

# P5' Status Report

Reamining systematics and approximate profiled FC  
or “How we spent our xmas vacation”

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for the  $P'_5$  group

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- We have estimated the remaining systematics uncertainties, only one missing;
  - ▶ Efficiency shape
  - ▶ Mass distribution
  - ▶ Feed-through background
  - ▶ Uncertainty from fixed pdf parameters **Still missing**
    - ★ Will describe the proposed solution, after the discussion with the StatCom
- Statistical uncertainties
  - ▶ After long and hard discussion with StatCom, against our scientific opinion, validated with data, the outcome was that **none** of the many methods we proposed to estimate the statistical uncertainties were good enough,
  - ▶ we had implement yet a different method to compute the statistical uncertainty;
    - ★ the proposed solution is to build the Neyman construction
    - ★ the full 2D construction is simply impossible: time estimate  $\mathcal{O}(\text{years})$
    - ★ instead, we agreed upon a 1D FC approach by profiling the likelihood on data at fixed  $P_1$  ( $P'_5$ ) values.
    - ★ here will describe the procedure and preliminary results;

# Systematics uncertainties: Efficiency shape

- Fit high statistics control regions ( $J/\psi/\psi(2S)$ ) using the efficiency and compare  $F_L$  with word-average values
  - $J/\psi$  165 000 signal events
    - $F_L = 0.537 \pm 0.002$  (stat) vs  $0.571 \pm 0.007$  (stat+syst)
  - $\psi(2S)$  for completeness also here, lower stat:
    - $F_L = 0.538 \pm 0.008$  (stat) vs  $0.463^{+0.028}_{-0.040}$  (stat+syst)
- Then uncertainties on  $F_L$  is propagated (with 200 toy experiments) to other bins

$q^2$ bin	$P_1$	$P'_5$
0	$\pm 0.017$	$\pm 0.005$
1	$\pm 0.048$	$\pm 0.060$
2	$\pm 0.093$	$\pm 0.065$
3	$\pm 0.094$	$\pm 0.045$
5	$\pm 0.083$	$\pm 0.059$
7	$\pm 0.100$	$\pm 0.060$
8	$\pm 0.068$	$\pm 0.041$

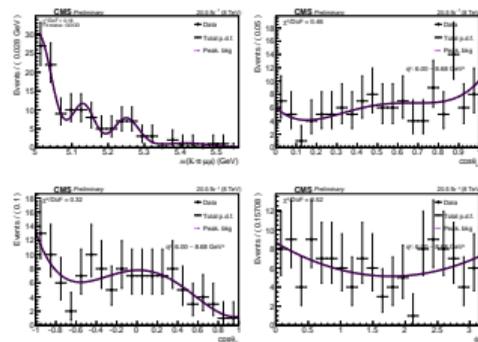
## Systematics: Mass distribution

- it is used in the final fit on data
  - ▶ pdf used is a **two-gaussian with common mean**
  - ▶ separately for correctly and wrongly tagged events
  - ▶ the parameters (mean, four  $\sigma$  and two ratios) are taken from the high statistics MC
- we use the two control samples ( $J/\psi, \psi(2S)$ ) to fit all params of the mass distribution and compare the results on  $P_1$  and  $P'_5$
- The maximum changes in the measured values in the two control channels when the parameters are varied are taken as the systematic uncertainty for all  $q^2$  bins.
- The maximum change of  $P_1$  is 0.012, of  $P'_5$  is 0.019.

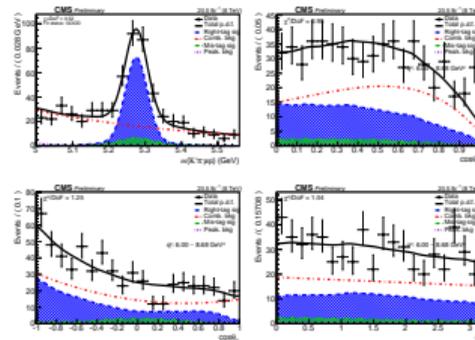
# Systematics: Feed-through background

- In bins just below or above the resonances, there could be a contamination from  $B^0 \rightarrow J/\psi K^{*0}$  and  $B^0 \rightarrow \psi' K^{*0}$  feed-through events
- Feed-through is modelled by a dedicated pdf for bin 3 and 5, using control region simulation

An example for bin 3 just below the  $J/\psi$  control region



Fit with additional pdf component



(barely visible in purple)

$q^2$ bin	P1	P5p
3	$\pm 0.004$	$\pm 0.012$
5	$\pm 0.012$	$\pm 0.020$
7	$\pm 0.011$	$\pm 0.024$

