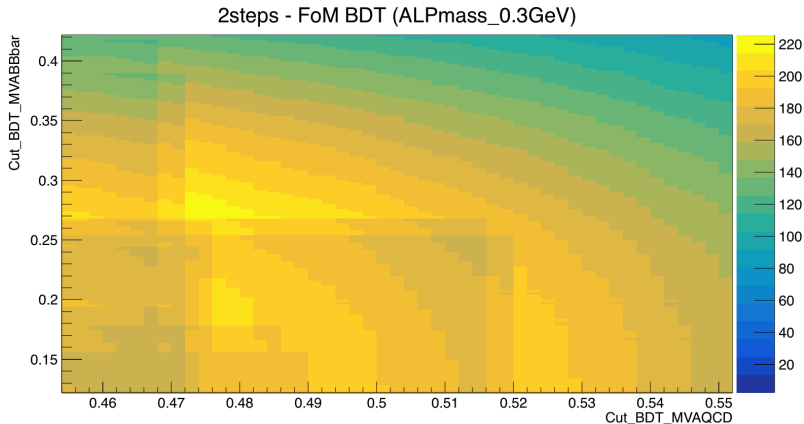
MVA stability for different ALP masses

Uncertainties



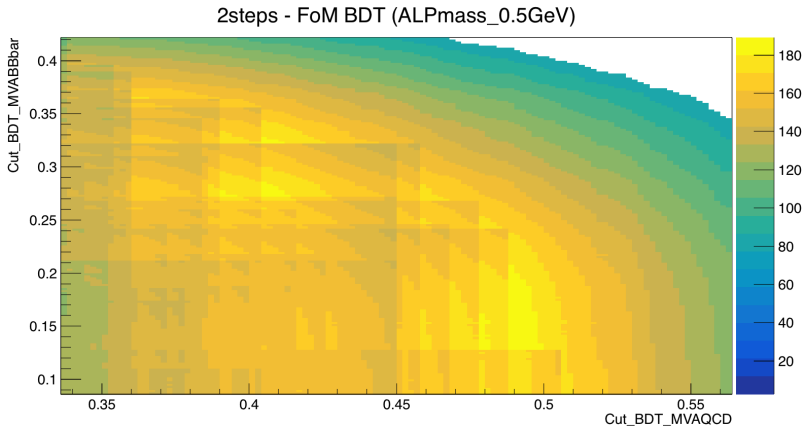
MVA stability in cut space

0.3 GeV



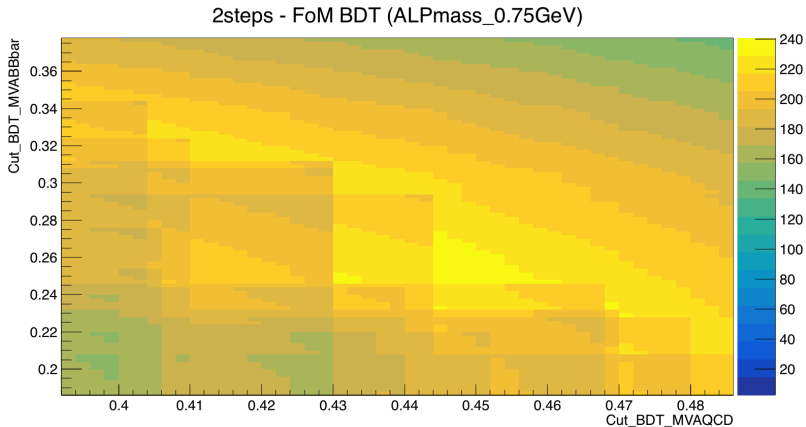
MVA out

0.5 GeV



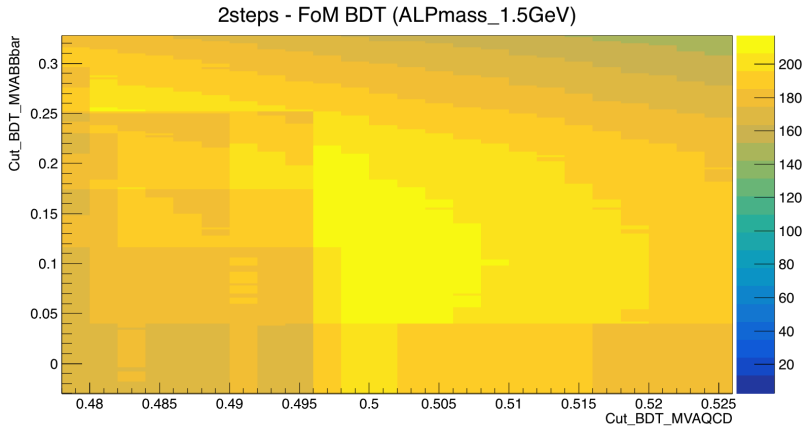
MVA out

