

STUDY OF $B_{(S)} \rightarrow (HH) \gamma$ DECAYS

Boris Quintana

LPC Clermont-Ferrand



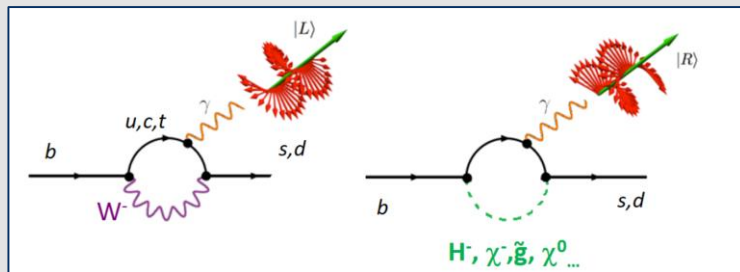


LHCb
~~LHCb~~

YCM01
UX85

RADIATIVE DECAYS

Photons in $b \rightarrow s\gamma$ (FCNC process) are predominantly left-handed in the SM, since the W boson couples to left-handed quarks. New particles could enhance the right-handed contribution !



First observation at CLEO (1993) in $B_d \rightarrow K^* \gamma$

Physical observables :

Branching Ratios :

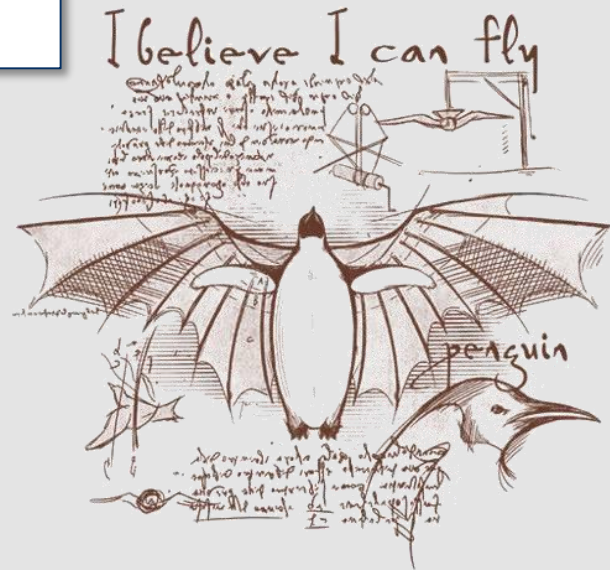
$$BR(B_d \rightarrow K^* \gamma) = \frac{N(B_d \rightarrow K^* \gamma)}{N(B_d)}$$

CP Asymetries :

$$A_{CP} = \frac{N(B_d \rightarrow K^* \gamma) - N(\bar{B}_d \rightarrow \bar{K}^* \gamma)}{N(B_d \rightarrow K^* \gamma) + N(\bar{B}_d \rightarrow \bar{K}^* \gamma)}$$

Photon polarisation : angular analysis / Time dependent decay rate

CKM Matrix elements : $|V_{td}|$, $|V_{ts}|$

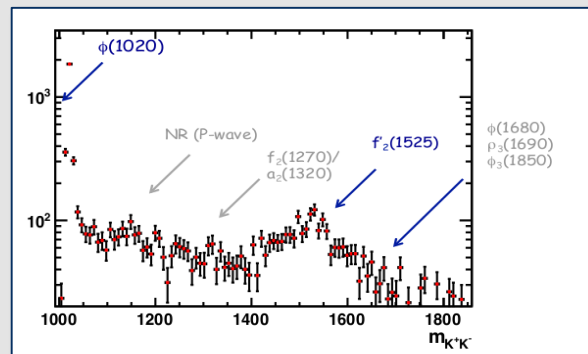
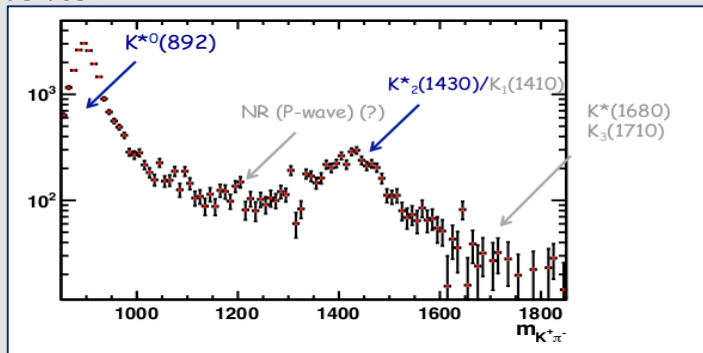


MOTIVATIONS

A simultaneous analysis of $B^0 \rightarrow K\pi\gamma$, $B_s^- \rightarrow KK\gamma$ (and $\Lambda_b^- \rightarrow pK\gamma$) :

Aim to improve $K^*\gamma / \Phi \gamma$ models with better understanding of NR and high-mass contamination + improve inclusive $b \rightarrow s \gamma$ BR in terms of sum over exclusives

h^+h^- spectrum :



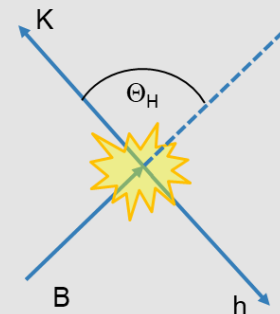
Already observed decay modes :

$B^0 \rightarrow (K^{*0} \rightarrow K^+ \pi^-) \gamma$:	$B_{\text{vis}} = 28.9 \pm 1.0 \times 10^{-6}$:	Cleo, Babar, Belle, LHCb
$B^0 \rightarrow (K^{*2} \rightarrow K^+ \pi^-) \gamma$:	$B_{\text{vis}} = 4.1 \pm 0.8 \times 10^{-6}$:	Babar (5.8σ), Belle (3σ), Cleo
$B_s^- \rightarrow (\varphi \rightarrow K^+ K^-) \gamma$:	$B_{\text{vis}} = 18 \pm 2 \times 10^{-6}$:	Belle, LHCb

Some limits on other modes :

$B^0 \rightarrow (K^* (1410) \rightarrow K^+ \pi^-) \gamma$:	$B_{\text{vis}} < 9 \times 10^{-6}$: Belle
$B^0 \rightarrow (K^+ \pi^-) \text{NR} \gamma$	$B_{\text{vis}} < 2.6 \times 10^{-6}$: Belle ($m_{K\pi}$ in [1.25, 1.60] GeV/ c^2)

Helicity angle

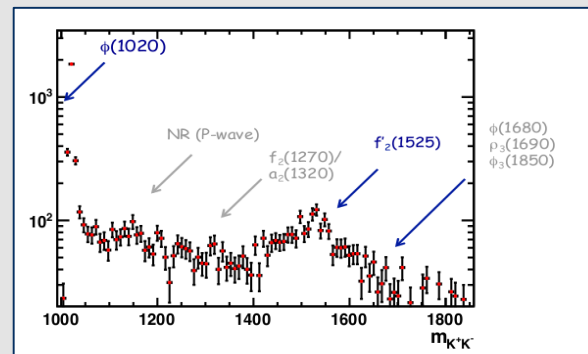
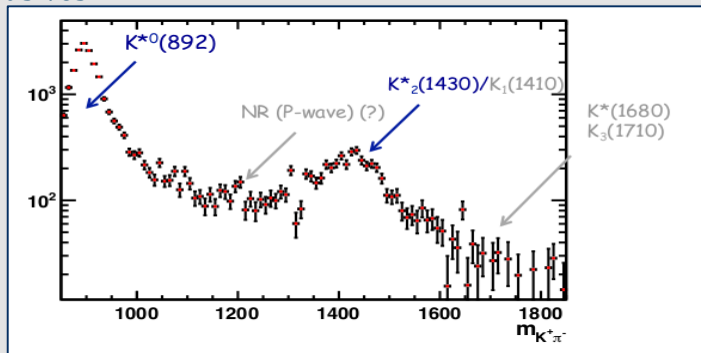


MOTIVATIONS

A simultaneous analysis of $B^0 \rightarrow K\pi\gamma$, $B_s^- \rightarrow KK\gamma$ (and $\Lambda_b^- \rightarrow pK\gamma$) :

Aim to improve $K^*\gamma / \Phi \gamma$ models with better understanding of NR and high-mass contamination + improve inclusive $b \rightarrow s \gamma$ BR in terms of sum over exclusives

h^+h^- spectrum :

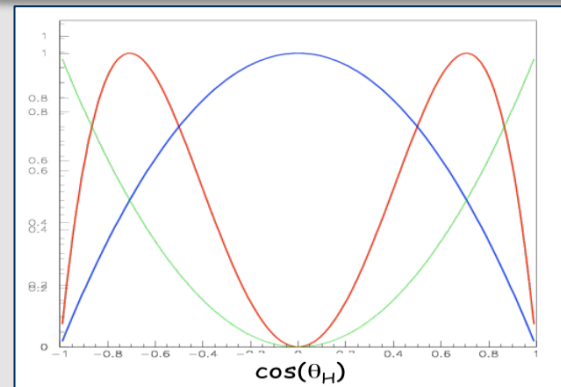


Already observed decay modes :

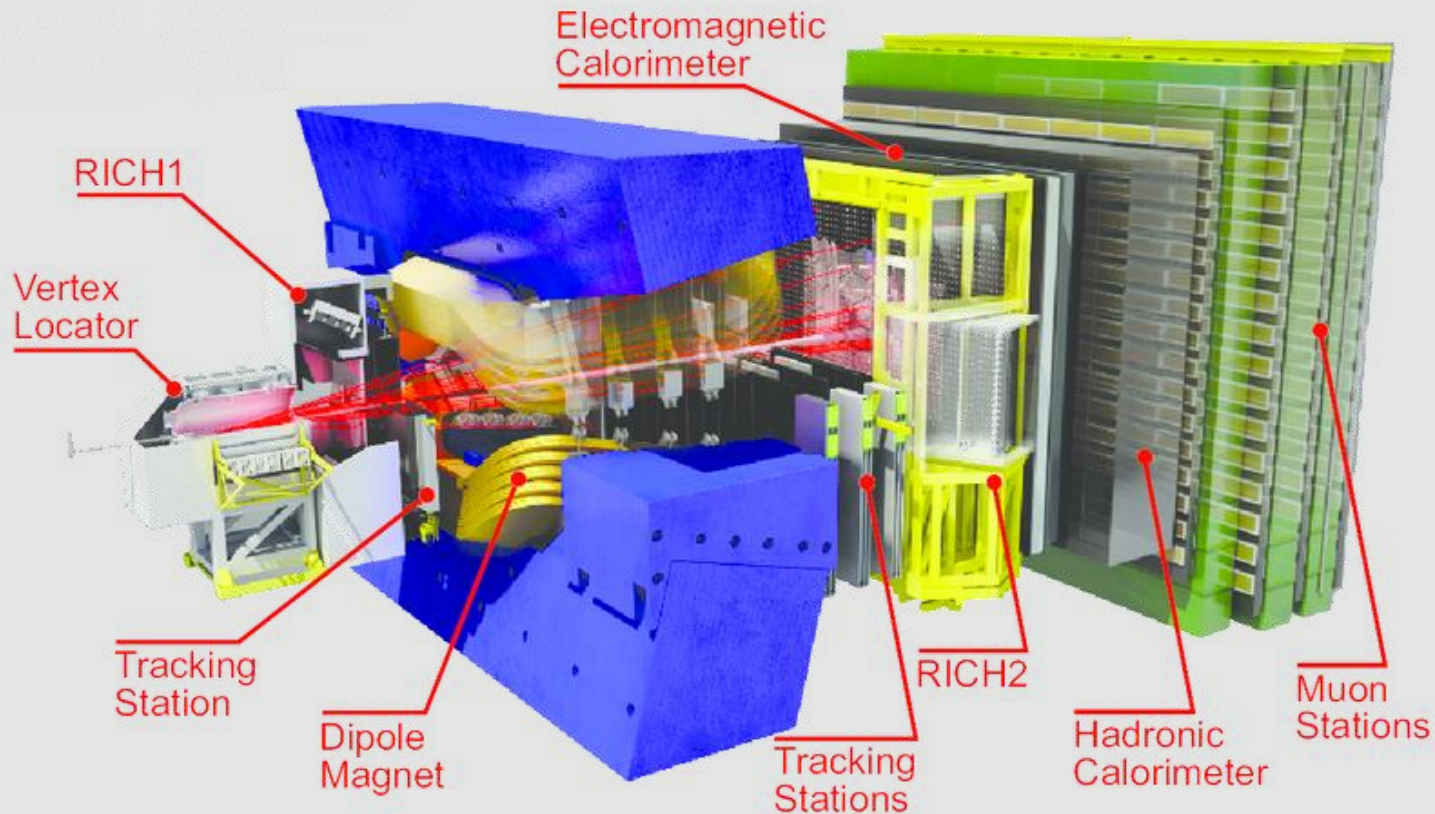
$$\begin{array}{lll}
 B^0 \rightarrow (K^{*0} \rightarrow K^+ \pi^-) \gamma : & B_{\text{vis}} = 28.9 \pm 1.0 \times 10^{-6} : & \text{Cleo, Babar, Belle, LHCb} \\
 B^0 \rightarrow (K^{*2} \rightarrow K^+ \pi^-) \gamma : & B_{\text{vis}} = 4.1 \pm 0.8 \times 10^{-6} : & \text{Babar (5.8}\sigma\text{), Belle(3}\sigma\text{), Cleo} \\
 B_s^- \rightarrow (\varphi \rightarrow K^+ K^-) \gamma : & B_{\text{vis}} = 18 \pm 2 \times 10^{-6} : & \text{Belle, LHCb}
 \end{array}$$

Some limits on other modes :

$$\begin{array}{ll}
 B^0 \rightarrow (K^* (1410) \rightarrow K^+ \pi^-) \gamma & : B_{\text{vis}} < 9 \times 10^{-6} : \text{Belle} \\
 B^0 \rightarrow (K^+ \pi^-) \text{NR} \gamma & B_{\text{vis}} < 2.6 \times 10^{-6} : \text{Belle (} m_{K\pi} \text{ in [1.25, 1.60] GeV/c}^2\text{)}
 \end{array}$$



LHCb DETECTOR



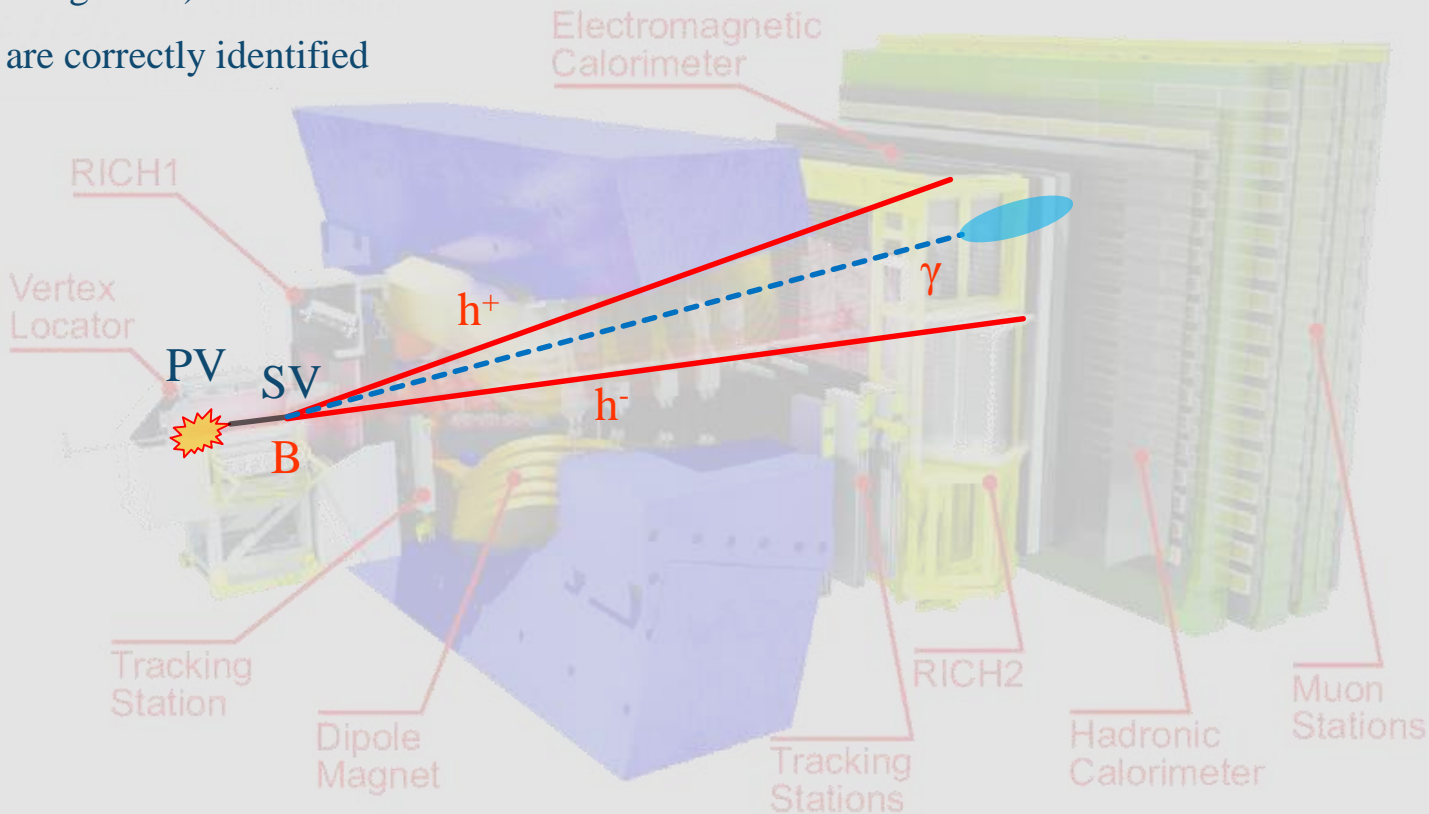
SIGNAL AND BACKGROUNDS

Signal (and Physical background) :

- final states particles are correctly identified

$$B_d \rightarrow K^* \gamma$$

$$B_s \rightarrow K^* \gamma$$



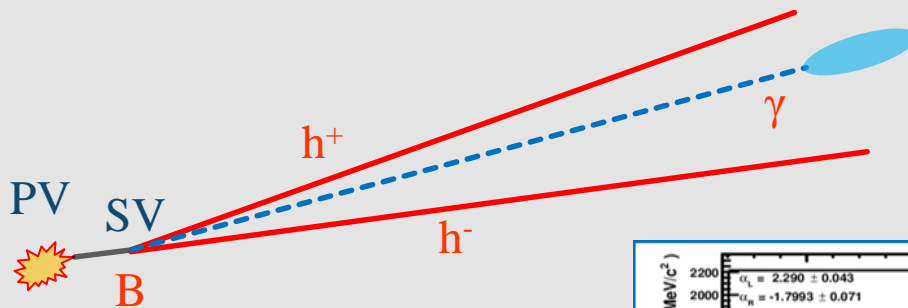
SIGNAL AND BACKGROUNDS

Signal (and Physical background) :

- final states particles are correctly identified

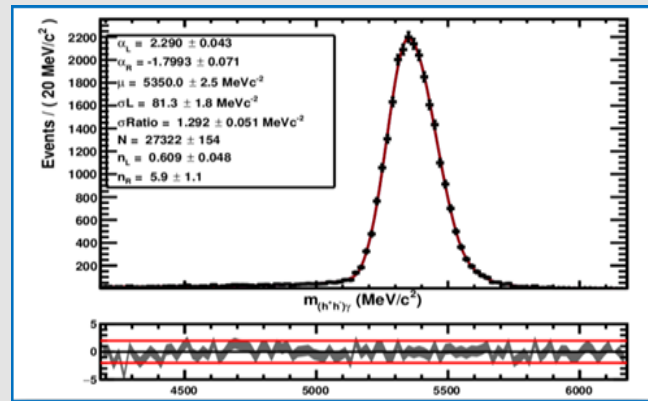
$$B_d \rightarrow K^* \gamma$$

$$B_s \rightarrow K^* \gamma$$



Expected yields and shape of the two contributions are extracted from Monte Carlo simulation :

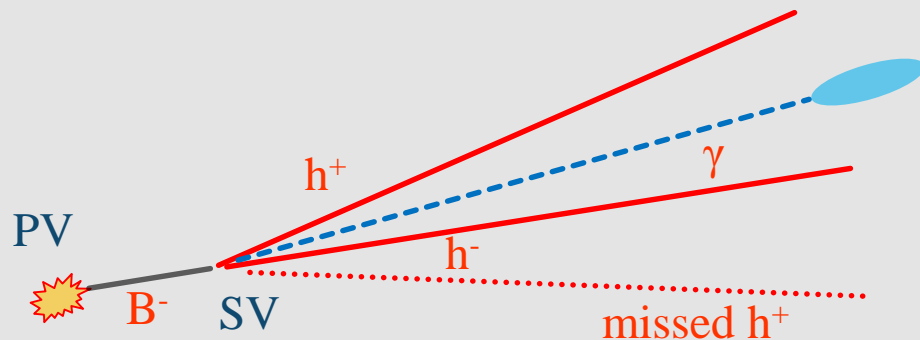
Contamination : 0,8 %



BACKGROUNDS

Partially reconstructed :