- In the final pdf some of the parameters ( $F_L$ ,  $F_S$ , and  $A_S$ ) are fixed from BPH-13-010
- Our initial idea was to get uncertainties (and correlation) from previous work (based on same dataset) and propagate to  $P_1$  and  $P'_5$ 
  - ▶ fit with fixed values randomly chosen via a 3D gauss around ( $F_L$ ,  $F_S$ , and  $A_S$ )
  - ▶ preliminary results were produced (not the most important syst for any bin)
  - ▶ but we got some criticism from StatCom:
    - ★ we would have not taken into account the correlation between the fixed
- Different approach based on toy (once again!)
  - ▶ generate a large statistics [ $\mathcal{O}(100 \times \text{Data})$ ] toy using as pdf the one with data best fit parameters;
  - ▶ fit the toy with all parameters free to float
  - ▶ compare the statistical errors of  $P_1$  and  $P'_5$  with the ones of a fit with three params fixed;
  - ▶ **sys uncert to reproduce the scale factor**
  - ▶ the scale factor between free and partially-fixed fit is precisely the correlation coefficient
    - ★ see e.g. [Bivariate Normal Distribution](#)
- Unless there are objection, we will proceed with this second option.

One dimensional Feldman-Cousins approach, profiling the likelihood on data at fixed  $P_1$  ( $P'_5$ ) values.

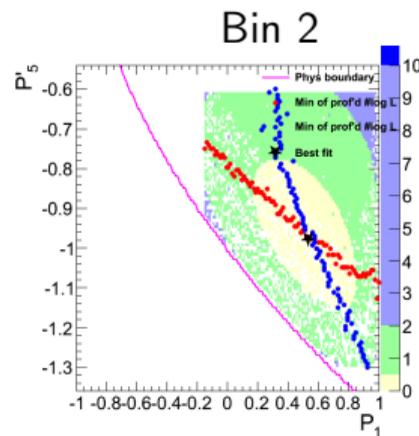
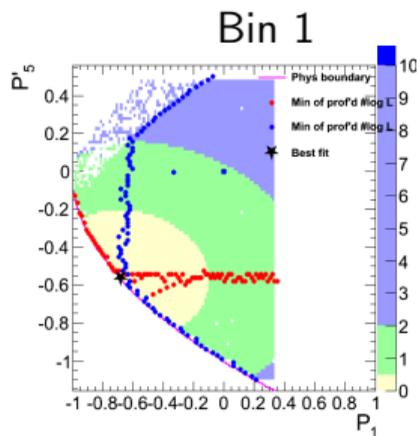
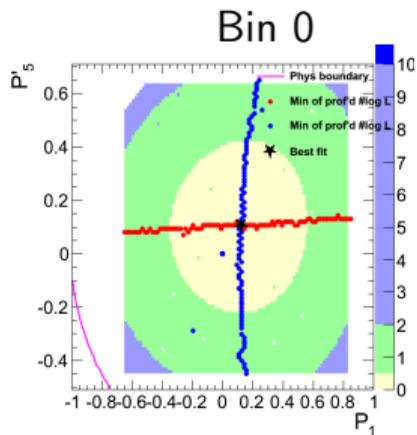
## Procedure description in the PAS

To ensure correct coverage for the uncertainties of the angular observables, the Feldman-Cousins method <sup>[?]</sup> is used with nuisance parameters. Two main sets of pseudo-experimental samples are generated to compute the coverage for the two angular observables  $P_1$  and  $P'_5$ , respectively. The first (second) set, used to compute the coverage for  $P_1$  ( $P'_5$ ), is generated by assigning values to the other observables as obtained by profiling the likelihood on data at fixed  $P_1$  ( $P'_5$ ) values. When fitting the pseudo-experimental samples the same fit procedure as in data is applied.