0.75 GeV



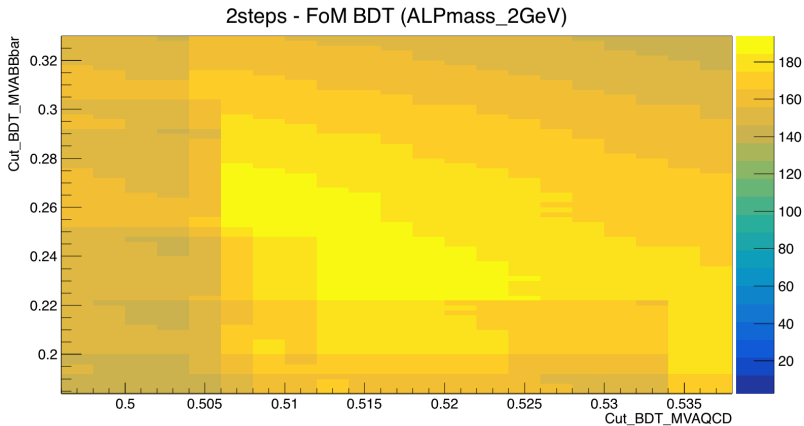
MVA out

1.5 GeV



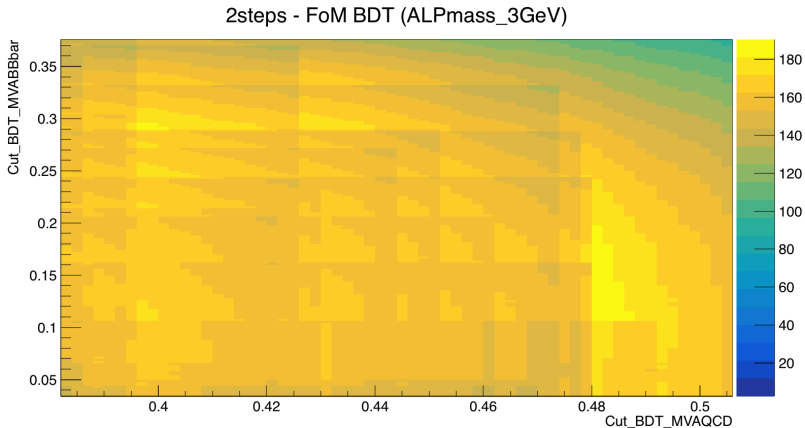
MVA out

2 GeV



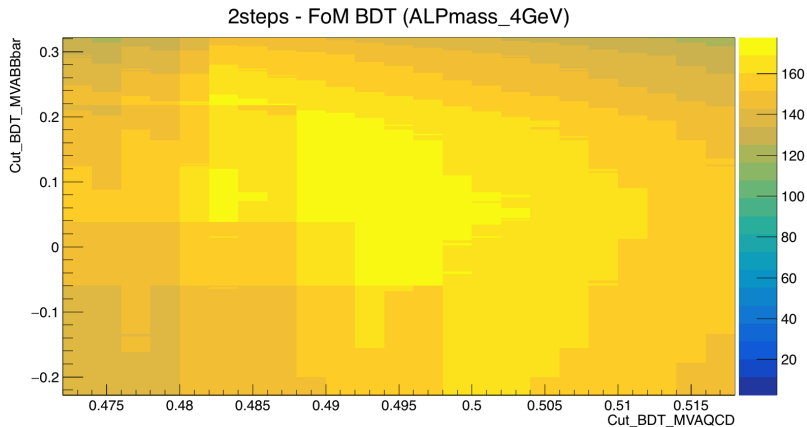
MVA out

3 GeV



MVA out

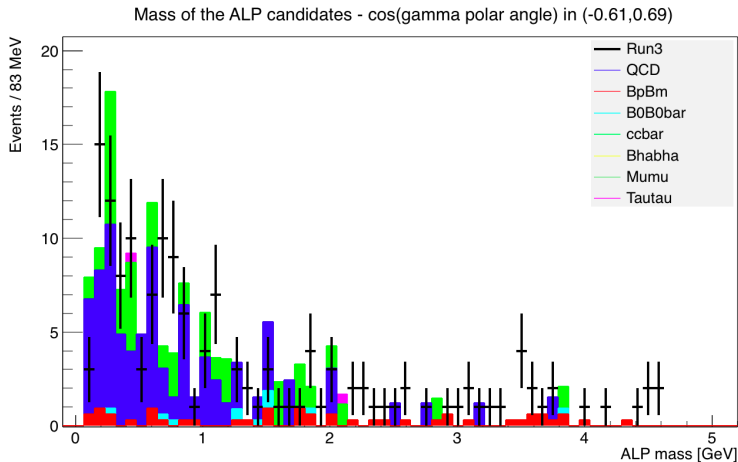
4 GeV



Polar angle investigation