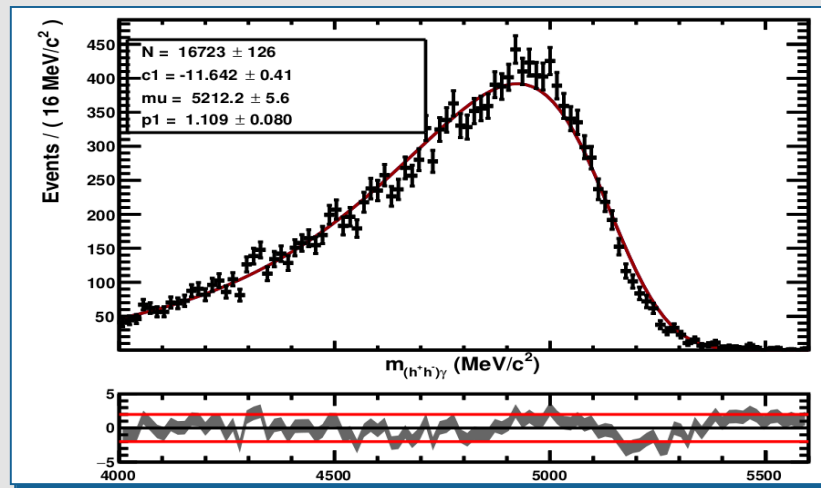
Ex : $B^+ \rightarrow K^{l+} \gamma \rightarrow K^+ \pi^- \pi^+ \gamma$



Expected yields and shape are extracted from

Monte Carlo simulation :

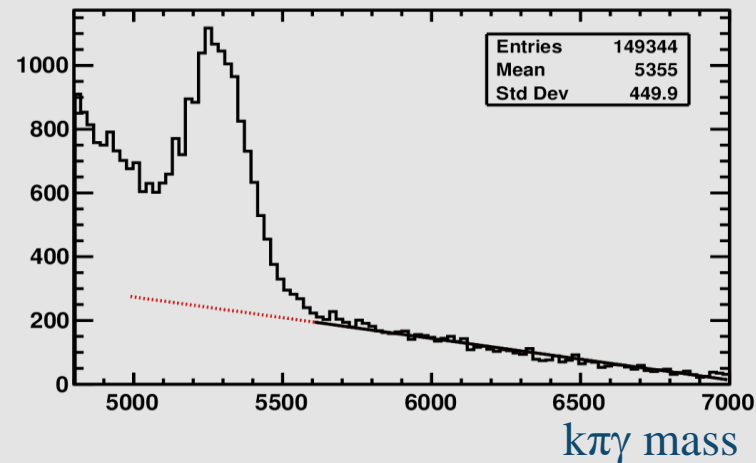
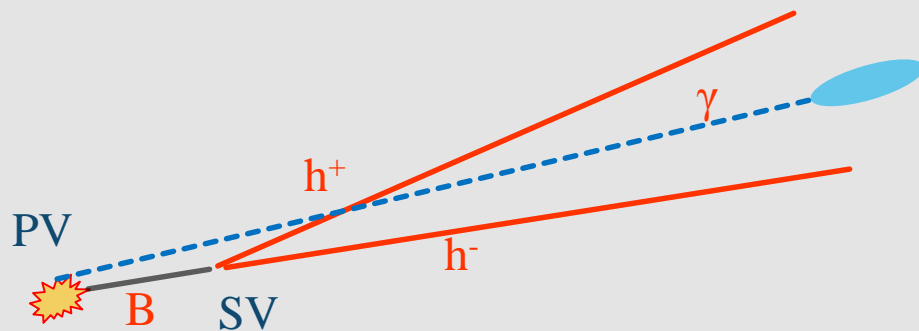
Contamination : $\sim 1\%$



BACKGROUNDS

Combinatorial background :

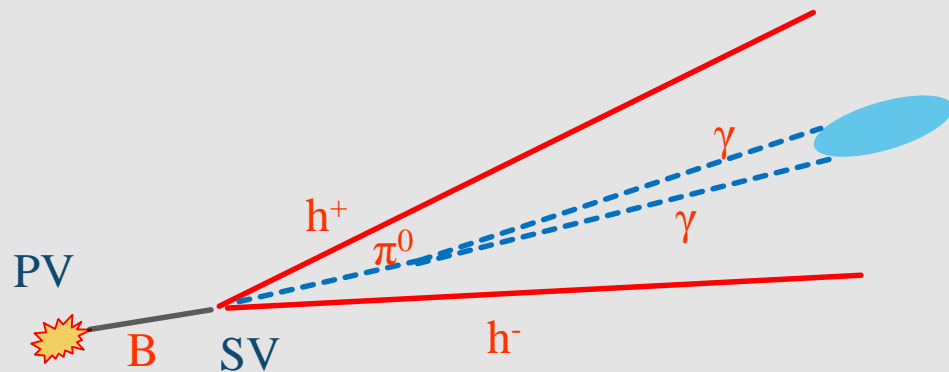
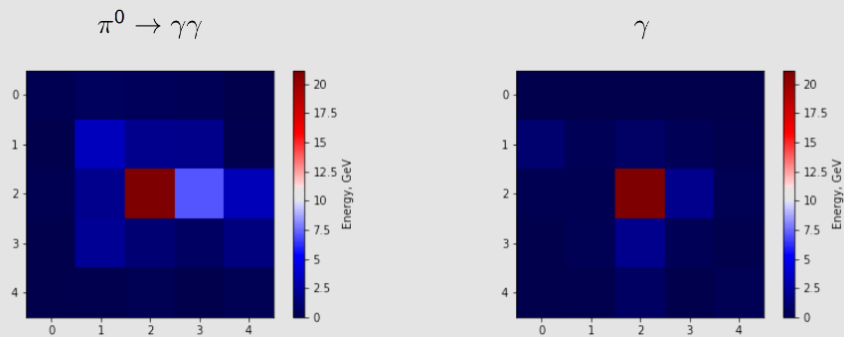
Final state particles don't all come from the (same) B decay



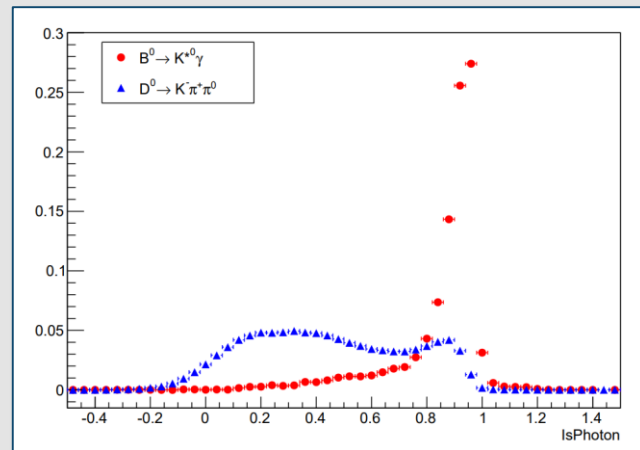
BACKGROUNDS

γ/π^0 Mis-identification :

$\pi^0 \rightarrow \gamma\gamma$ looks like a single γ in the ECAL :



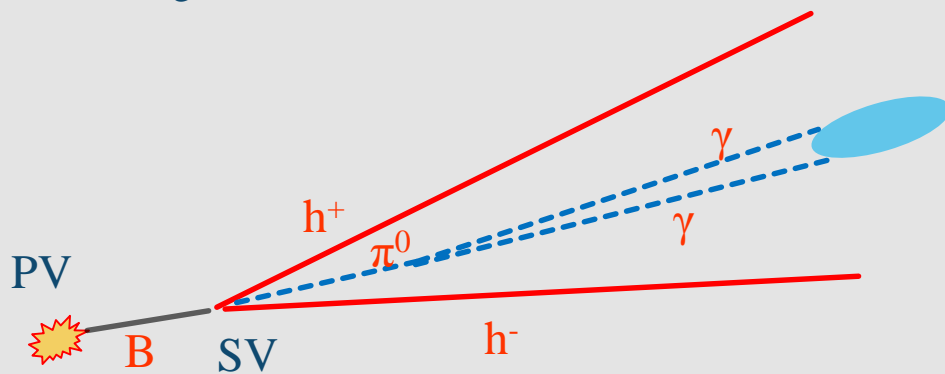
We use a Neural Network trained to distinguish between the two :



BACKGROUNDS

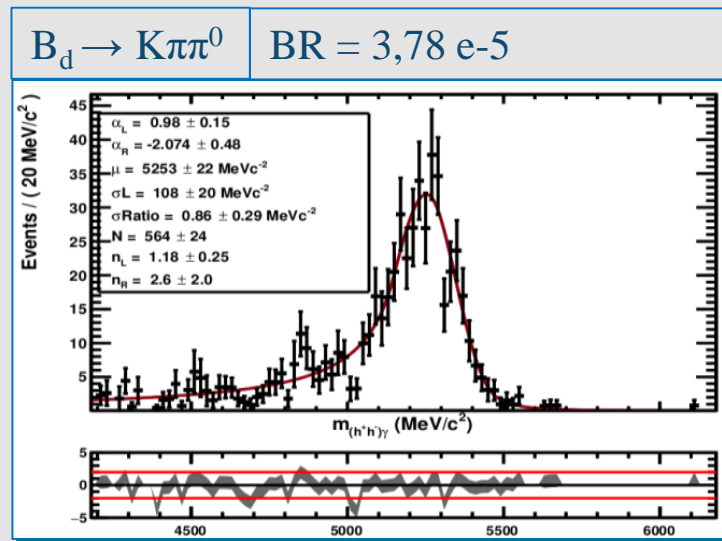
$\gamma \rightarrow \pi^0$ Mis-identification :

Peaks at the signal mass !



IsPhoton $> 0,6$: $\sim 80\%$ background rejection

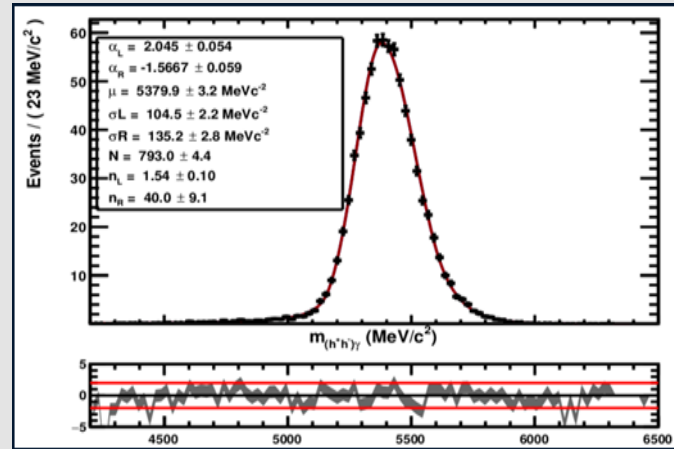
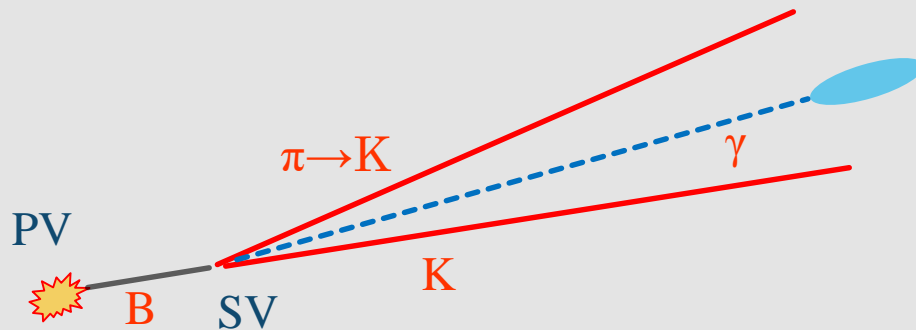
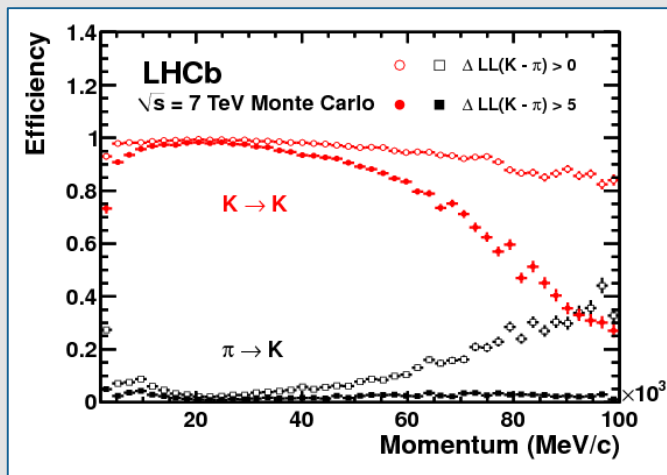
Contamination : 2-5%



BACKGROUNDS

Charged tracks Mis-identification :

Peaks under the signal !



Contamination :

- 0,01 % for double mis-id
- From 1 to 10% for simple Mis-ID

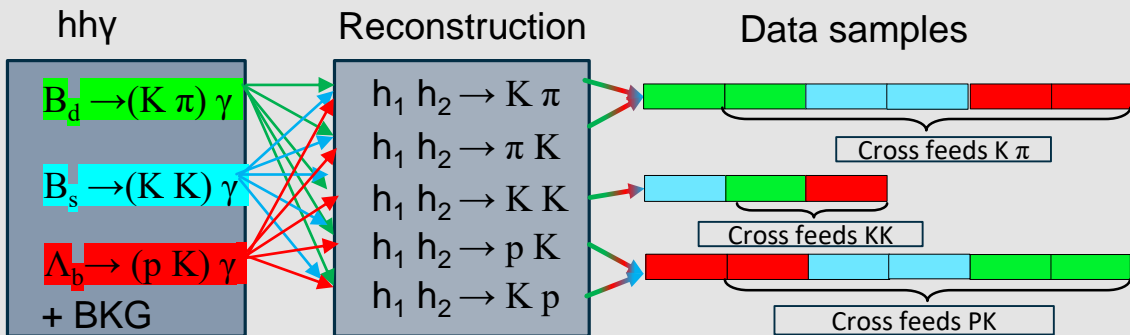
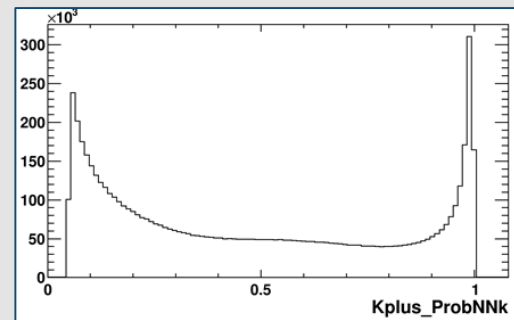
ANALYSIS SCHEME

After gathering all hhy events :

Need to build exclusive samples \rightarrow PID selection

Classify $h^+ h^- \gamma$ candidates according to largest h-ProbNN x_i ($x_i = \pi, K, p$)

$h = x_i$ if $\text{probNN}_{x_i}(h) > 0.1$ & $\text{probNN}_{x_i}(h) > \text{probNN}_{x_j}(h)$ for all $x_j \neq x_i$



PID FOM

The FoM computed for each (x_{h1}, x_{h2}) cut values is : $\text{FoM} = \frac{S}{\sqrt{S+B}}$

With :

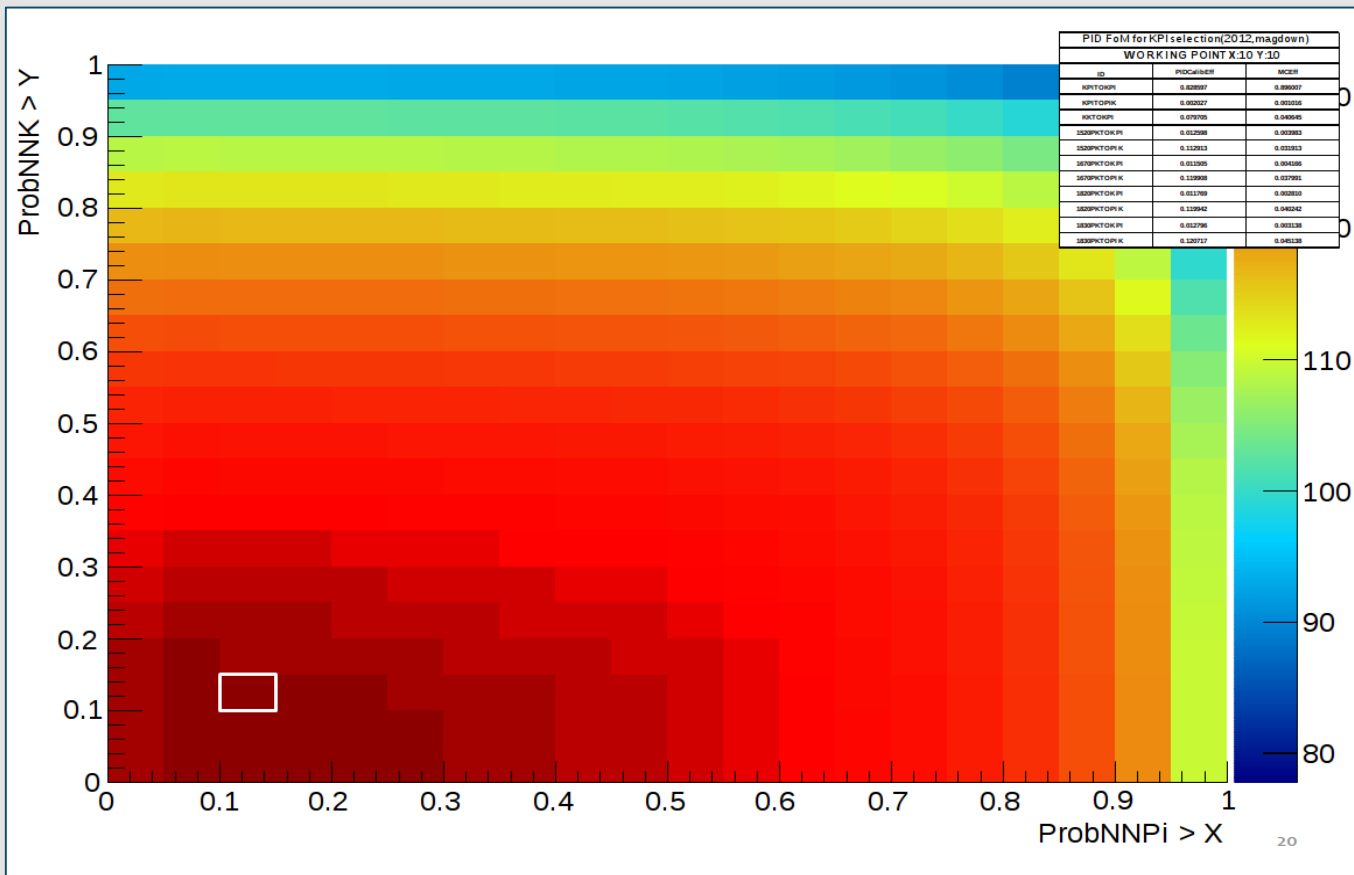
$$S = L * 2 * \sigma(\text{bb}) * f_{b \rightarrow (B_0/B_s/\Lambda_b)} * \text{BR}(B \rightarrow \text{hh}\gamma) * \epsilon_{\text{MC}} * \epsilon_{\text{preSel}} * \epsilon_{\text{PID}}(x_{h1}, x_{h2})$$

$$B = \sum^{\text{cross feeds}} L * 2 * \sigma(\text{bb}) * f_{b \rightarrow (B_0/B_s/\Lambda_b)} * \text{BR}(B \rightarrow [\text{hh}]_{\text{trueID}}\gamma) * \epsilon_{\text{MC}} * \epsilon_{\text{preSel}} * \epsilon_{\text{misID}}(x_{h1}, x_{h2})$$

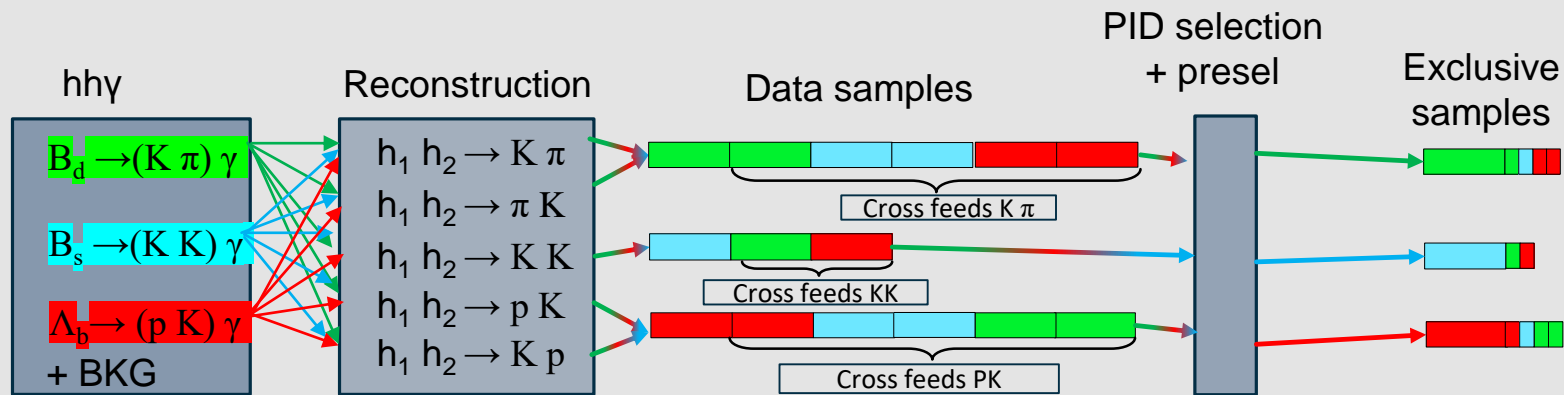
This FoM has been drawn for a scan performed by steps of 0,05 in ProbNN.

The $\epsilon_{\text{PID}}(x_{h1}, x_{h2})$ is taken from calibration data.