What it means in the following, together with preliminary results

# Procedure description

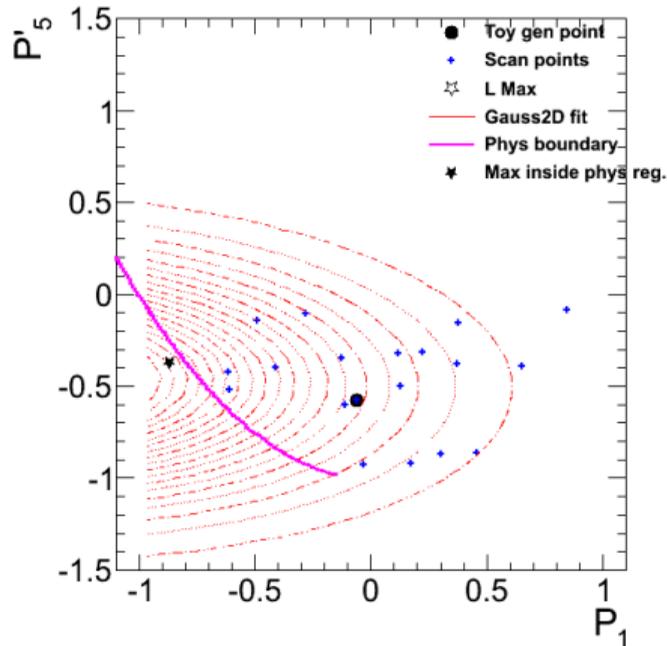
- start from the 2D  $\mathcal{L}(P_1, P'_5)$  computed on data, taking into account the physical boundaries
- Then we profile it vs  $P_1$  and  $P'_5$ , respectively
  - ▶ if we hit a physical boundary, the minimum can be along the boundary itself
- Then we generate 100 (data-like size) toys using as input parameters  $P_1$  and  $P'_5$ .
  - ▶ To save CPU time not for all points, but we start around  $\Delta \log \mathcal{L} = 0.5$



# Procedure description (cont'ed)

Each toy is fitted with the full pdf as done for data

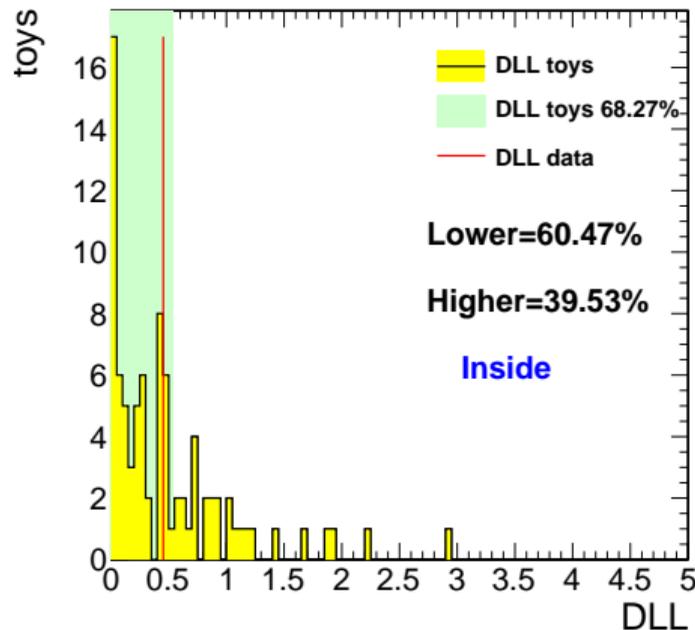
- we repeat the fit with 20 different set of 20 initial values of  $P_1$  and  $P'_5$ 
  - the 20 points are chosen randomly a 2D gaussian distribution around the toy generator point;
  - or around the min of a fully free fit if it converges
  - the results of the 20 fits provide 20 likelihood values in 20 sets of  $(P_1, P'_5)$ ,
  - to find the absolute max, we fit the 20 values with a 2D gauss function
  - the max must be inside the physical region
- Eventually, we have 100 toys, and 100 values for the likelihood.



# Procedure description (cont'ed)

- We compute  $\Delta \log \mathcal{L}$  for each toy (compared with the min along the profile) [black histo] →
- and  $\Delta \log \mathcal{L}$  for data for that gen point [red line] →
- **ratio** = (# toys with  $DLL(\text{toy}) < DLL(\text{Data})$ ) / (#toys)
- If **ratio** < 68.27% [green area] → then generation point is inside the  $1\sigma$  boundary for data, otherwise it's outside.
  - ▶ In principle there should be 100 toys
  - ▶ some failure (10-15%) due to gauss2D fit failure to be investigated
  - ▶ some job failing (batch system): recover
  - ▶ should we fit the  $DLL(\text{toys})$  distribution? With what?  $\chi^2(\#\text{DoF})$ ?
- repeat for  $P_1(P'_5)$  upper(lower) bound: 4 "directions"