$\cos(\gamma \text{ polar angle})$ in $(-0.61; 0.69)$

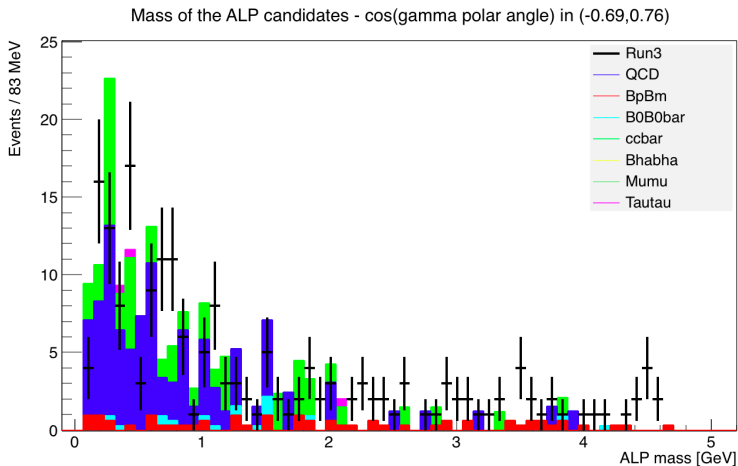
Polar angle of γ s from ALP candidate.



Polar angle investigation

$\cos(\gamma \text{ polar angle})$ in $(-0.69; 0.76)$

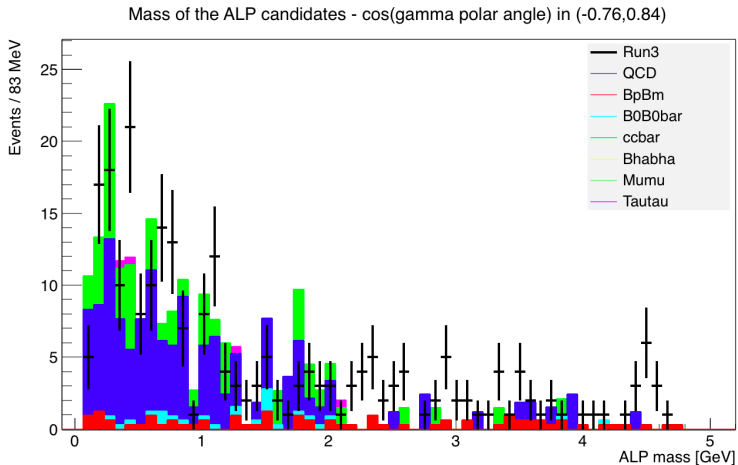
Polar angle of γ s from ALP candidate.