PID FOM FOR KPI SELECTION

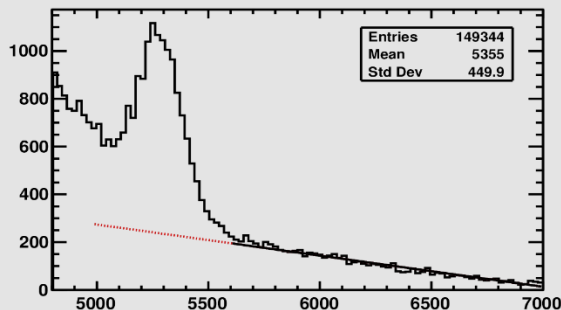


ANALYSIS SCHEME



After the PID selection, a large contribution of combinatorial background remains :

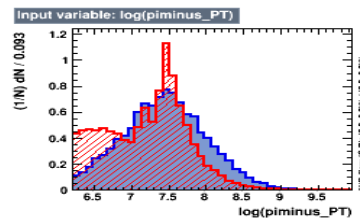
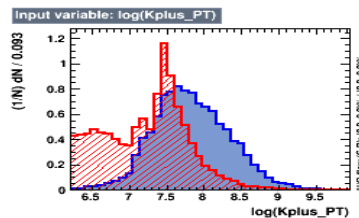
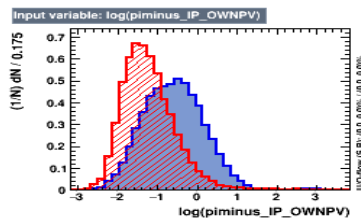
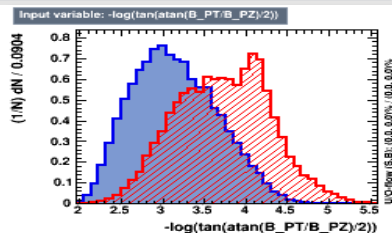
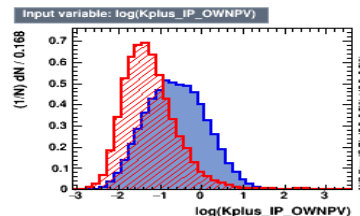
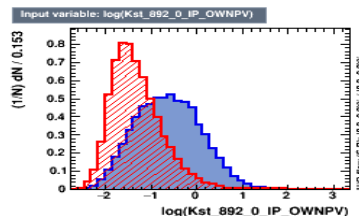
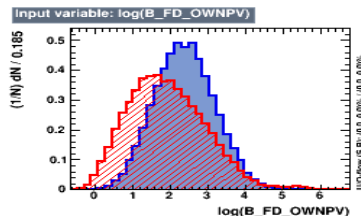
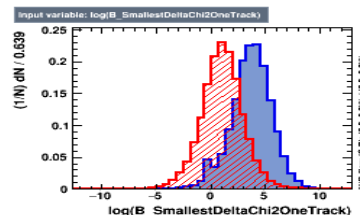
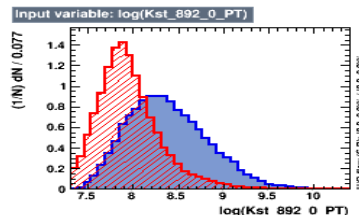
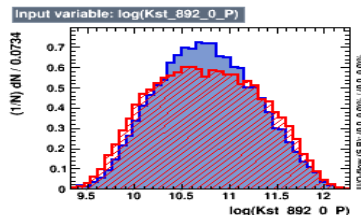
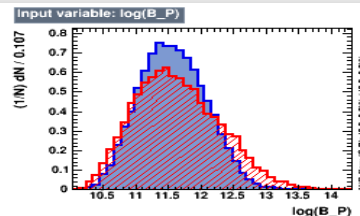
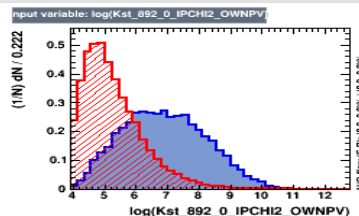
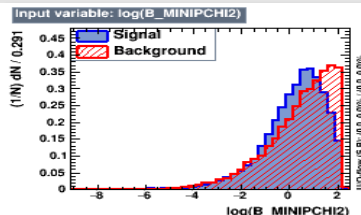
- We use a Booster Decision Tree (Machine Learning) to reduce it !



BDT VARIABLES

Final set :

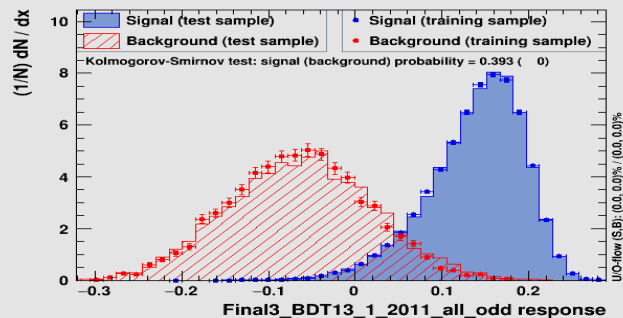
- $B \min \chi_{IP}^2$
- B pseudorapidity
- B momentum
- B Flight Distance
- B vertex isolation (Smallest $\Delta\chi^2$)
- χ_{IP}^2 of the (hh) resonance
- IP of the resonance
- Momentum of the resonance
- IP of the tracks
- Transverse momenta of the tracks



BDT PERFORMANCES

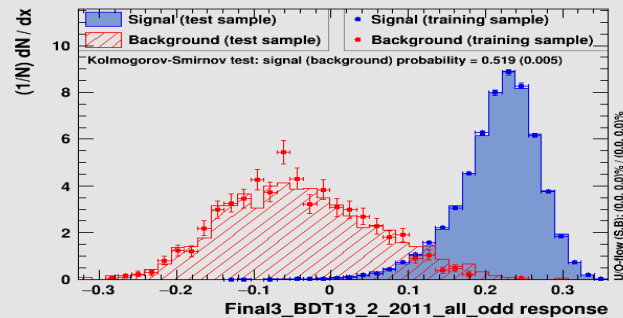
$K\pi\gamma$

TMVA overtraining check for classifier: Final3_BDT13_1_2011_all_odd

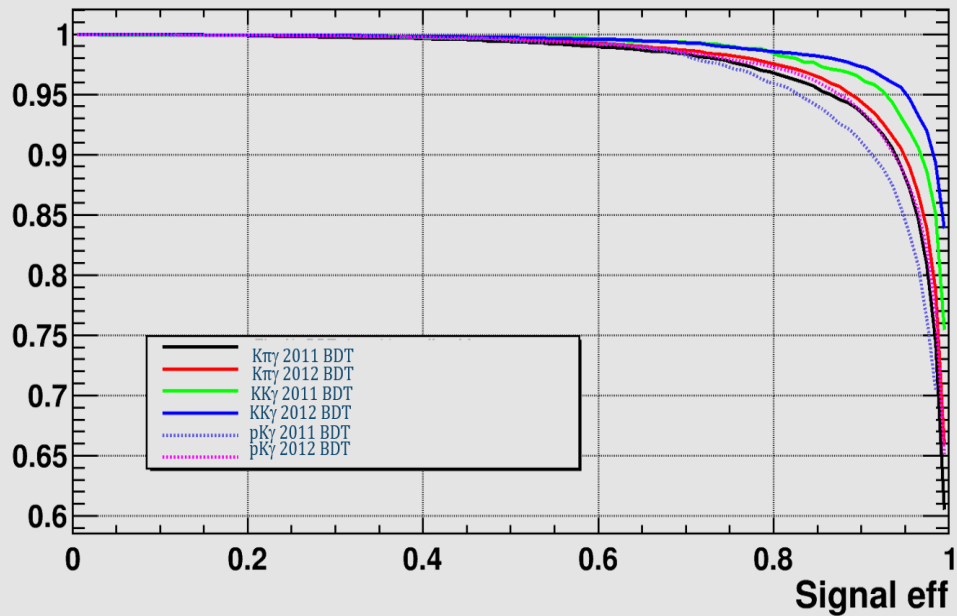


$KK\gamma$

TMVA overtraining check for classifier: Final3_BDT13_2_2011_all_odd



Backgr rejection (1-eff)



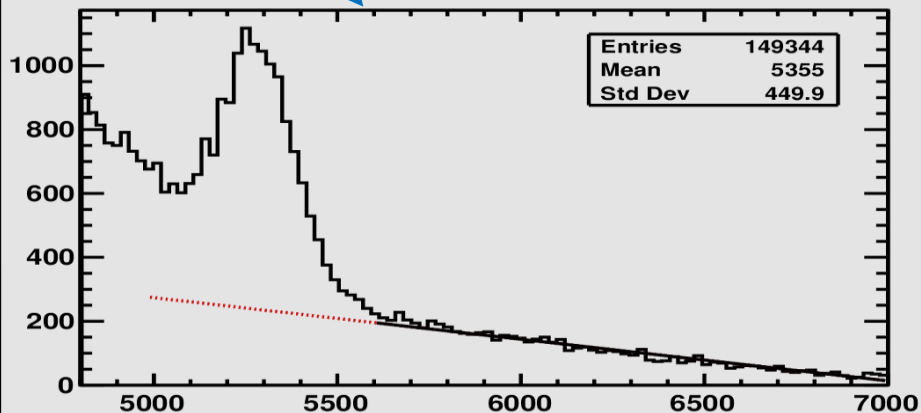
OPTIMAL CUT

$$\text{FoM} = \frac{S}{\sqrt{S+B}}$$

$$\text{With } S = L * 2 * \sigma(\text{bb}) * f_{\text{b} \rightarrow (\text{B}^0/\text{B}_s/\Lambda\text{b})} * \text{BR}(\text{B} \rightarrow \text{hh}\gamma) * \epsilon_{\text{MC}} * \epsilon_{\text{presel}} * \epsilon_{\text{PID}} * \epsilon_{\text{BDT_sig}}$$

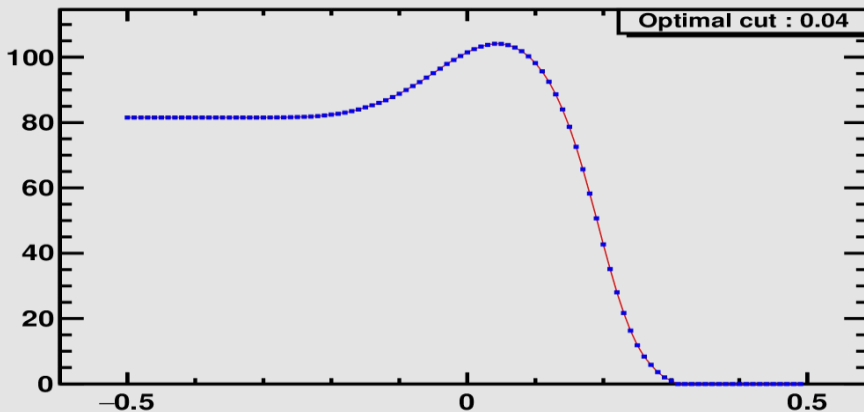
$$B = N^{3\sigma}_{\text{bkg}} * \epsilon_{\text{BDT_bkg}}$$

From linear fit of combinatorial



$\kappa\pi\gamma$ mass

FoM



BDT cut

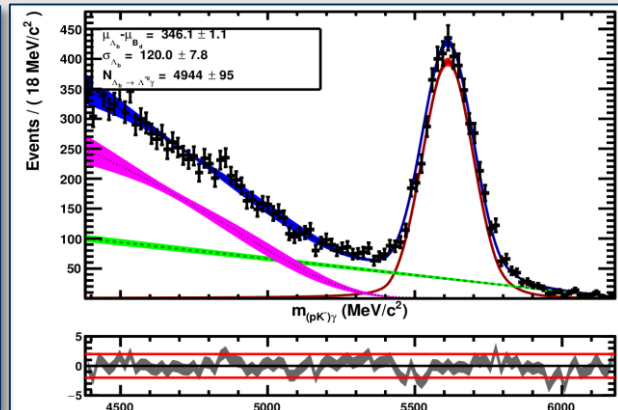
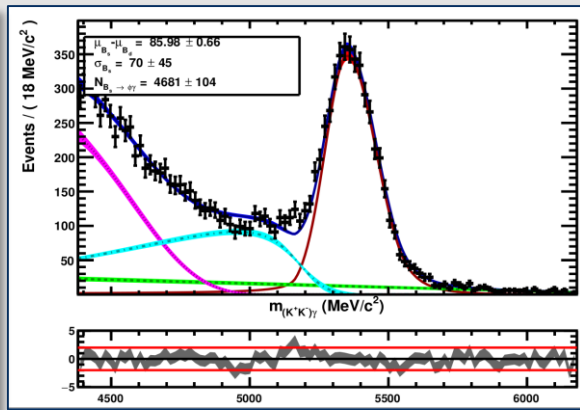
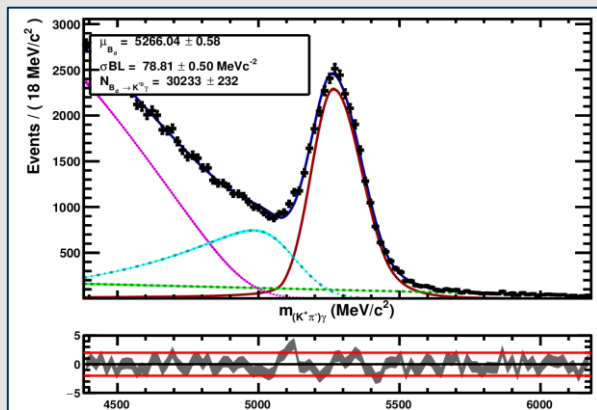
BDT EFFICIENCIES

Training Sample	BDTCut	Efficiency (K* / ϕ)	Efficiency (K1+K*2 / f^2)	BKG rejec.
K* γ _2011	> 0,04	94,2 %	??	88,8 %
K* γ _2012	> 0	91,3 %	79,6 %	93,3 %
(K*+K*2+K1) γ _2012	>0,01	89,6 %	88,3 %	93,8 %
(K*2+K1) γ _2012	>-0,01	88,5 %	88,7 %	93,8 %
$\phi\gamma$ _2011	> 0,13	93,4 %	??	93,7 %
$\phi\gamma$ _2012	> 0,07	92 %	43,6 %	96,5 %
($\phi + f^2$) γ _2012	> 0,07	89 %	76,3 %	96 %
$f^2\gamma$ _2012	> 0,05	85 %	81,3 %	95,8 %
(pK) γ _2011	> 0,1	83,3 %	x	94,5 %
(pK) γ _2012	> 0,07	79,2 %	x	97,6 %

SIMULTANEOUS FIT

2012 massfit (preliminary) :

External constraints : $(m_B - m_{B_s})$ & $(m_B - m_{\Lambda_b})$



Total yields expected Run I + Run II :

$K\pi\gamma$: ~380k

$KK\gamma$: ~40k

$pK\gamma$: ~35k

2D MASS FIT

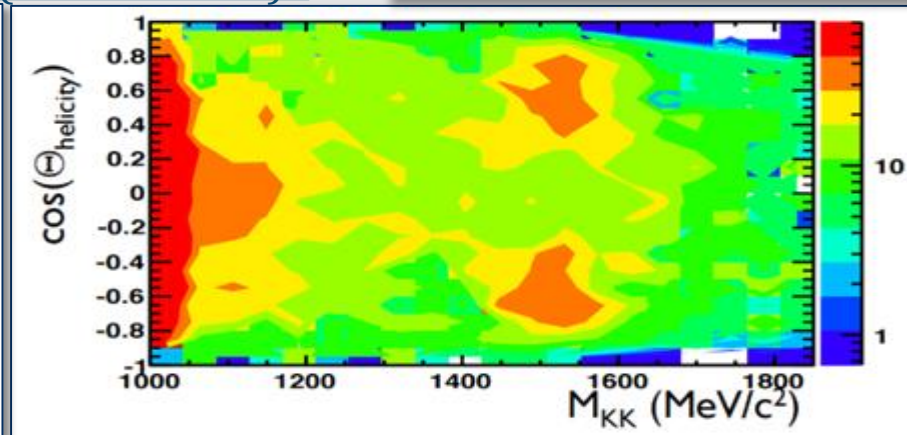
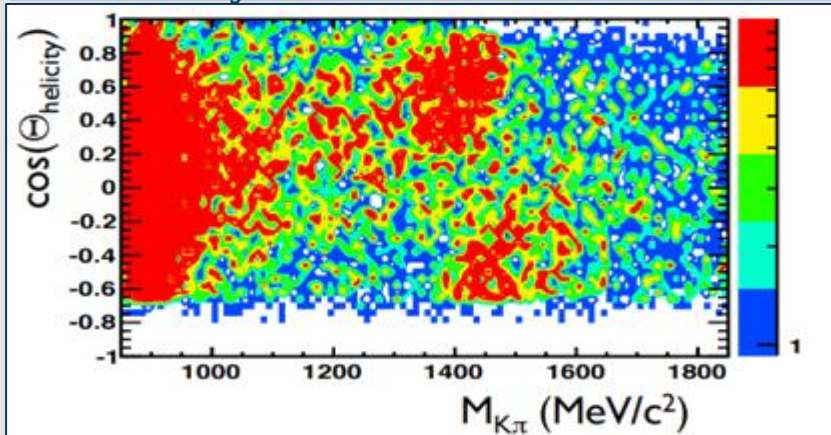
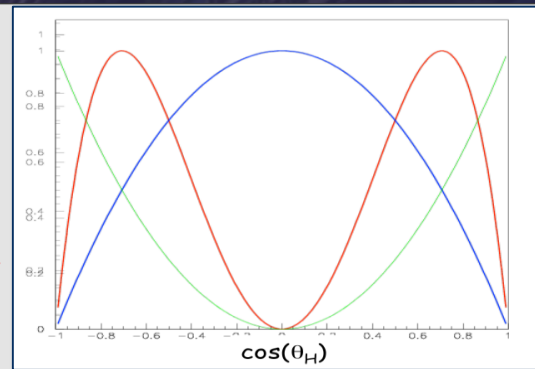
Make use of the helicity angle

$\text{Cos}^2(\Theta_H)$ for S-waves : π^0 backgrounds

$\text{Sin}^2(\Theta_H)$ for P-waves : $K^*(892)\gamma$, $K_1(1410)\gamma$, $\phi(1020)\gamma$

$\text{Cos}^2(\Theta_H).\text{sin}^2(\Theta_H)$ for D-waves : $K^*_2(1430)\gamma$, $f'_2(1525)\gamma$

Preliminary look to the data run1+run2 (2.7 + 1.0 fb⁻¹) :



- Clear D-wave contribution in the $K^*_2(K\pi)$ and $f'_2(KK)$ region
- Expecting a first observation of $B_s \rightarrow f'_2(1525)\gamma$ and $B^0 \rightarrow K_1(1410)\gamma$

CONCLUSION AND PROSPECTS

- We achieved the selection of a high amount of radiative decay events (Run I + Run II)
- We'll be eventually able to measure new modes' BR
- $B_s \rightarrow f'_2(1525) \gamma$ would be the second $b \rightarrow s \gamma$ in B_s decays

- Need to develop the 2D fit to extract BR

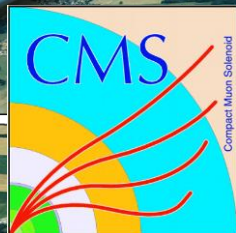
- Start writing the thesis ...



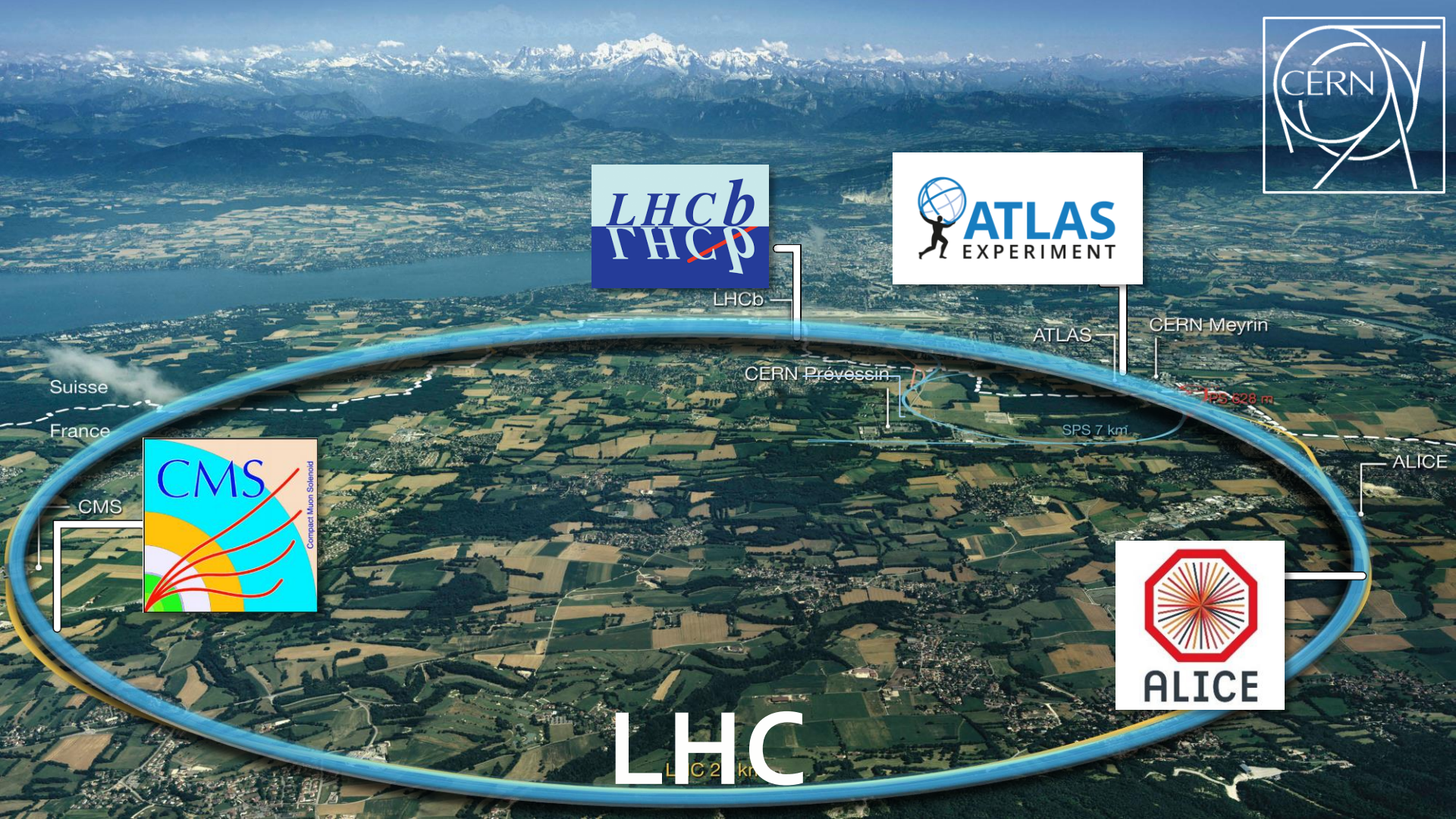


THANK YOU !