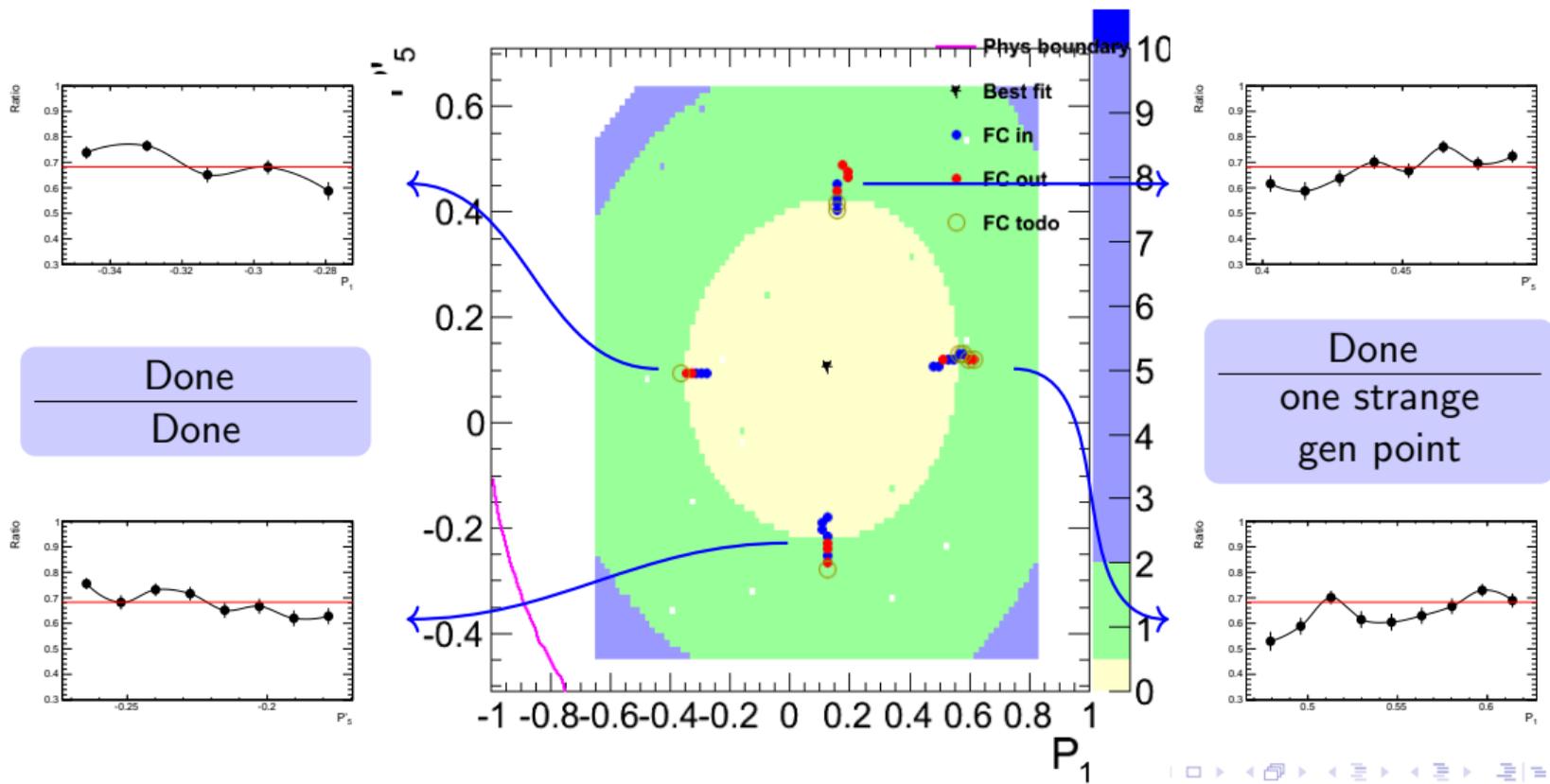
One example of DLL toy distribution compared with **DLL(Data)** (red)



## Some numbers

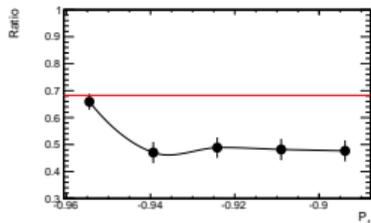
- The procedure is complex, and very time consuming
  - ▶ Each toy, for which we scan 20 points, takes about 1h of CPU
  - ▶ Each Gen Point has 100 toys
  - ▶ We need to evaluate approx 5-10 GenPoints per directions (x4), for each bin (x7)
  - ▶ **The math is left as a simple exercise to to class**

hint  $\approx 3.5 \cdot 10^5$  fits,  $> 10k$  jobs,  $\mathcal{O}(1 \cdot 10^5)$  CPU-hours, babysitting time *you-don't-want-to-know*
- Lot of babysitting!
  - ▶ Each time a set of GenPoint finish, we have to evaluate if we crossed the  $1\sigma$  boundary
  - ▶ if not, decide which new GenPoints should be submitted
- On top of that, the local batch system we are using is not working well
  - ▶ If we submit at once too many jobs, it just collapses, leaving the jobs in a weird state which must be recovered by hand. Happened twice last week.
  - ▶ We have a full support from local IT, so situation is improving, but still quite close to a nightmare
- Have I said lot of babysitting? **It is not, by all means, just CPU**