Polar angle investigation

$\cos(\gamma \text{ polar angle})$ in $(-0.76; 0.84)$

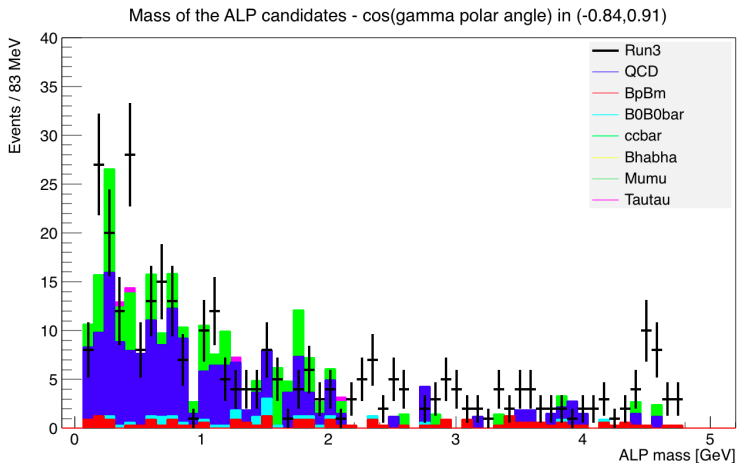
Polar angle of γ s from ALP candidate.



Polar angle investigation

$\cos(\gamma \text{ polar angle})$ in $(-0.84; 0.91)$

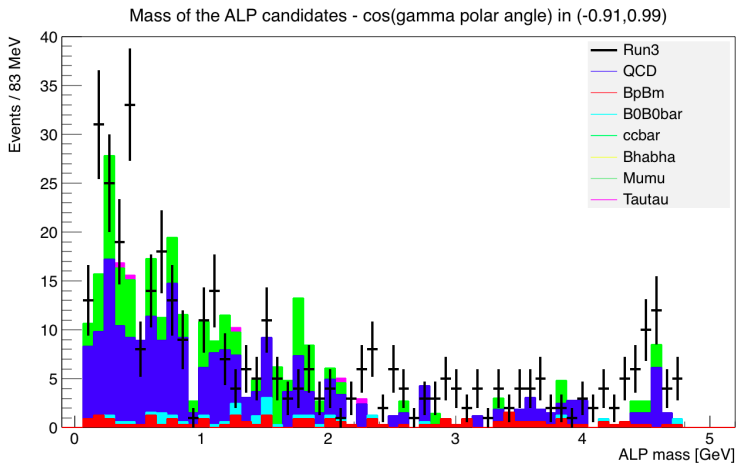
Polar angle of γ s from ALP candidate.



Polar angle investigation

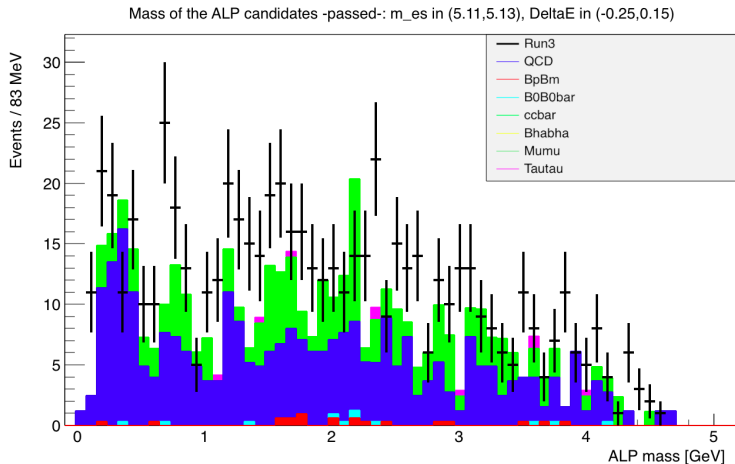
$\cos(\gamma \text{ polar angle})$ in $(-0.91; 0.99)$

Polar angle of γ s from ALP candidate.