BACK UP AND STUFF...



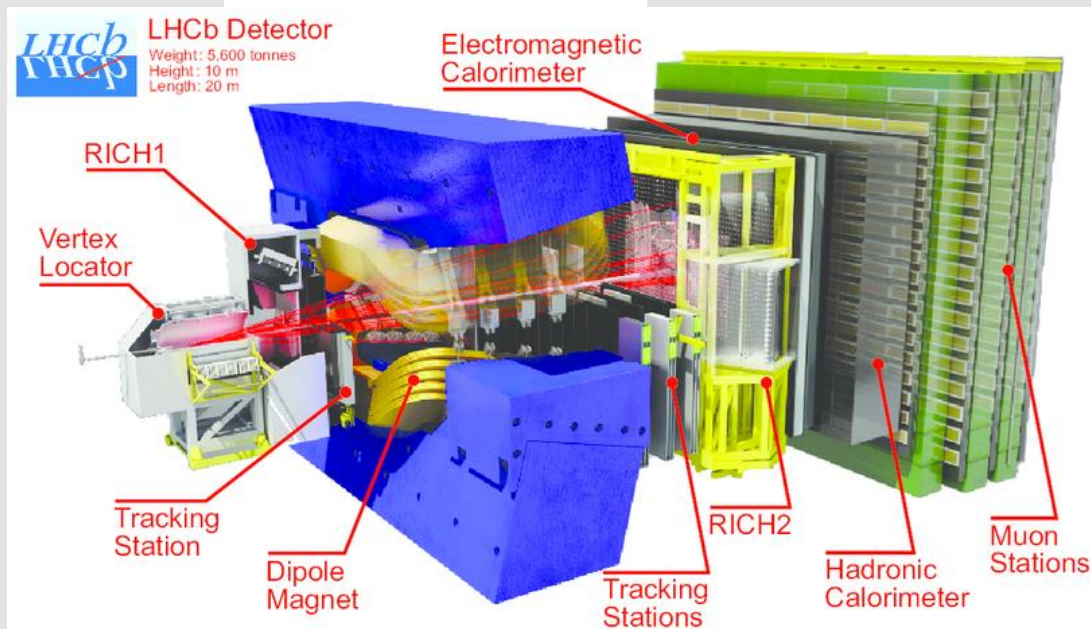
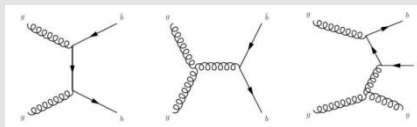
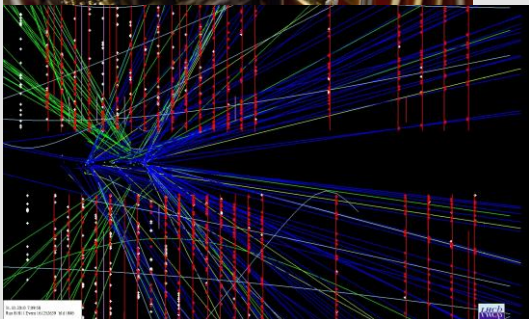
LHC



The Vertex Locator :

Bunch crossing rate : 40M/s

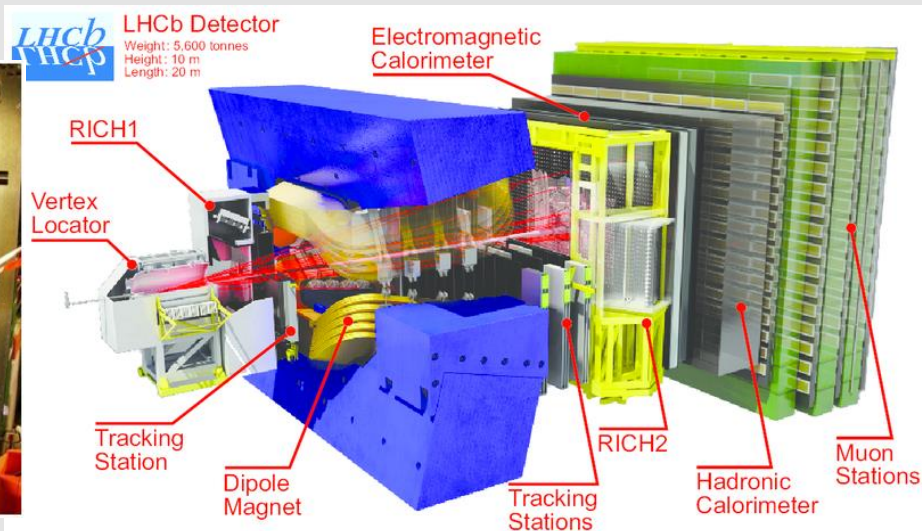
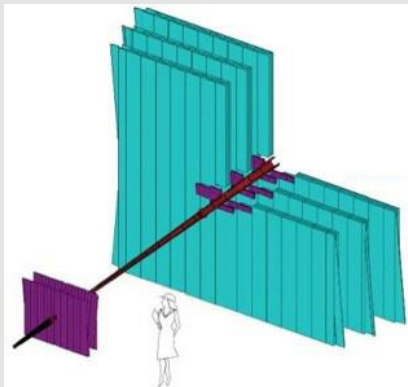
pp collision per crossing : $\sim 2,57$



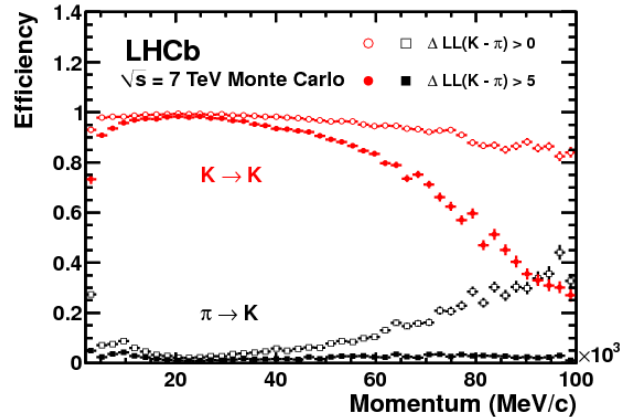
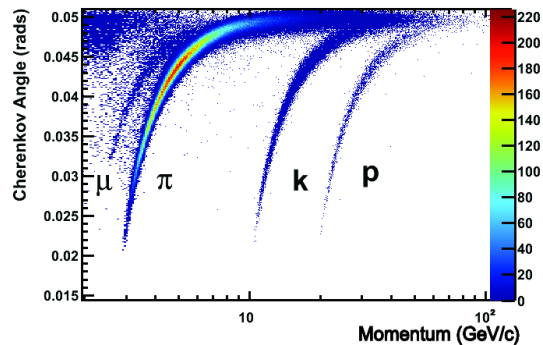
The Tracking system

Magnetic field : 3,6Tm

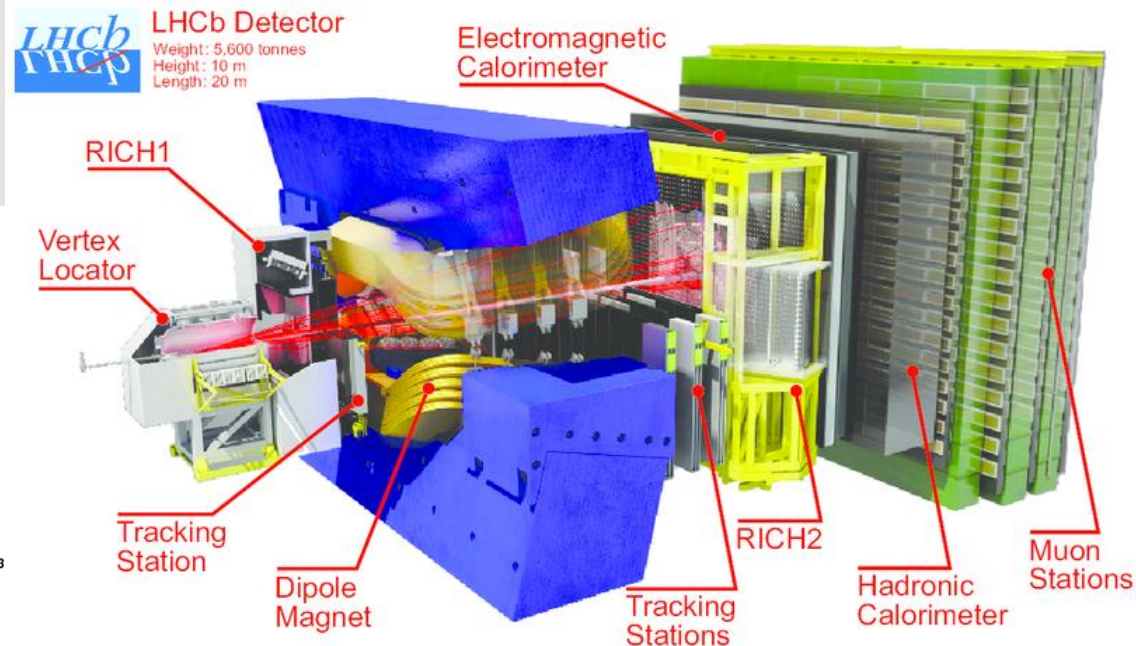
-> measure momenta : $R = \frac{mv}{qB}$



RICH detectors

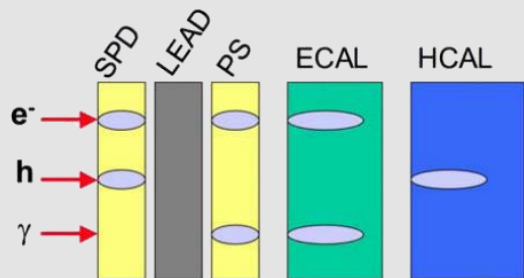
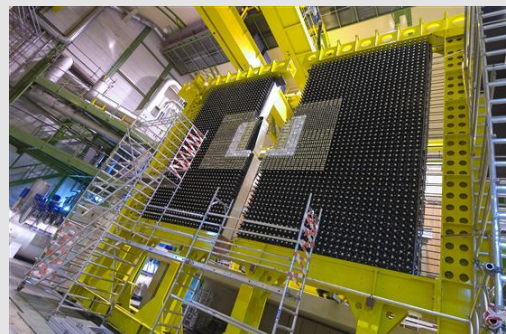


$$\cos \theta = \frac{1}{\beta n}$$

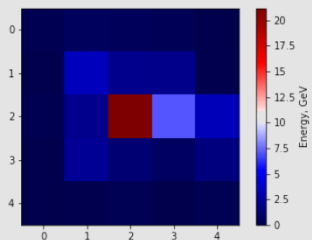


Calorimeters :

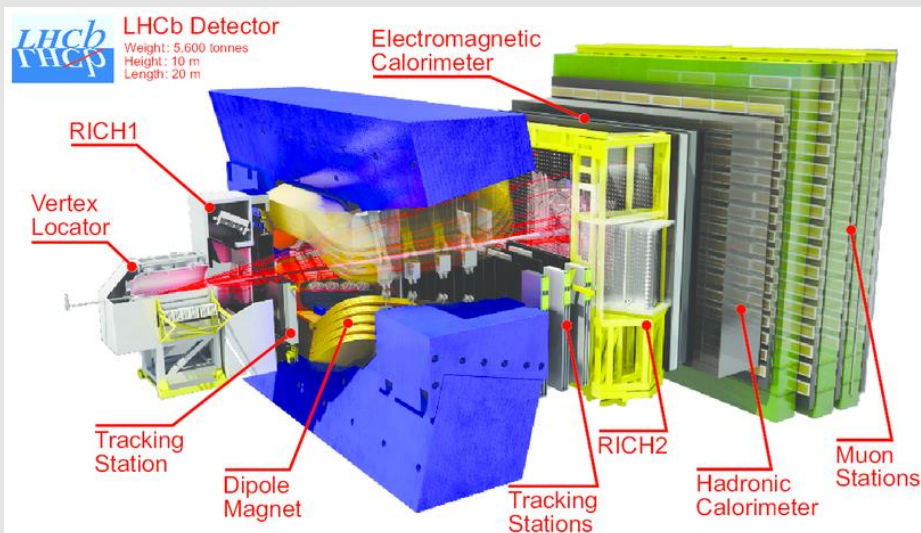
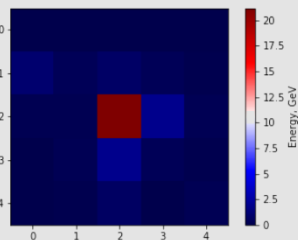
Resolution on B mass in radiatives : 80-100MeV



$$\pi^0 \rightarrow \gamma\gamma$$

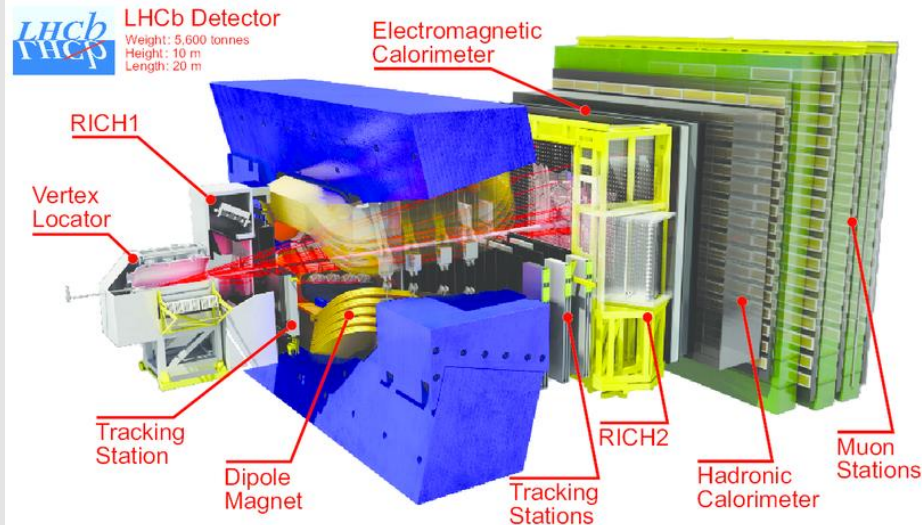
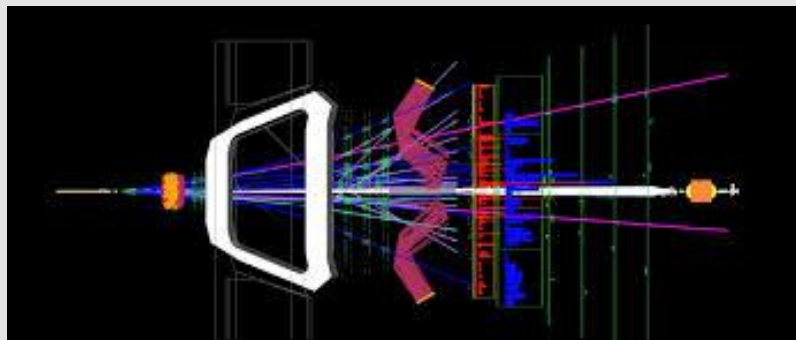
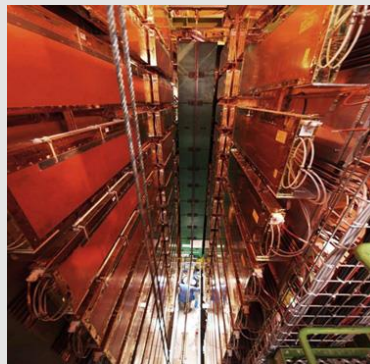


$$\gamma$$

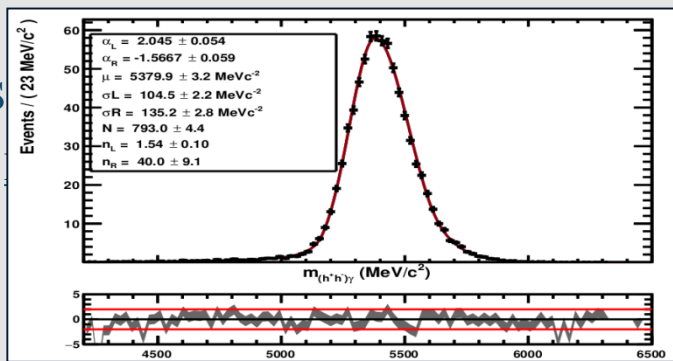


Muon trackers :

B- \rightarrow $\mu\mu$ / B- \rightarrow K* $\mu\mu$



CHARGED MIS-ID



normalization fixe

Note : normalisation wrt simult. fit :

$$\frac{N(K\pi\gamma \rightarrow KK\gamma)}{N(K\pi\gamma)} \neq \text{Contamination}$$

Contribution	Contamination	
KK -> KPI	2,4 %	2,6 %
PK -> PIK	9 %	14,2 %
PK -> KPI	0,89 %	1,2 %
KPI -> PIK	0,22 %	0,25 %
KPI -> KK	2,1 %	3,2 %
PK -> KK	5,8 %	7,4 %
KPI -> KP	13 %	9 %
KPI -> PK	1 %	0,8 %
KK -> PK	14,5 %	14,2 %

REWEIGHTING MC ?? (FOR LATER)

- ▶ MC
- ▶ Reweighting
- ▶ Efficiency studies ?
- ▶ Contamination computation ?

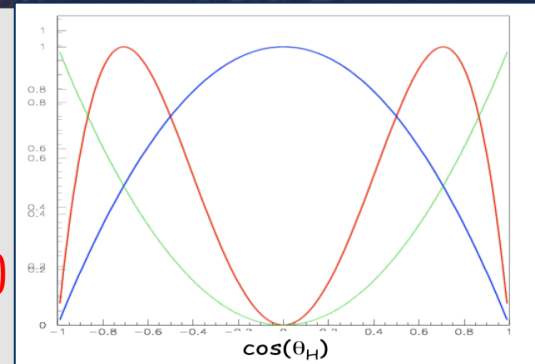
2D MASS FIT

Make use of the helicity angle

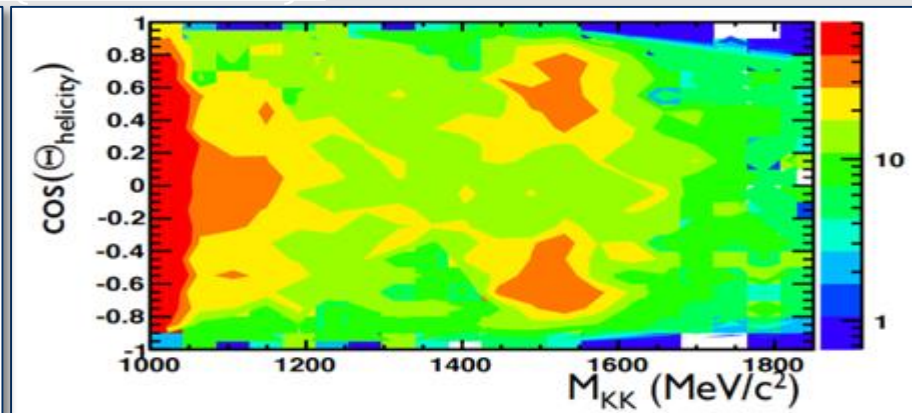
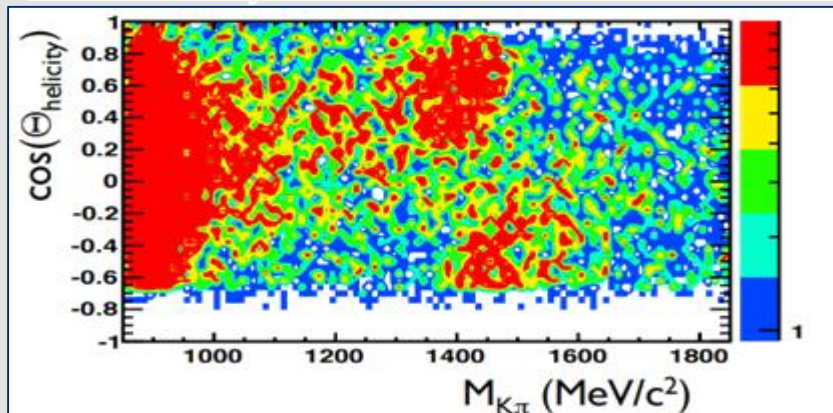
$\text{Cos}^2(\theta_H)$ for S-waves : π^0 backgrounds

$\text{Sin}^2(\theta_H)$ for P-waves : $K^*(892)\gamma$, $K_1(1410)\gamma$, $\phi(1020)\gamma$

$\text{Cos}^2(\theta_H).\text{sin}^2(\theta_H)$ for D-waves : $K^*_2(1430)\gamma$, $f'_2(1525)$



> release the anti S-waves cut
Preliminary look to the data run1+run2 (2.7 + 1.0 fb⁻¹) :



- Clear D-wave contribution in the $K^*_2(K\pi)$ and $f'_2(KK)$ region
- Expecting a first observation of $B_s \rightarrow f'_2(1525)\gamma$ and $B^0 \rightarrow K_1(1410)\gamma$

PRESELECTION

- ▶ Standard $V\gamma$ selection with :
 - ▶ - a release of the h^+h^- mass window (below the charm threshold)
 - ▶ - no cut on the helicity
 - ▶ - $m(h\gamma_{\rightarrow\pi^0}) > 2000 \text{ MeV}/c^2$

- ▶ List of cuts :

$h_MINIPCHI2 > 16$

$\max(h1_PT, h2_PT) > 1200$

$\min(h1_PT, h2_PT) > 500$

$h_P > 500$

$h_P < 100000$

$h1_eta > 1.5$

$h1_eta1 < 5.0$

$res_ENDVERTEX_CHI2 < 9$

$h_TRACK_CHI2NDOF < 3$

$h_TRACK_GhostProb < 0.4$

$B_DiraAngle < 0.06$

$B_PT > 2000$

$B_IPCHI2_OWNPV < 9.0$

$gamma_CL > 0.2$

$gamma_PP_IsPhoton > 0.6$

$gamma_PT > 3000$

$B_M01 < 1850$ (< 3000 for $pK\gamma$)

PID SELECTION

- ▶ In order to reduce the cross feeds and build exclusive samples the following cut is used :
 - ▶ $\text{probNN}h_i(h) > \mathbf{x}h_i$ & $\text{probNN}h_i(h) > \text{probNN}h_j(h)$ for all $h_j \neq h_i$ ($h_i = \pi, K, p$)
 - ▶ Optimisation : Scan (xh_1, xh_2) cut values
 - ▶ The corresponding efficiency of each (h_1, h_2) to be reconstructed as (kpi) , (kk) , (pk) etc... is extracted from calibration samples using nTracks-reweighted MC samples and the PIDCalib tool.

 - ▶ MC used (Sim8 2011 / 2012):
 - ▶ $B_d \rightarrow K^*(892) \gamma$
 - ▶ $B_s \rightarrow \phi(1020) \gamma$
 - ▶ $\Lambda_b \rightarrow \Lambda^*(1520/1670/1820/1830) \gamma$
- 1 sample / possible reconstruction
(including misID $kpi \rightarrow kk$, etc...)
is build using truthmatching
- ▶ - preselection and truthmatching applied on MC
 - ▶ PIDCalib settings :
 - ▶ - Custom 3D binning in $\{\text{nTracks}(5), \text{PT}(8), \text{ETA}(8)\}$ based on MC distributions
 - ▶ - V2 ProbNN variables

PID FOM

▶ The FoM computed for each (x_{h1}, x_{h2}) cut values is : $\text{FoM} = \frac{S}{\sqrt{S+B}}$

▶ With :

$$S = L * 2 * \sigma(\text{bb}) * f_{b \rightarrow (B0/Bs/\Lambda B)} * \text{BR}(B \rightarrow \text{hh}\gamma) * \epsilon_{\text{MC}} * \epsilon_{\text{presel}} * \epsilon_{\text{PID}}(x_{h1}, x_{h2})$$

▶ assuming $\text{BR}_{\text{vis}}(B \rightarrow \text{K}\pi\gamma) = \text{BR}(B \rightarrow \text{K}^*(892)\gamma) / 0,65$

▶ $\text{BR}_{\text{vis}}(B \rightarrow \text{KK}\gamma) = \text{BR}(B \rightarrow \text{phi}\gamma) / 0,30$

From preliminary fit

▶ $\text{BR}_{\text{vis}}(B \rightarrow \text{pK}\gamma) = 3.39 \cdot 10^{-5}$ (from Vicente's thesis)

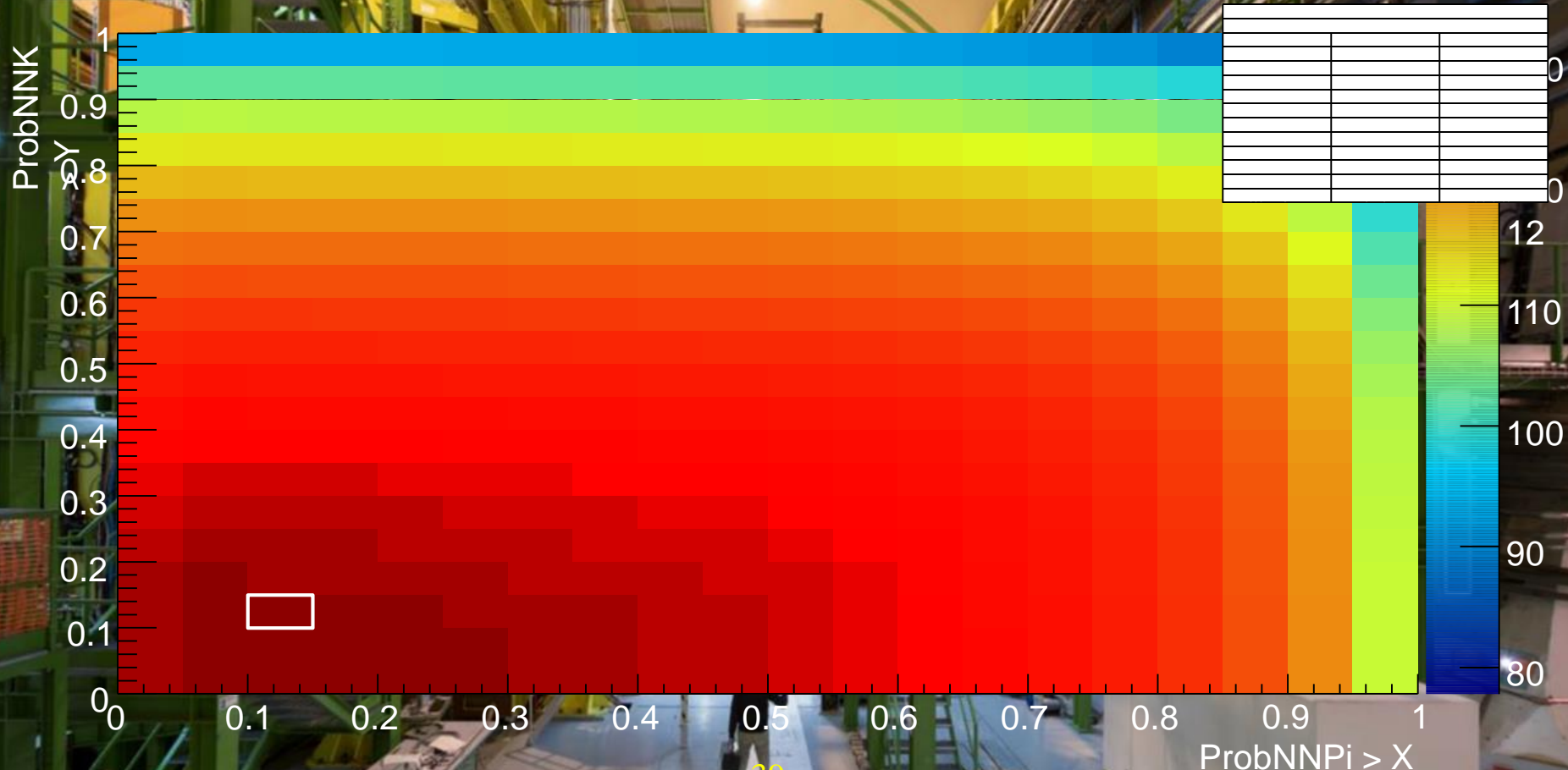
▶ ϵ_{PID} is taken as the mean of the PIDCalib eff in the reference sample

$$B = \sum_{\text{cross feeds}} L * 2 * \sigma(\text{bb}) * f_{b \rightarrow (B0/Bs/\Lambda B)} * \text{BR}(B \rightarrow [\text{hh}]_{\text{trueID}}\gamma) * \epsilon_{\text{MC}} * \epsilon_{\text{presel}} * \epsilon_{\text{misID}}(x_{h1}, x_{h2})$$

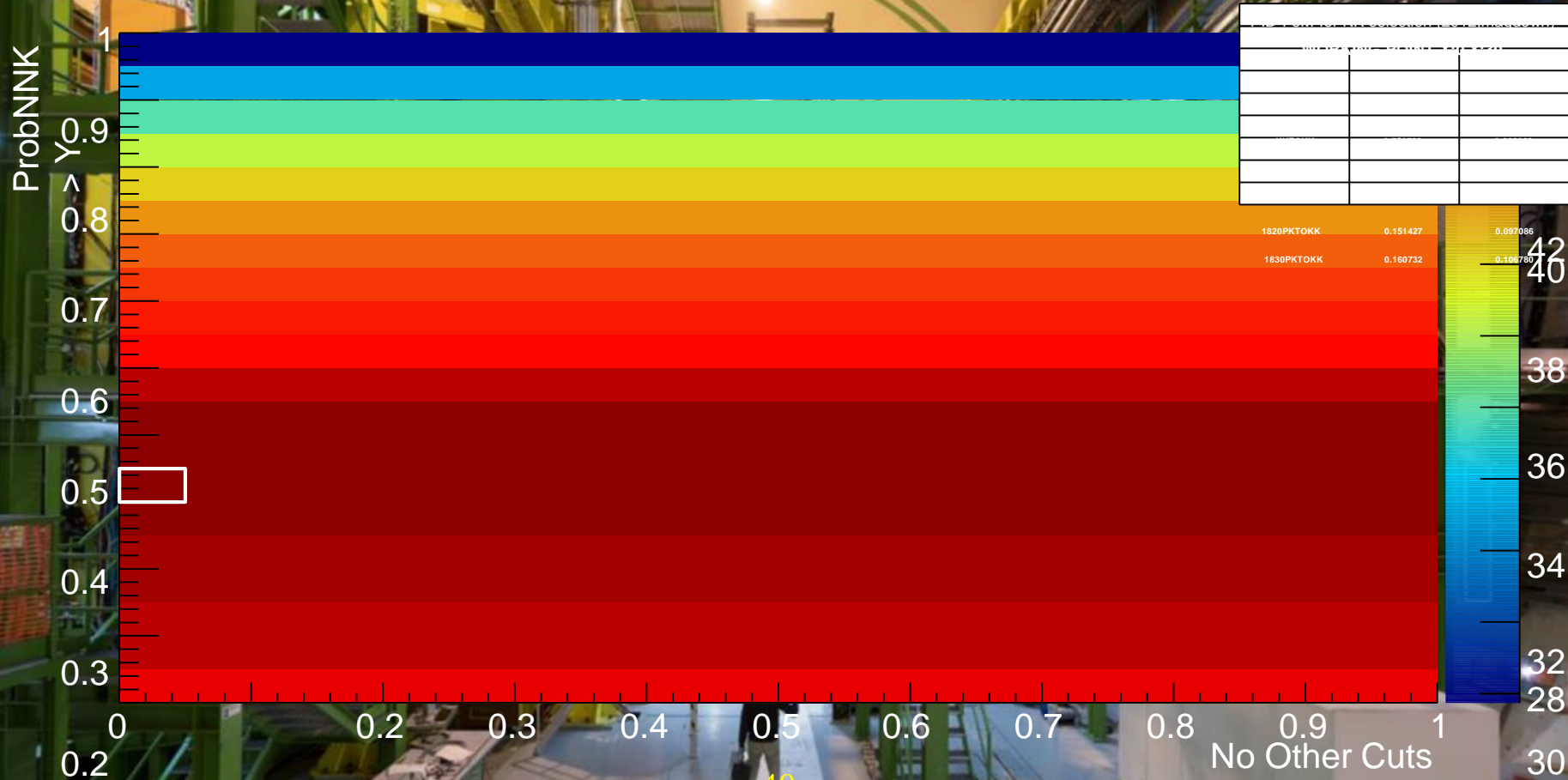
▶ This FoM has been drawn for a scan performed by steps of 0,05 in ProbNN

▶ and for each final state/year/magnet polarity

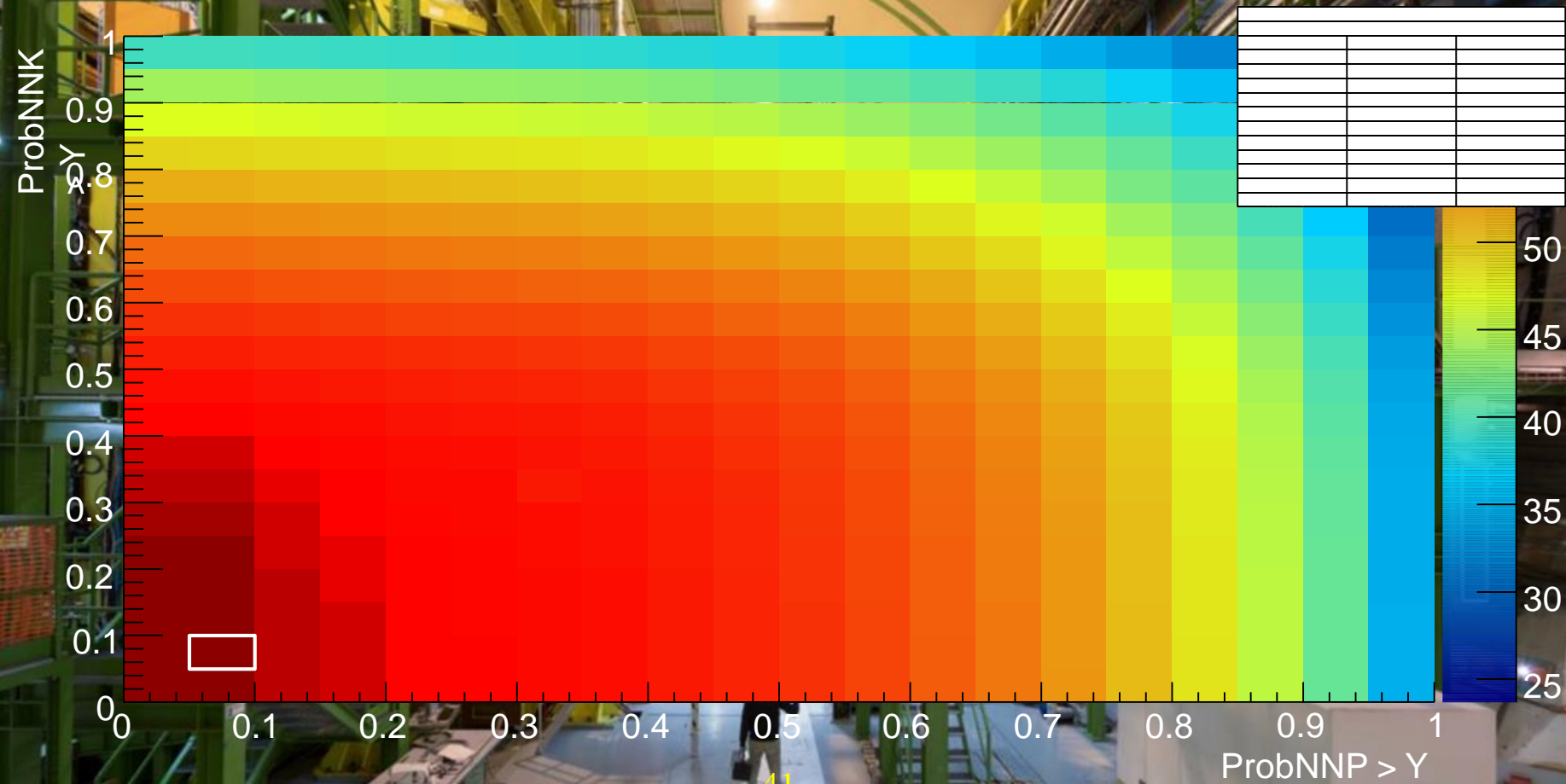
PID FOM FOR KPI SELECTION (2012,MAGDOWN)



PID FOM FOR KK SELECTION (2012,MAGDOWN)



PID FOM FOR PK SELECTION (2012, MAGDOWN)



ProbNNP > Y

YCM01
UX85

FINAL WORKING POINTS

► The Fol

Final State	Cut h1	Cut h2	Eff for the signal
$K\pi\gamma$	ProbNNK > 0,1	ProbNNpi > 0,1	83 %
$KK\gamma$	ProbNNK > 0,3	ProbNNK > 0,3	73%
$pK\gamma$	ProbNNp > 0,05	ProbNNK > 0,05	60%

► Errors to be computed...

MULTIVARIATE ANALYSIS

▶ The next step is to train a Multivariate selection to reduce combinatorial contamination

▶ Signal samples:

▶ MC used for $K\pi\gamma$:

$$B^0 \rightarrow (K^*(892) \rightarrow K\pi)\gamma \quad (2011/12)$$

$$B^0 \rightarrow (K^*_2(1430) \rightarrow K\pi)\gamma \quad (2012)$$

$$B^0 \rightarrow (K_1(1410) \rightarrow K\pi)\gamma \quad (2012)$$

▶ MC used for $KK\gamma$:

$$B_s \rightarrow (\phi \rightarrow K K)\gamma \quad (2011 / 2012)$$

$$B_s \rightarrow (f'_2(1525) \rightarrow K K)\gamma \quad (2011)$$

▶ MC used for $KK\gamma$:

$$\Lambda_b \rightarrow (\Lambda^*(1520) \rightarrow p K)\gamma$$

$$\Lambda_b \rightarrow (\Lambda^*(1670) \rightarrow p K)\gamma$$

$$\Lambda_b \rightarrow (\Lambda^*(1820) \rightarrow p K)\gamma$$

$$\Lambda_b \rightarrow (\Lambda^*(1830) \rightarrow p K)\gamma$$

We train 3 MVA :
One with only K^* for 2011 and 2012
One with a mix of all the resonances for 2012
One with K^*_2 / K_1 for 2012

We train 3 MVA :
One with only $\phi(1020)$ for 2011 and 2012
One with a mix of the two resonances
One with only f'_2

We train one MVA for each year

▶ All the MC samples are reweighted on nTracks & PIDCalib_eff

MULTIVARIATE ANALYSIS

	Sample	MC Signal	SB	SB (no PID cut)
▶ BK	$K\pi\gamma_{2011}$	107544	6540	15999
▶ S	$K\pi\gamma_{2012}$	29340	30970	75291
▶	$KK\gamma_{2011}$	84080	460	3421
▶	$KK\gamma_{2012}$	30208	1815	163309
	$pK\gamma_{2011}$	40249	384	7707
	$pK\gamma_{2012}$	45344	1988	40429

Low stats for  training...

▶ Since the number of events in the Side Band is quite low for $KK\gamma$ and $pK\gamma$ samples after the PID cut, we choose to train on a SB without this cut.

▶ We first tested this for $K\pi\gamma$, by comparing two MVA trained on each of the SB

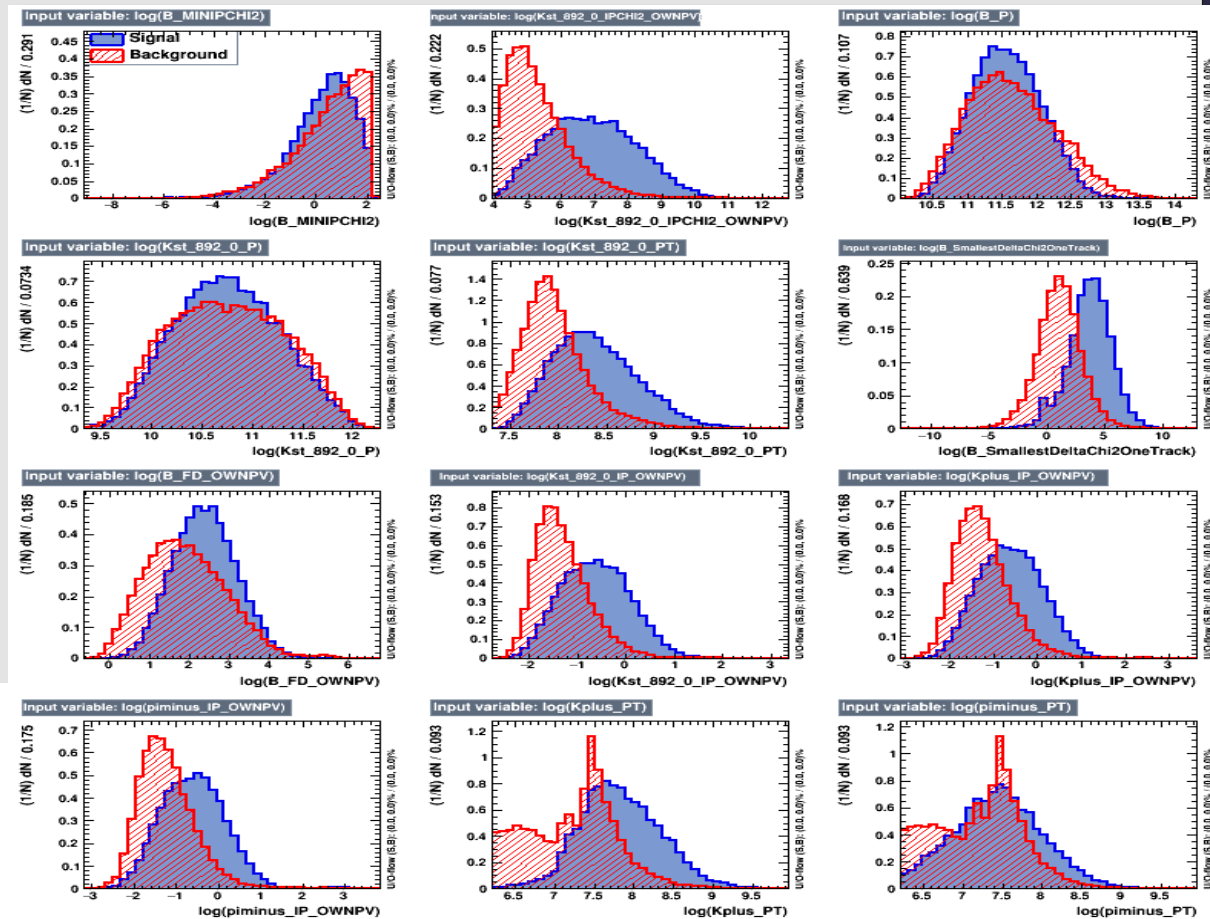
MVA BUILDING

- ▶ Different algorithms and settings have been tested (on $K\pi\gamma$):
 - ▶ Best algorithm : BDT (AdaBoost)
 - ▶ Best Settings : - Ntrees=800
 - ▶ - MaxDepth=3
 - ▶ - minNodSize=2,5%
 - ▶ - nCuts=200
 - ▶ - AdaBoost $\beta=1$
 - ▶ Samples have been splitted in 2 according to (evtNumber%2=0), each subsample being tested on the other one.
 - ▶ Starting from a set of 40 variables, the minimum set has been chosen removing variables one by one according to their rank during the training.
 - ▶ We end up with 13 variables. Removing any one more seems to worsen significantly the performances (ROC AUC).