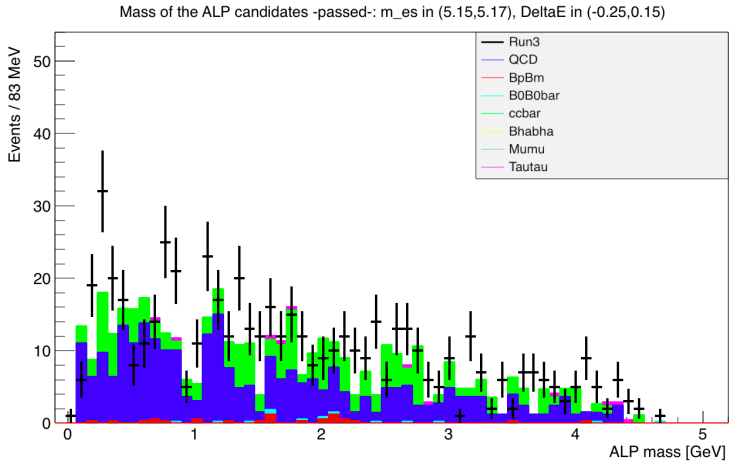


m_{ES} window

5.11 - 5.13

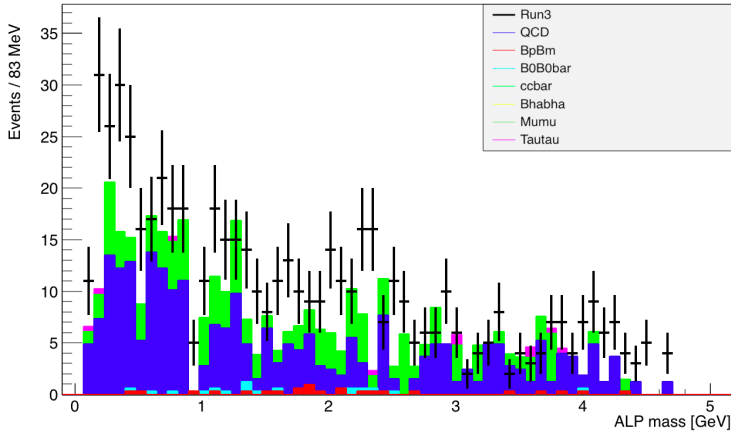


m_{ES} window 5.15 - 5.17



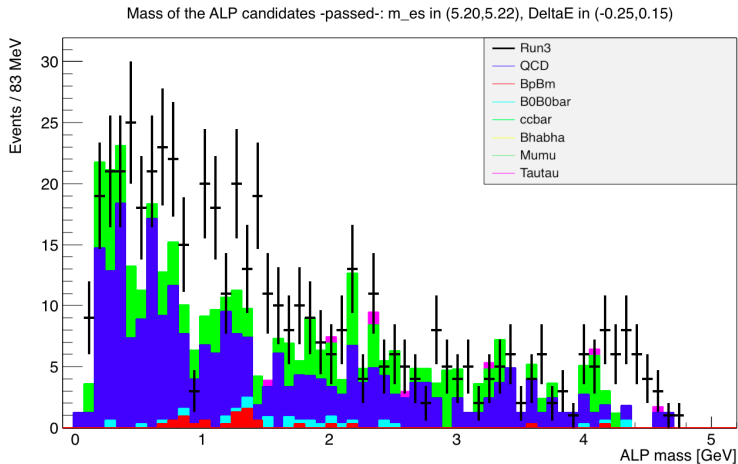
m_{ES} window 5.17 - 5.19

Mass of the ALP candidates -passed-: m_{es} in (5.17,5.19), ΔE in (-0.25,0.15)



m_{ES} window

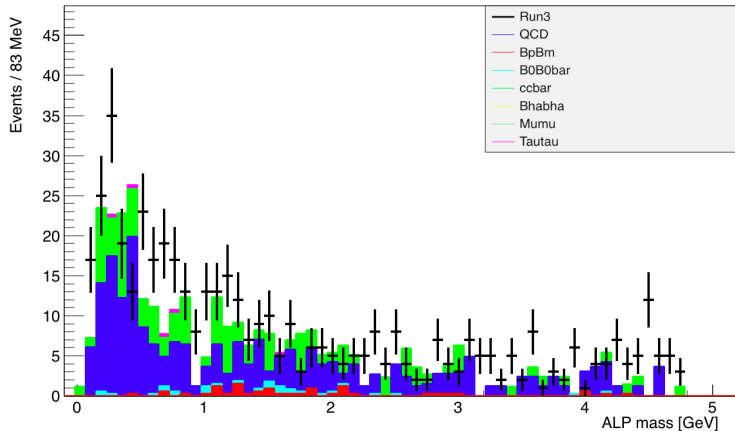
5.20 - 5.22



m_{ES} window

5.23 - 5.25

Mass of the ALP candidates -passed-: m_{es} in (5.23,5.25), ΔE in (-0.25,0.15)



- <https://arxiv.org/pdf/1611.09355.pdf> paper from which the analysis is based on;
- <https://arxiv.org/pdf/0707.2798.pdf> comparison paper for $B \rightarrow K\pi^0$.