BDT VARIABLES

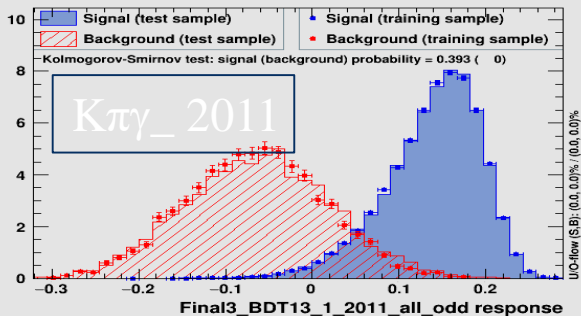
Final set :

- $B \min \chi_{IP}^2$
- B pseudorapidity
- B momentum
- B Flight Distance
- B vertex isolation (Smallest $\Delta\chi^2$)
- χ_{IP}^2 of the (hh) resonance
- IP of the resonance
- Momentum of the resonance
- IP of the tracks
- Transverse momenta of the tracks

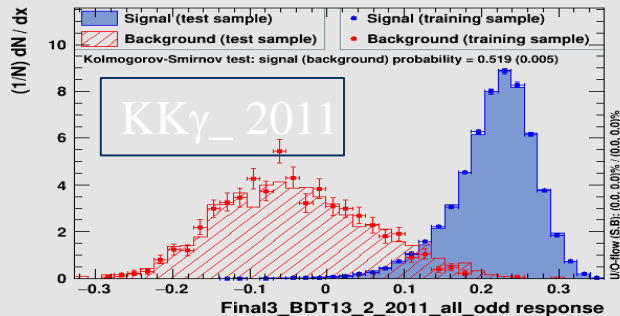


OVERTRAINING TESTS

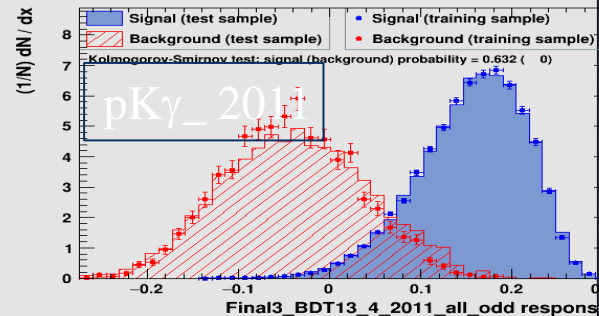
TMVA overtraining check for classifier: Final3_BDT13_1_2011_all_odd



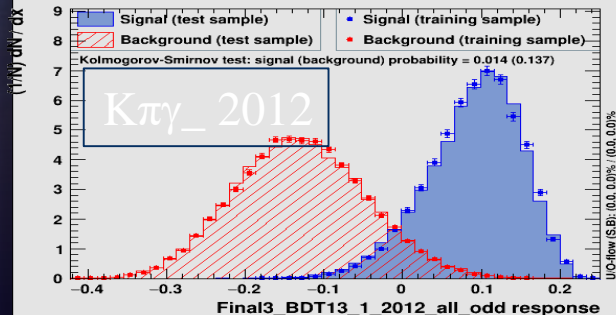
TMVA overtraining check for classifier: Final3_BDT13_2_2011_all_odd



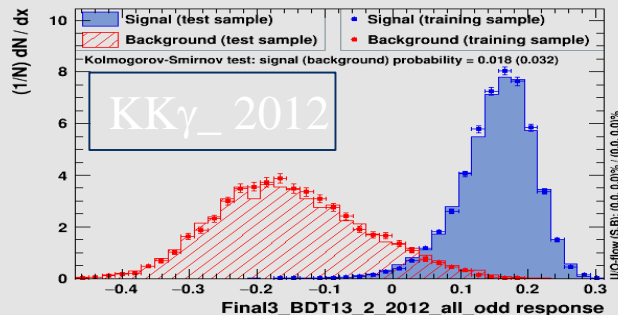
TMVA overtraining check for classifier: Final3_BDT13_4_2011_all_odd



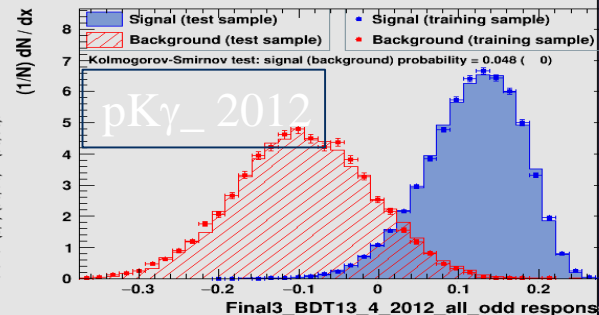
TMVA overtraining check for classifier: Final3_BDT13_1_2012_all_odd



TMVA overtraining check for classifier: Final3_BDT13_2_2012_all_odd

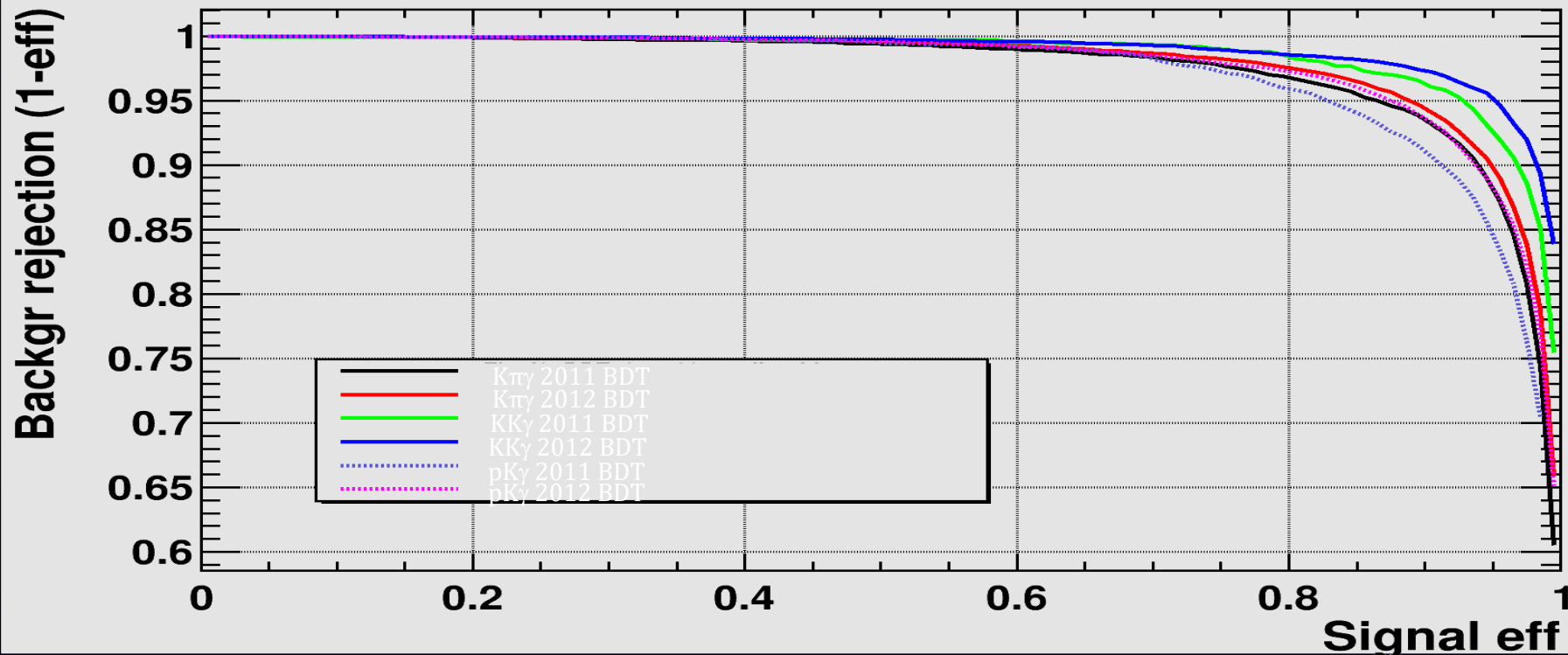


TMVA overtraining check for classifier: Final3_BDT13_4_2012_all_odd



BDT PERFORMANCES

The variable set and settings optimised on $K\pi\gamma$ are used to train BDT on the two other modes :



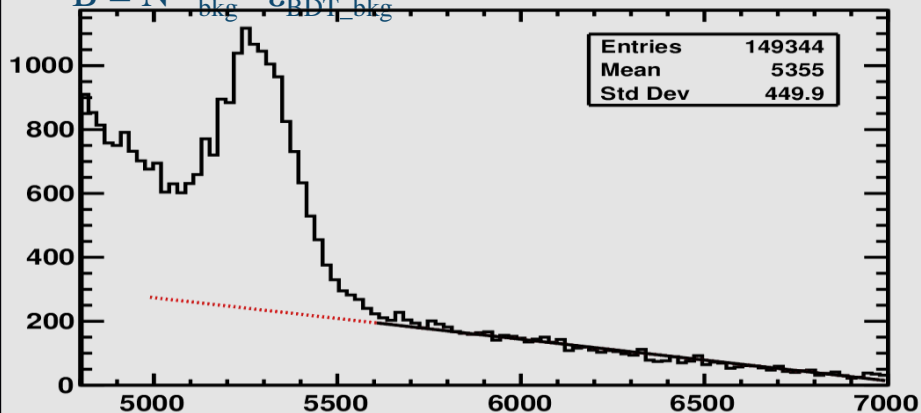
OPTIMAL CUT

$$\text{FoM} = \frac{S}{\sqrt{S+B}}$$

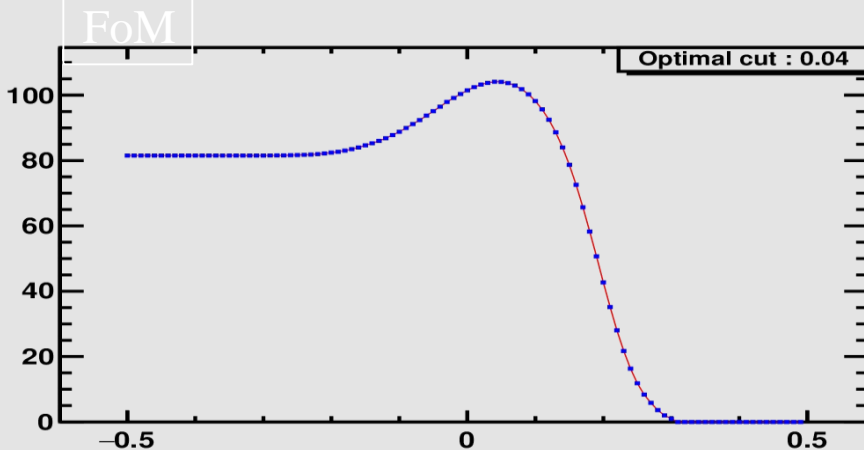
From linear fit of combinatorial

$$\text{With } S = L * 2 * \sigma(bb) * 1_{b \rightarrow (B^0/B_s^0/Ab)} * \text{BR}(B \rightarrow hh\gamma) * \epsilon_{MC} * \epsilon_{\text{presel}} * \epsilon_{\text{PID}} * \epsilon_{\text{BDT_sig}}$$

$$B = N_{\text{bkg}}^{3\sigma} * \epsilon_{\text{BDT_bkg}}$$



$\kappa\pi$ mass



BDT cut

BDT EFFICIENCIES

Training Sample	BDTCut	Efficiency (K^*/ϕ)	Efficiency (K_1+K_2/ϕ^2)	BKG rejec. (SB in test samples)	BKG rejec. (on SB after PID cut)
<u>Kγ</u>					
$K^*\gamma_{2011}$	> 0,04	94,2 %	??	88,8 %	89,2 %
$K^*\gamma_{2012}$	> 0	91,3 %	79,6 %	93,3 %	94 %
$(K^*+K_2+K_1)\gamma_{2012}$	>0,01	89,6 %	88,3 %	93,8 %	95,6%
$(K_2+K_1)\gamma_{2012}$	>-0,01	88,5 %	88,7 %	93,8 %	95,6 %
<u>KKγ</u>					
$\phi\gamma_{2011}$	> 0,13	93,4 %	??	93,7 %	93,1 %
$\phi\gamma_{2012}$	> 0,07	92 %	43,6 %	96,5 %	95,6 %
$(\phi + \phi^2)\gamma_{2012}$	> 0,07	89 %	76,3 %	96 %	95,6 %
$\phi^2\gamma_{2012}$	> 0,05	85 %	81,3 %	95,8 %	95,6 %
<u>PKγ</u>					
$(pK)\gamma_{2011}$	> 0,1	83,3 %	x	94,5 %	92,3 %
$(pK)\gamma_{2012}$	> 0,07	79,2 %	x	97,6 %	94,8 %

We see that :

- BDT trained with only K^*/ϕ MC samples gives a bad efficiency for the modes we are willing to see (43 % for $\phi^2\gamma$, 79% for $K_1/K_2\gamma$)

- Training on a SB with no cuts on PID gives a good background rejection on SB after PID cut

FITTING THE $HH\gamma$ MASS

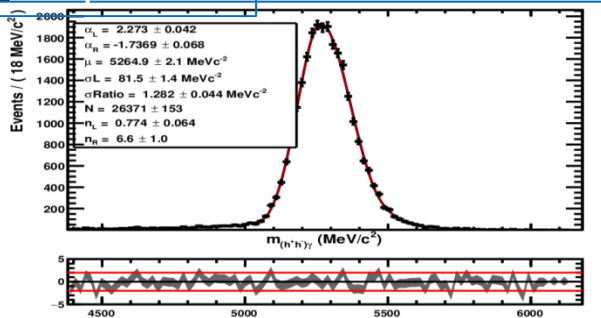
Signal :

- ▶ Asymmetric bifurcated CB ($\mu, \sigma_{\text{Left}}, N$; $\frac{\sigma_{\text{Left}}}{\sigma_{\text{Right}}}, \alpha_{\text{Left}}, n_{\text{Left}}, \alpha_{\text{Right}}, n_{\text{Right}}$)

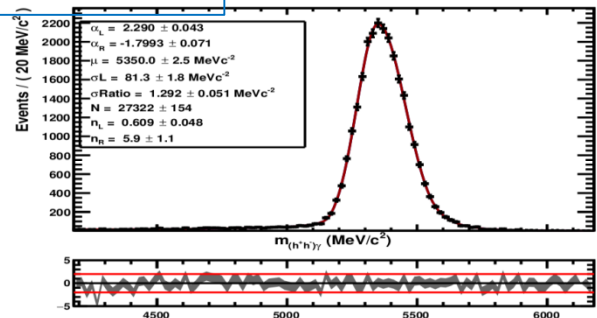
Free parameters

Fixed on MC

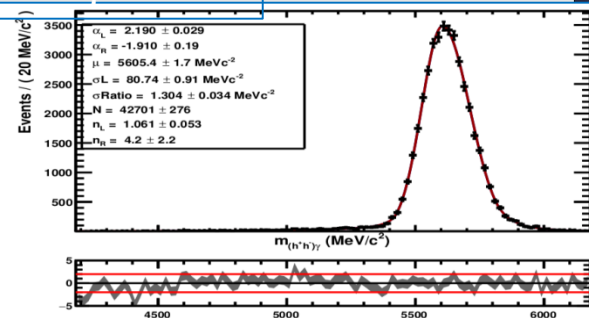
$K^* \gamma$ MC Fit



$\phi\gamma$ MC Fit



$\Lambda^* \gamma$ MC Fit



- Combinatorial BKG: Exponential (shape and N free)
- Inclusive partially reconstructed : $B \rightarrow hh \pi^0/\gamma + X$ ($X > 2$)
- ▶ Argus (free, $\mu = m_B - 2 m_\pi$) convoluted with Asym. Gaussian ($\mu=0, \sigma = \sigma_{\text{sig}}$) (CB convo ?!)

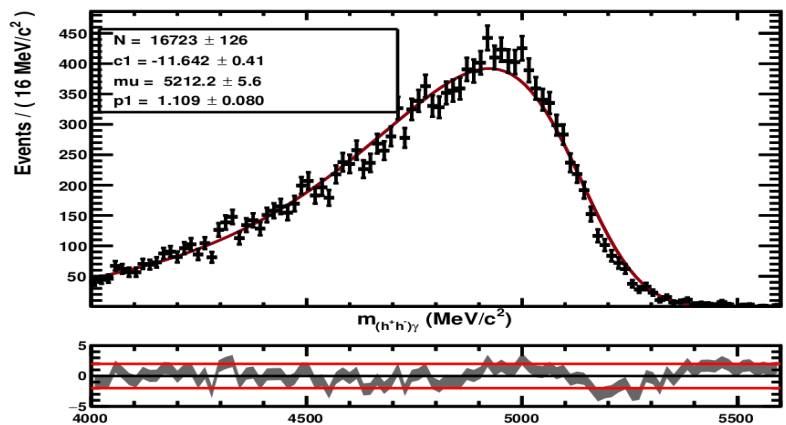
PARTIALLY RECONSTRUCTED

Partially reconstructed : $B \rightarrow (hh\pi)_{\text{res}} \gamma$

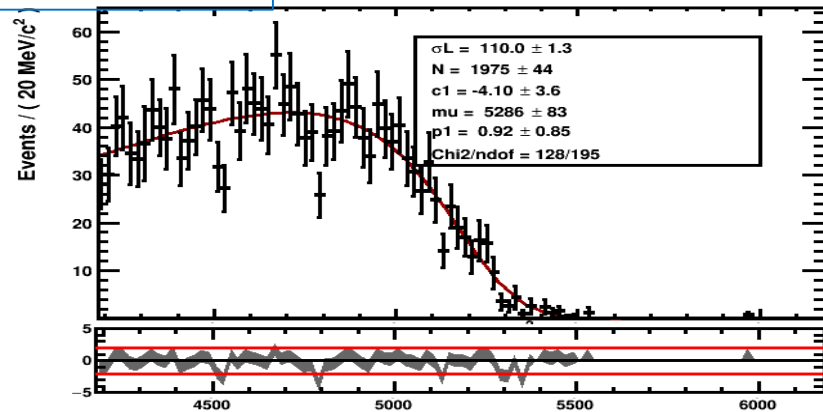
Shape : Argus ($\mu = m_B - m_\pi$, fixed on MC, free yield) convoluted with an

Asymmetric Gaussian ($\mu=0$, $\sigma = \sigma_{\text{sig}}$)

$B_u \rightarrow (k\pi\pi)\gamma$



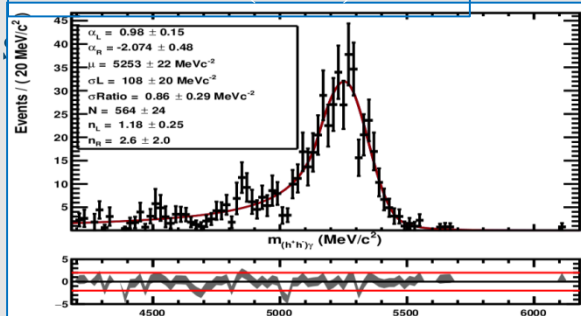
$B_s \rightarrow \phi\pi^0\gamma$



- For every shape we perform an extended maximum likelihood unbinned fit.
 - The MC used is reweighted to propagate the overall PID efficiencies, and the BDT selection is applied.
- TODO: study $m(\pi\pi) \rightarrow \gamma\gamma$, would be Argus, shape and yields fixed

Π^0/Γ MIS-ID CONTRIBUTIONS

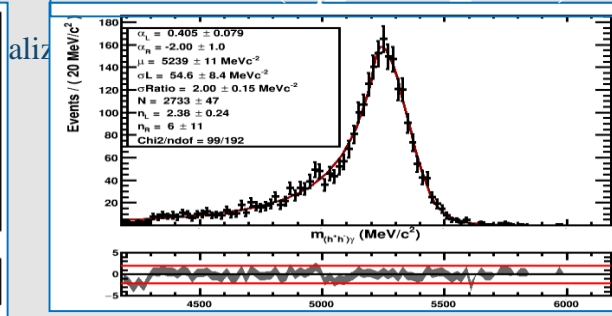
$B_d \rightarrow K\pi\pi^0$ (2012)



BR = 3,78 e-5

Contamination : C = 2,6%

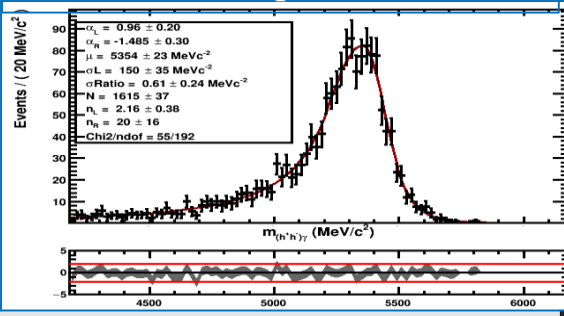
$B_d \rightarrow KK\pi^0$ (sq_Dalitz_2012)



BR = 0,22 e-5

Contamination : C = 2,5%

$B_s \rightarrow KK\pi^0$ (sq_Dalitz_2012)



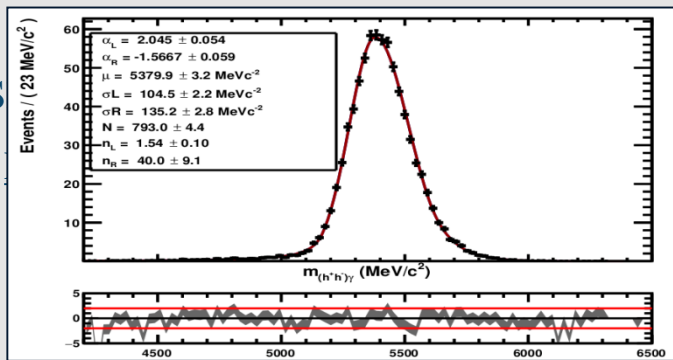
$$BR = \frac{BR(B_d \rightarrow K\pi\pi^0)}{BR(B_d \rightarrow K^*\gamma)} * BR(B_s \rightarrow \phi\gamma)$$

Contamination : C = 1,2%

Note : the contamination are lower in 2011, seems to be due to the BDT

TO DO : request better MC with new DecFile

CHARGED MIS-ID



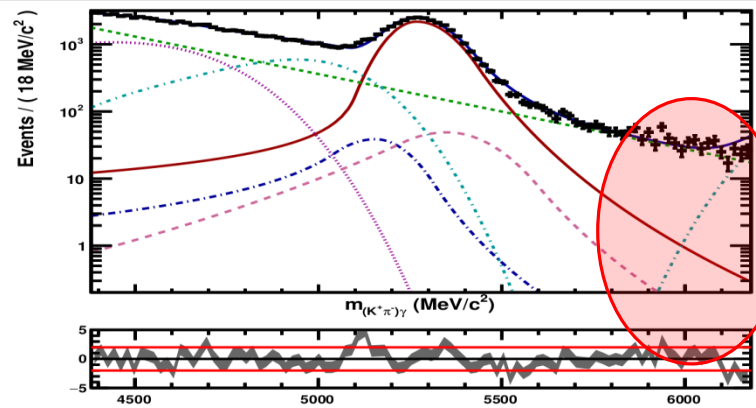
normalization fixe

Note : normalisation wrt simult. fit :

$$\frac{N(K\pi\gamma \rightarrow KK\gamma)}{N(K\pi\gamma)} \neq \text{Contamination}$$

Contribution	Contamination	
KK -> KPI	2,4 %	2,6 %
PK -> PIK	9 %	14,2 %
PK -> KPI	0,89 %	1,2 %
KPI -> PIK	0,22 %	0,25 %
KPI -> KK	2,1 %	3,2 %
PK -> KK	5,8 %	7,4 %
KPI -> KP	13 %	9 %
KPI -> PK	1 %	0,8 %
KK -> PK	14,5 %	14,2 %

SIMULTANEOUS FIT



s):

WORK IN PROGRESS !

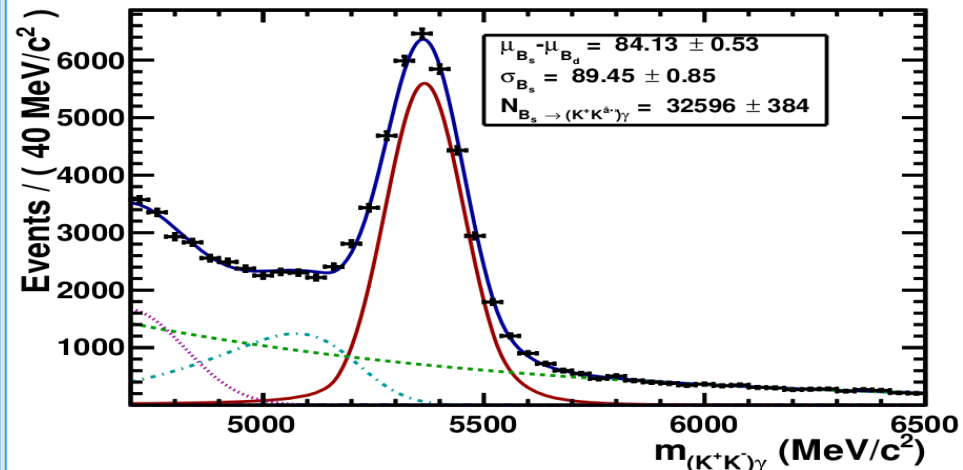
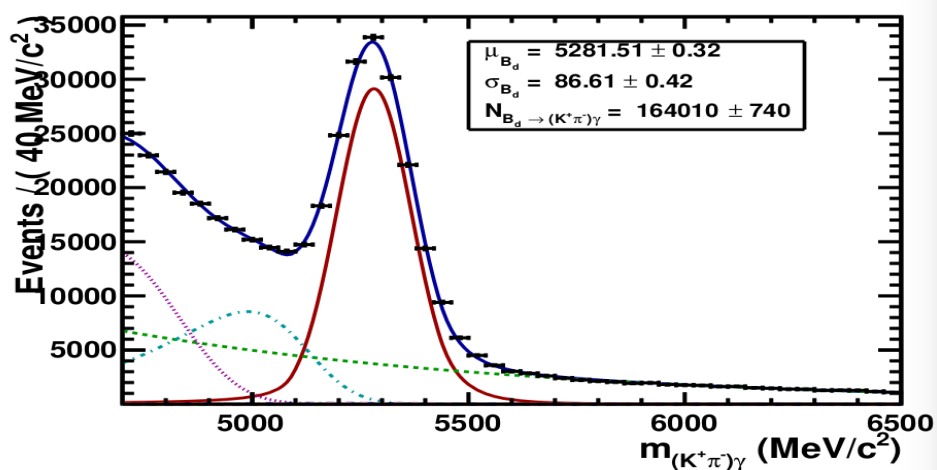
RUN 2 ANALYSIS

All the requested 2015/6 MC samples for kpi and kk are (almost) here.

PID Calib efficiency tables have been generated.

BDT for Run 2 with new isolation variables will be soon developed

▶ Run 2 approximated yields :



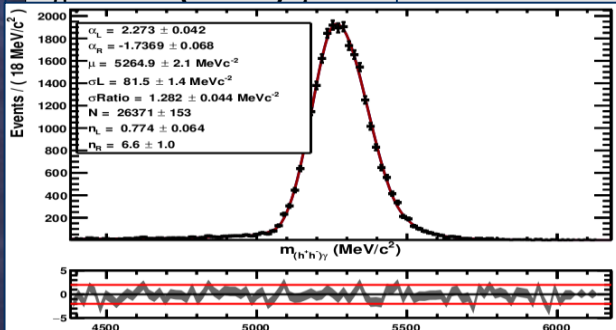
Work is starting to apply a similar analysis procedure for $K_s \text{ hhy run1+run2..}$

BACK UP

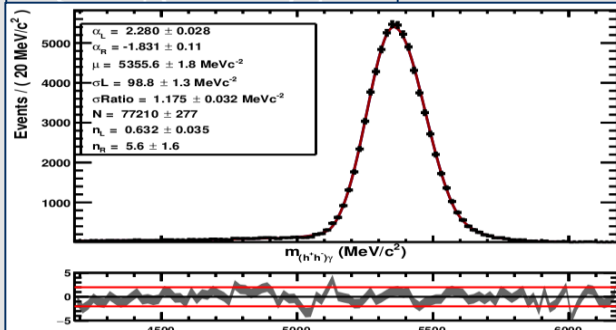
SIGNAL AND BKG SHAPES

- For every shape we perform an extended maximum likelihood unbinned fit.
- The MC used is reweighted to propagate the overall PID efficiencies, and the BDT

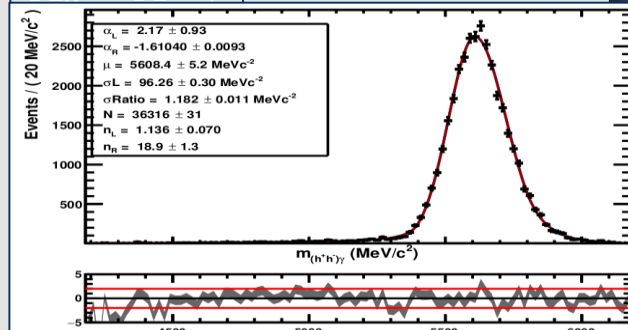
$B_d \rightarrow \psi^*(892) \gamma$



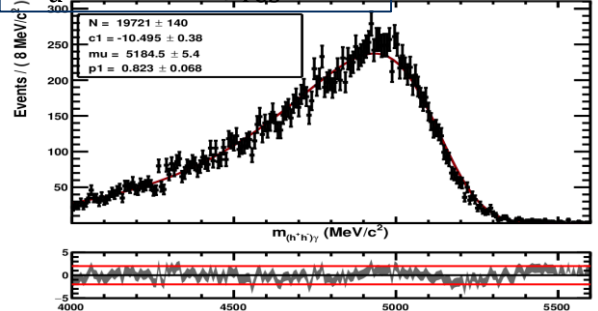
$B_s \rightarrow \phi(1020) \gamma$



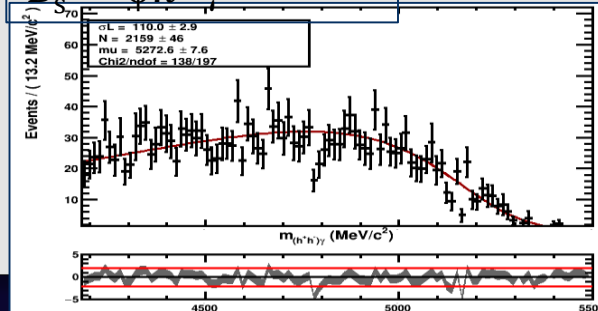
$\Lambda_b \rightarrow \Lambda^* \gamma$



$B_u \rightarrow (\text{K}\pi\pi)_{\text{res}} \gamma$



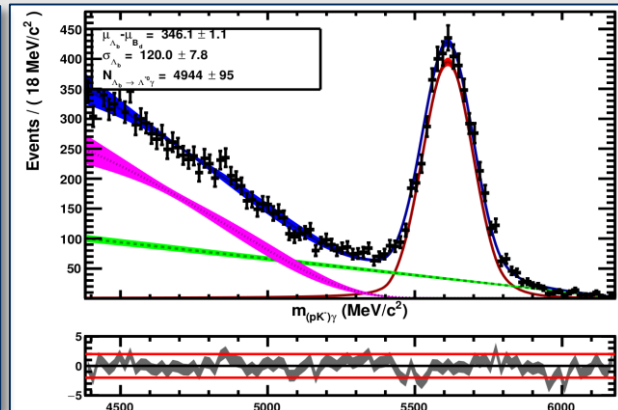
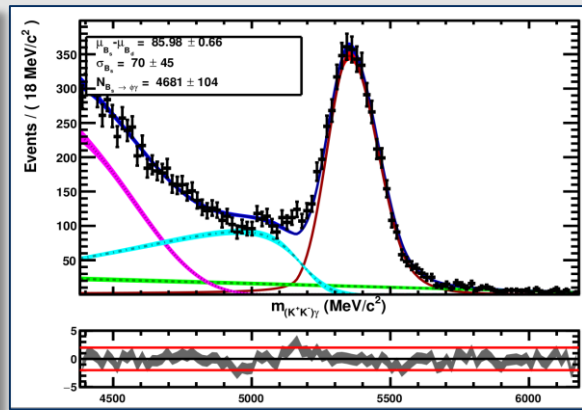
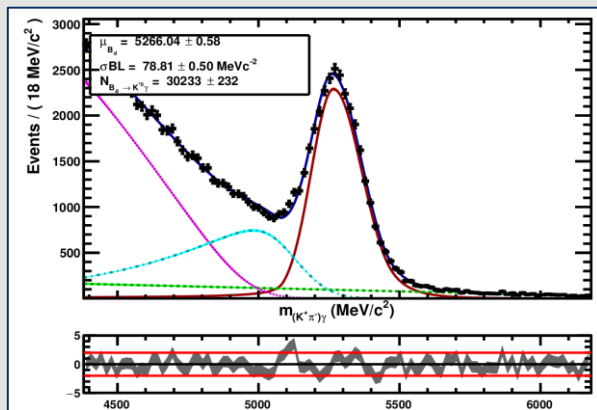
$B_s \rightarrow \phi \pi^0 \gamma$



SIMULTANEOUS FIT

2012 massfit

External constraints : $(m_B - m_{B_s})$ & $(m_B - m_{\Lambda_b})$



CONCLUSIONS AND PERSPECTIVES

Run 1 selection OK, few crosschecks left for the PID and MVA studies

Run 1 Yields according to preliminary fit :

▶ $B^0 \rightarrow (K^* \rightarrow K \pi) \gamma : \sim 38500$

▶ $B_s \rightarrow (\phi \rightarrow K K) \gamma : \sim 6000$

▶ $\Lambda_b \rightarrow \Lambda^* \gamma : \sim 6300$

▶ Next steps:

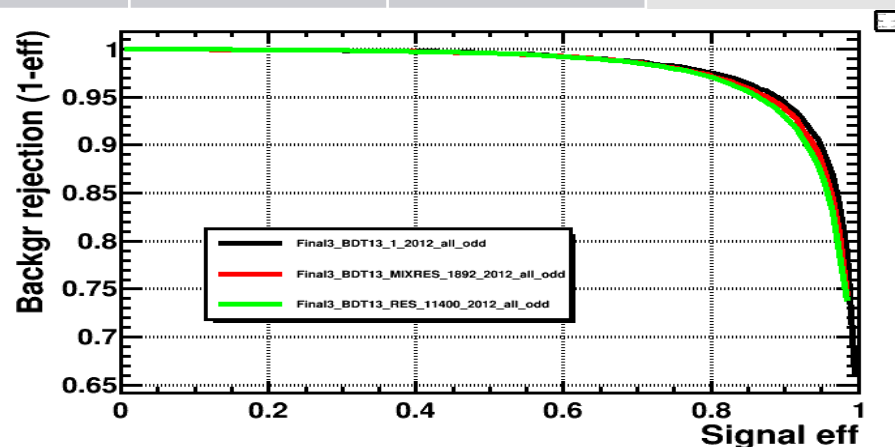
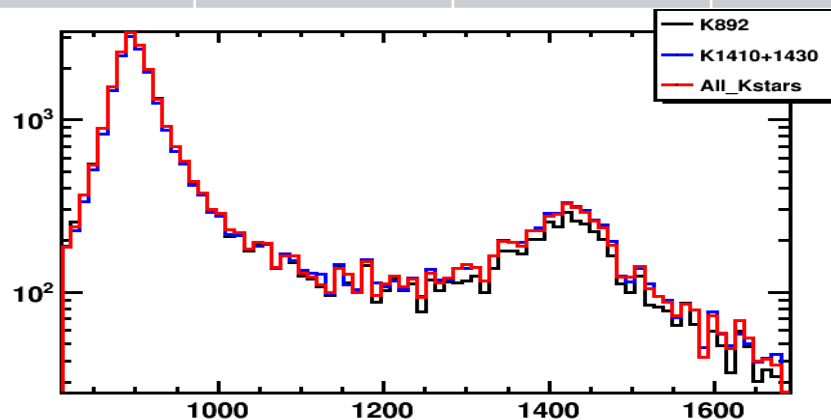
- Selection run2: MC production ongoing

- Still need to train BDT for 2011 with f_2 / K_{res} MC

- Background studies : ongoing

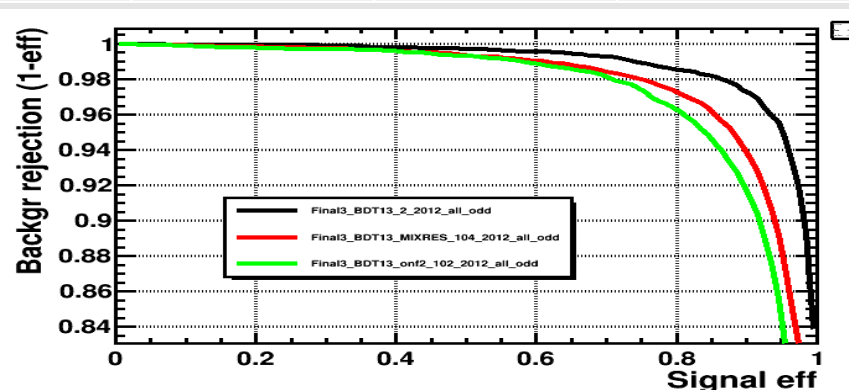
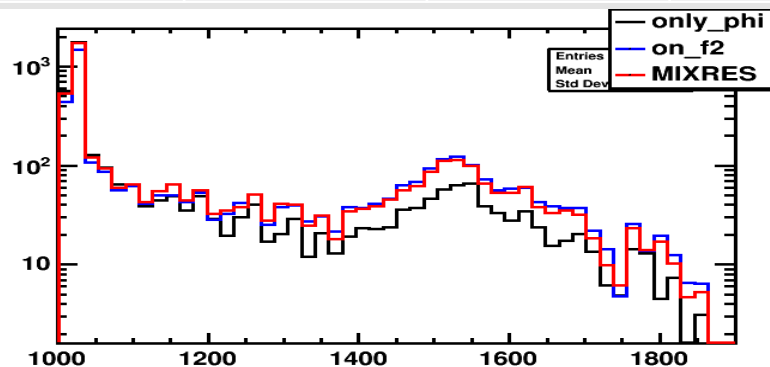
BDT ON K1(1410)/K2(1430)

BDT training MC	cut	Sig Eff	k1 eff	Rej_SB	Rej_Filtered SB
K2 + k1	$> -0,01$	$\sim 88,5 \%$	$\sim 88,7 \%$	$\sim 93,8 \%$	$\sim 95,6 \%$
$K^* + k2 + k1$	$> 0,01$	$\sim 89,6 \%$	$\sim 88,3 \%$	$\sim 93,8 \%$	Same #evt
Only K^*	> 0	$\sim 91 \%$	$\sim 79,6 \%$	$\sim 93,3 \%$	Same #evt



BDT ON F²(1525)

BDT training MC	cut	FoM	Sig Eff	F ² eff	Rej_SB	Rej_FilteredSB
Only f ²	>0,05	50	~81,3 %	~81,3 %	~95,8 %	~95,6 %
Mix f ² /phi	>0,07	53,3	~84 %	~76,3 %	~96 %	~95,6 %
Only phi	>0,07	58,6	~92 %	~43,6 %	~96	~95,6 %



In the end we choose to use the BDT trained on Mix f²/phi

BDT EFFICIENCY

Sample	BDTCut	Signal efficiency	Background rejec. (Test Samples)	Background rejec. (filtered SB)
Kπγ_2011	BDToutput > 0,04	94,2 %	88,8 %	89,2 %
KKγ_2011	BDToutput > 0,13	93,4 %	93,7 %	93,1 %
ρKγ_2011	BDToutput > 0,1	83,3 %	94,5 %	92,3 %
Kπγ_2012	BDToutput > 0	91,3 %	93,3 %	94 %
KKγ_2012	BDToutput > 0,07	92 %	96,5 %	95,6 %
ρKγ_2012	BDToutput > 0,07	79,2 %	97,6 %	94,8 %

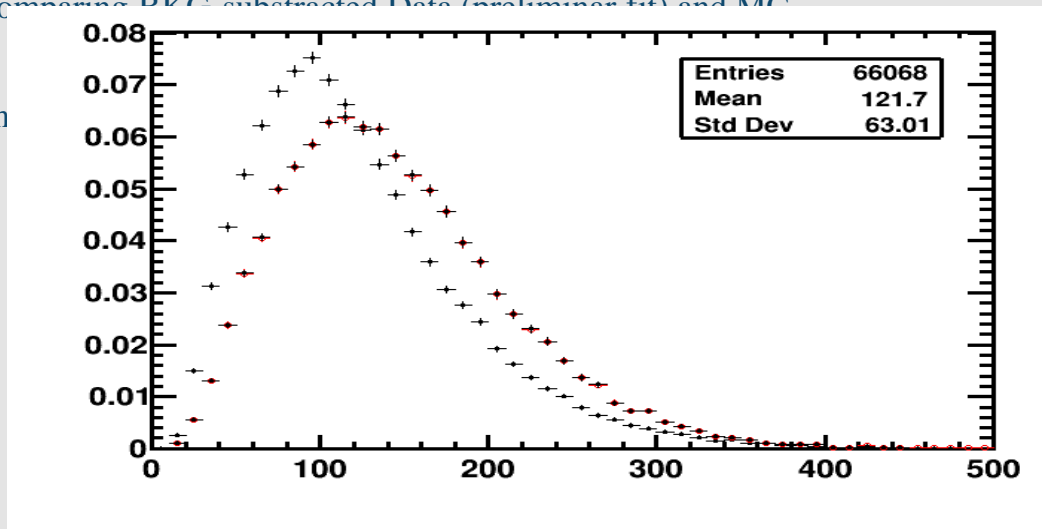
Bad efficiencies on Bs->K*2/K1 γ, Bs->f'2γ :

- efficiency of the BDT trained with Bs->φ γ MC on Bs->f'2γ MC : ~44% !
- equivalent for Bd -> k1 γ : ~80%

-> Train other BDTs with Mix of resonant states, and also only on f'2 / K1

NTRACKS REWEIGHTING

- ▶ Weights computed comparing R_{K^*} subtracted Data (preliminary fit) and MC
- ▶ distributions :
- ▶ - Example for $k^*\gamma$

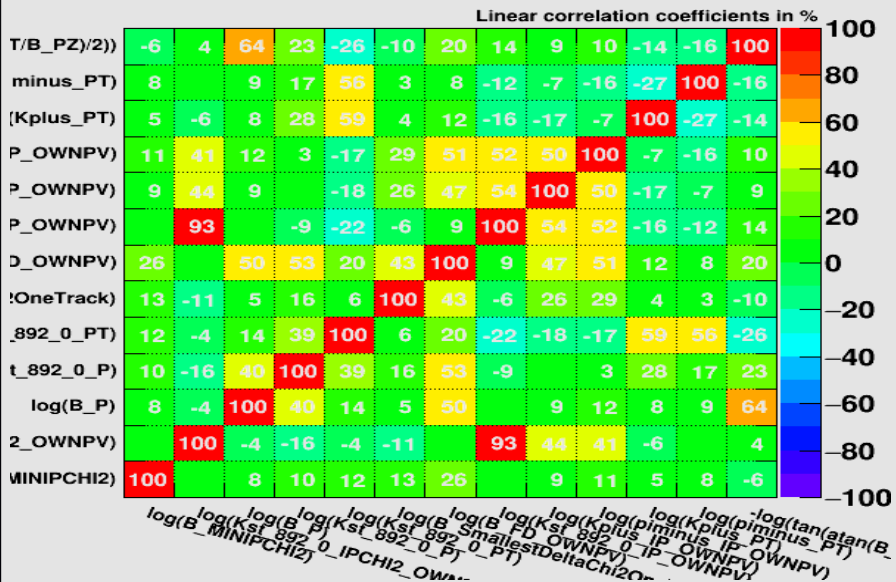


- ▶ - P & Eta well reproduced in MC, and not impacted by nTracks reweighting

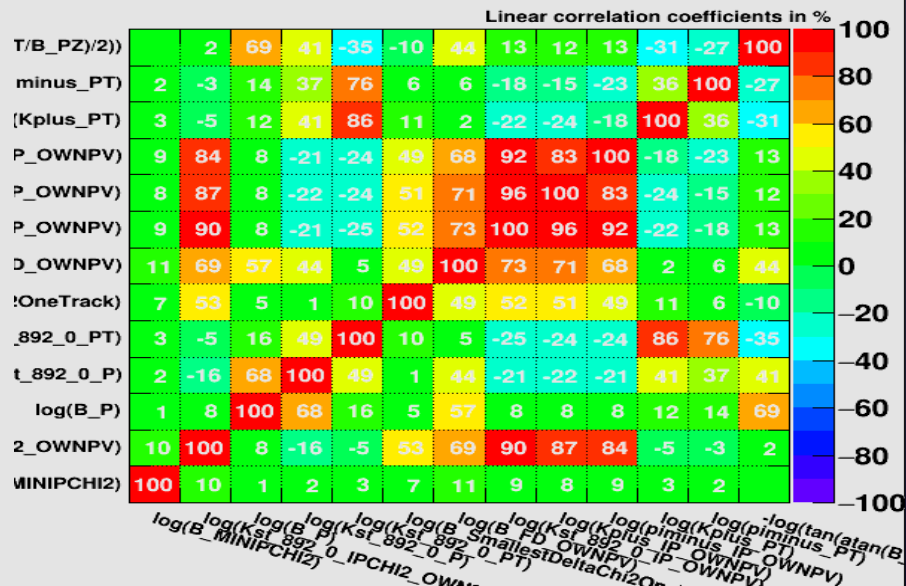
LINEAR CORRELATIONS

Some variables are highly correlated, yet seem necessary to keep good performances...
According to TMVA, BDT works fine with correlated variables...

Correlation Matrix (background)



Correlation Matrix (signal)

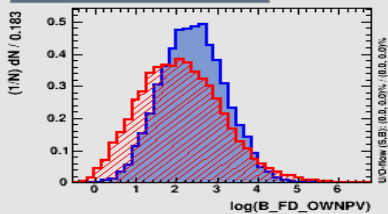


STRIPPING & PRESEL CUTS

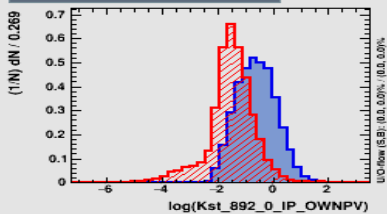
Common cuts	$K\pi\gamma$	$KK\gamma$	$pK\gamma$
h_MINIPCHI2 > 16	B_M01 < 1850	B_M01 < 1850	B_M01 < 3000
max(h1_PT,h2_PT) > 1200	// D0(Kpi)	// D0(KK)	
min(h1_PT,h2_PT) > 500		B_M02_Subst2_gamma2pi0 >2000	
h_P > 500	B_M02_Subst2_gamma2pi0 >2000 //D+(K+pi0)	//D+(K+pi0) (& K*+)	
h_P < 100000	(& K*+)		
h1_eta > 1.5	B_M12_Subst2_gamma2pi0 >2000	B_M12_Subst2_gamma2pi0 >2000	
h1_eta1+ < 5.0	// D-(pi-pi0) (& rho-)	// D-(K-pi0) (& K*-)	
res_ENDVERTEX_CHI2 < 9			
h_TRACK_CHI2NDOF < 3			
h_TRACK_GhostProb < 0.4			
B_DiraAngle < 0.06			
B_PT > 2000			
B_IPCHI2_OWNPV < 9.0			
gamma_CL > 0.2			
gamma_PP_IsPhoton > 0.6			
gamma_PT > 3000			

BDT VARIABLES

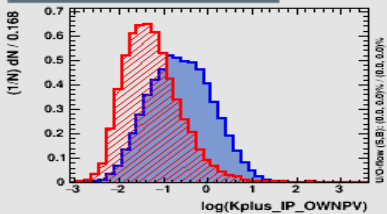
Input variable: $\log(B_FD_OWNPV)$



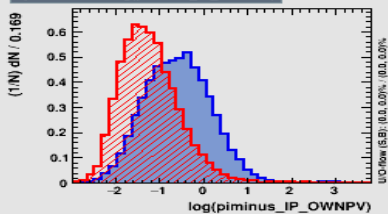
Input variable: $\log(Kst_892_0_IP_OWNPV)$



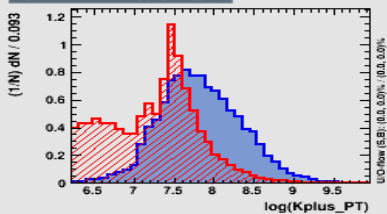
Input variable: $\log(Kplus_IP_OWNPV)$



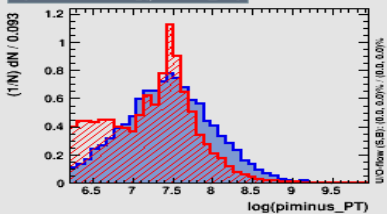
Input variable: $\log(piminus_IP_OWNPV)$



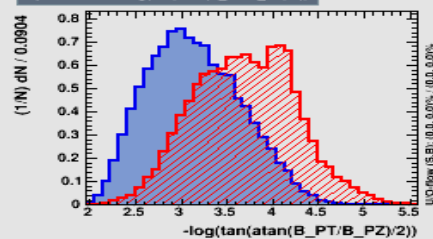
Input variable: $\log(Kplus_PT)$



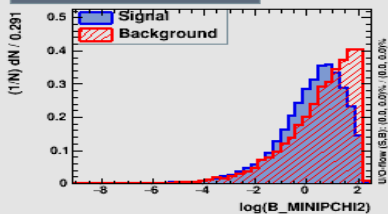
Input variable: $\log(piminus_PT)$



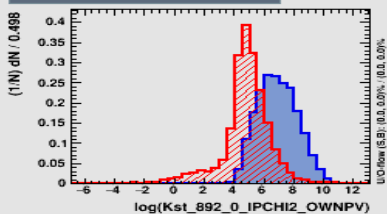
Input variable: $-\log(\tan(\text{atan}(B_PT/B_PZ)/2))$



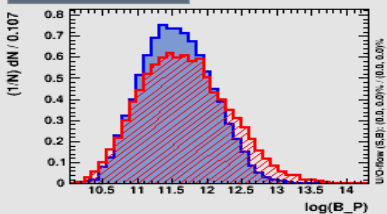
Input variable: $\log(B_MINIPCHI2)$



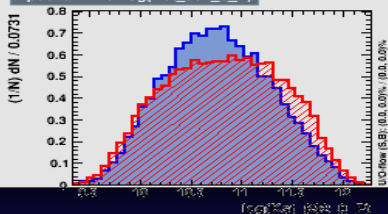
Input variable: $\log(Kst_892_0_IPCHI2_OWNPV)$



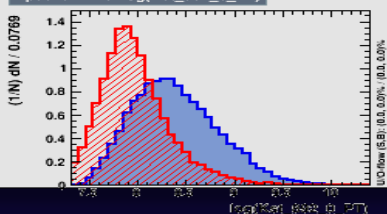
Input variable: $\log(B_P)$



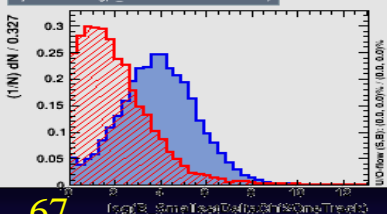
Input variable: $\log(Kst_892_0_P)$



Input variable: $\log(Kst_892_0_PT)$



Input variable: $\log(B_SmallestDeltaChi2OneTrack)$



ANALYSIS SCHEME (2)

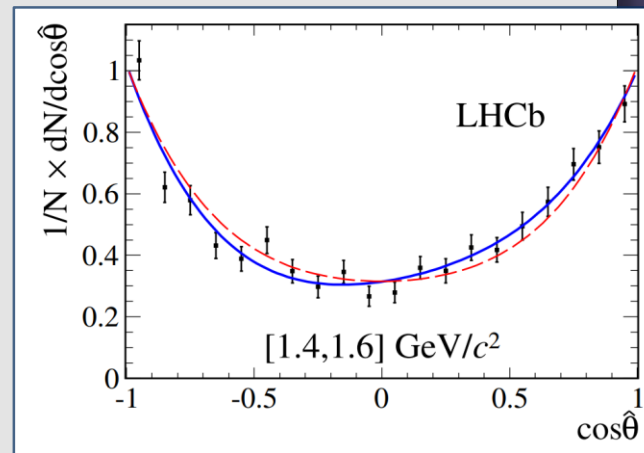
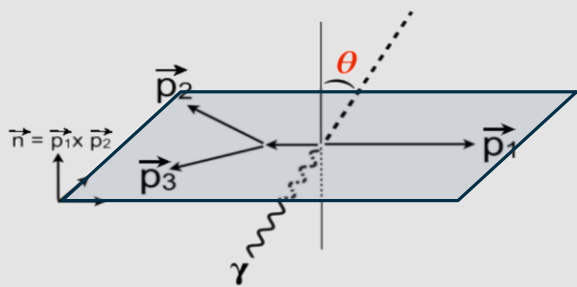
3. PID studies :
4. Develop MVA selection
5. Fit B masses
6. Fit (hh) mass VS helicity



PHOTON POLARIZATION (KSHHG)

On peut mesurer la distribution angulaire d'émission du photon par rapport au plan hadronique de l'état final :

$$\frac{d\Gamma}{ds ds_{13} ds_{23} d\cos\theta} \propto \sum_{i=0,2,4} a_i(s, s_{13}, s_{23}) \cos^i \theta + \lambda_\gamma \sum_{j=1,3} a_j(s, s_{13}, s_{23}) \cos^j \theta.$$



(PHOTON POLARIZATION)

Une analyse basée sur la dépendance temporelle de la largeur de désintégration :

$$\Gamma_{B(\bar{B})_{(s)}^0 \rightarrow \Phi^{CP} \gamma}(t) = |A|^2 e^{-\Gamma_{(s)} t} \left(\cosh(\Delta\Gamma_{(s)} t/2) + \mathcal{A}_\Delta \sinh(\Delta\Gamma_{(s)} t/2) \right. \\ \left. \pm \mathcal{C}_{CP} \cos(\Delta m_{(s)} t) \mp \mathcal{S}_{CP} \sin(\Delta m_{(s)} t) \right) \quad \tan\Psi = \left| \frac{\mathcal{A}_R}{\mathcal{A}_L} \right|$$

Avec :

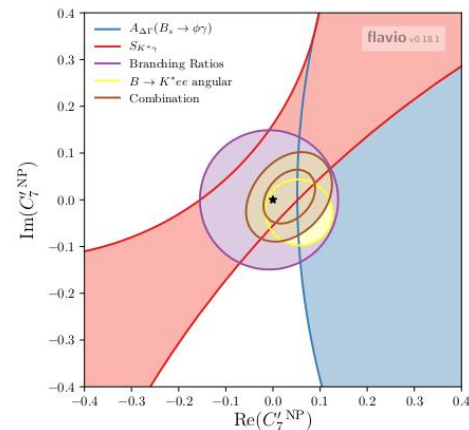
$$\mathcal{S}_{CP} \sim \sin 2\Psi \sin \phi_{(s)}$$

$$\mathcal{A}_\Delta \sim \sin 2\Psi \cos \phi_{(s)}$$

, $\phi_{(s)}$ la phase de violation de CP, et

+ plot Carla workshop kshhg ?

A^Δ is sensitive to $\text{Re}(C_7^{\prime NP})$



Constrains from radiative

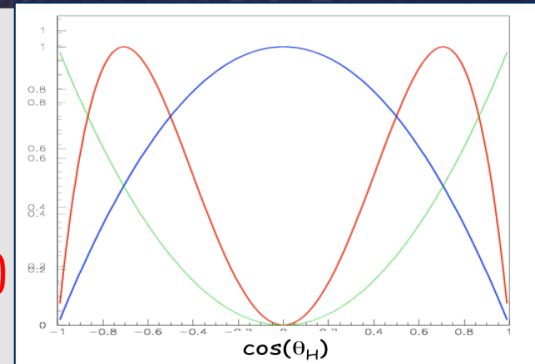
(2D MASS FIT)

Make use of the helicity angle

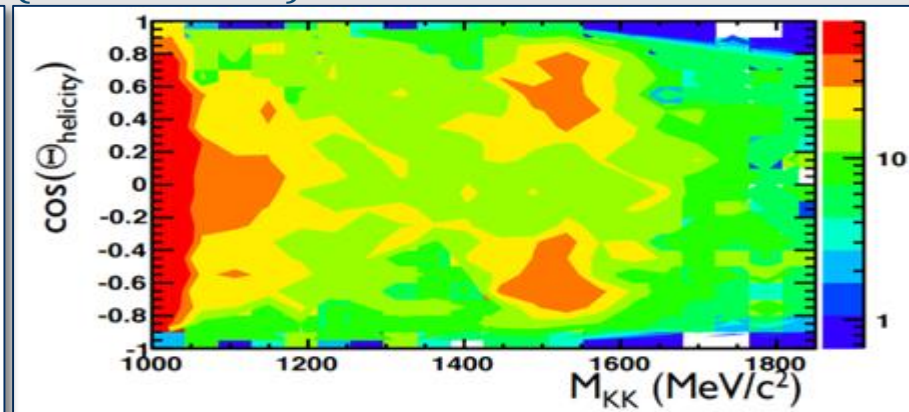
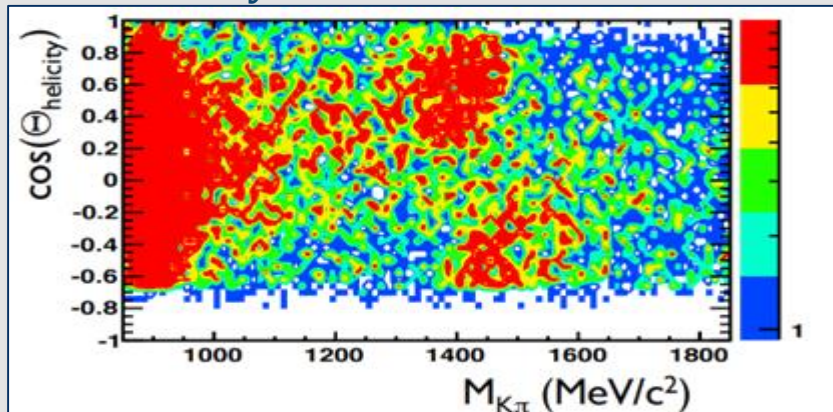
$\text{Cos}^2(\theta_H)$ for S-waves : π^0 backgrounds

$\text{Sin}^2(\theta_H)$ for P-waves : $K^*(892)\gamma$, $K_1(1410)\gamma$, $\phi(1020)\gamma$

$\text{Cos}^2(\theta_H).\text{sin}^2(\theta_H)$ for D-waves : $K^*_2(1430)\gamma$, $f'_2(1525)$



Preliminary look to the data run1+run2 (2.7 + 1.0 fb⁻¹) :



- Clear D-wave contribution in the $K^*_2(K\pi)$ and $f'_2(KK)$ region
- Expecting a first observation of $B_s \rightarrow f'_2(1525)\gamma$ and $B^0 \rightarrow K_1(1410)\gamma$

TRIGGER SYSTEM

Bunch crossing too high for the detector to record

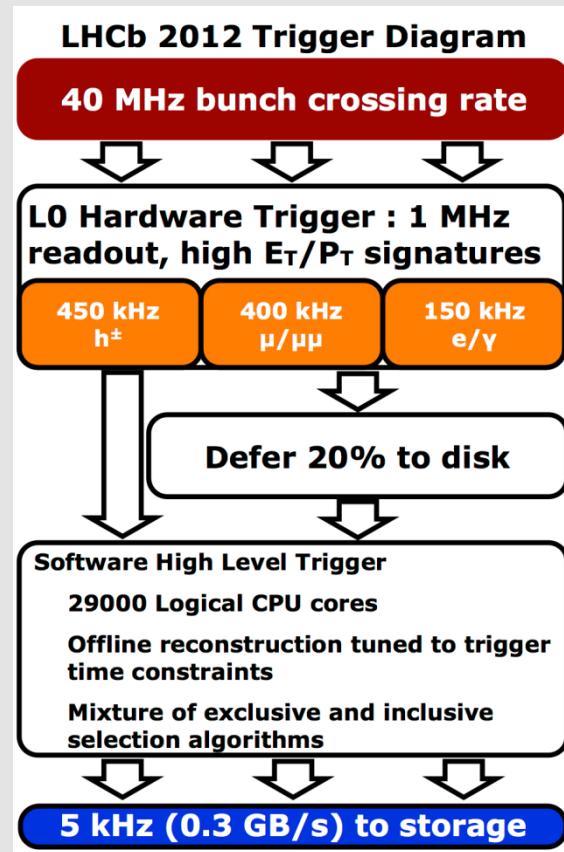
All -> Trigger selection implemented in the hardware

based on fast informations

(details(deposits))exemple tck ?

After reconstruction of event :

Further selection at HLT before storage



ANALYSIS SCHEME

1. Use data from inclusive line hh gamma :

Trigger :

L0 : L0 photon		L0 electron	(TOS)
HLT1 : Hlt1trackAllL0		Hlt1trackPhoton	(TOS)
HLT 2 : Hlt2RadiativeTopoTrack		Hlt2RadiativeTopoPhoton	(TOS)

Stripping :

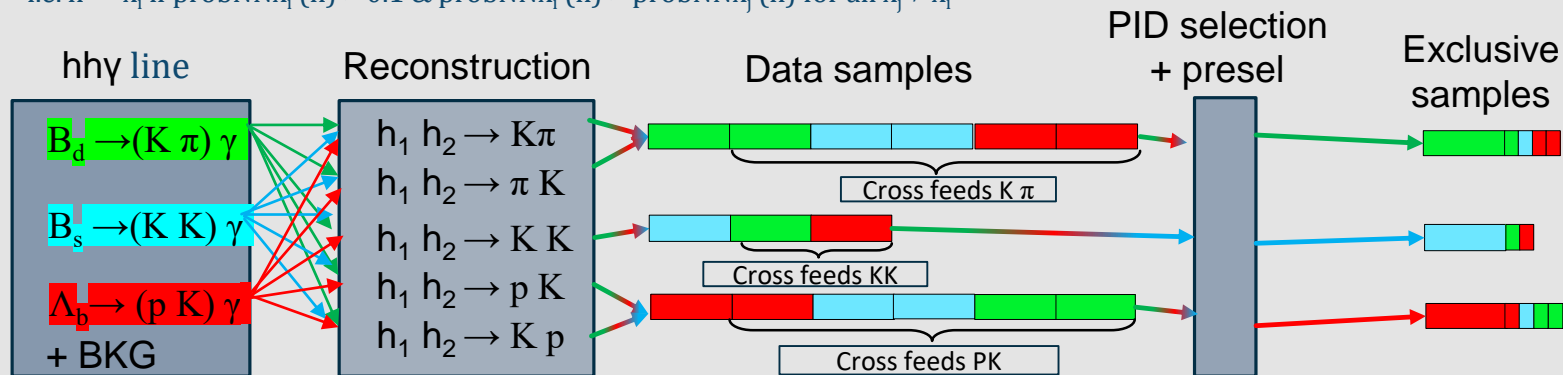
Inclusive hhy (2pi_Line)

Reconstruct the 2h gamma categories :

$K\pi$ & KK & pK hypothesis reconstructed by applying PID-substitution to stripped candidates

Classify $h^+ h^- \gamma$ candidates according to largest h -ProbNN $_i$ ($x_i = \pi, K, p$)

i.e. $h = x_i$ if $\text{probNN}_{x_i}(h) > 0.1$ & $\text{probNN}_{x_i}(h) > \text{probNN}_{x_j}(h)$ for all $x_j \neq x_i$



BDT SELECTION

- ▶ Principles (slides master2 ?)
- ▶ Combi bkg selection
- ▶ Training
- ▶ Testing
- ▶ FoM
- ▶ Efficiency