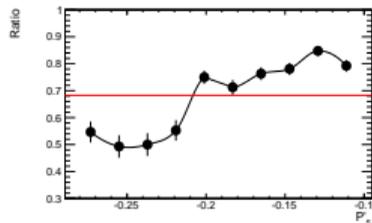
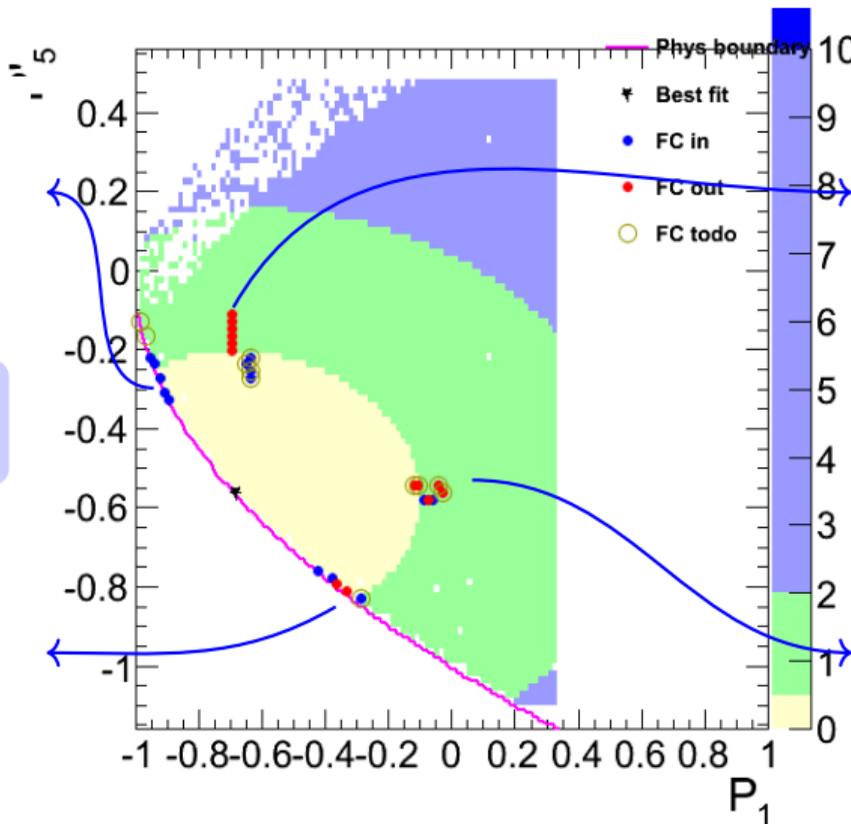
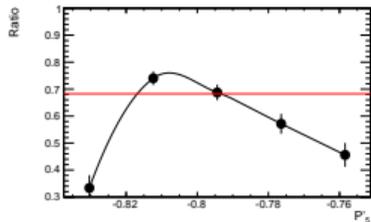


Done  
Done

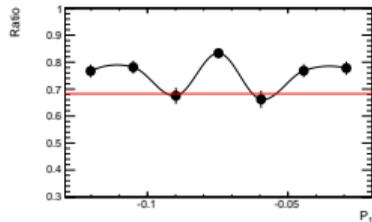
Done  
one strange  
gen point

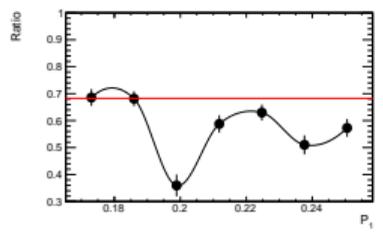


Need more GP  
Strange GP

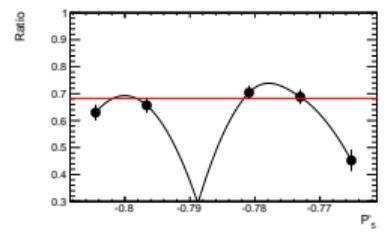
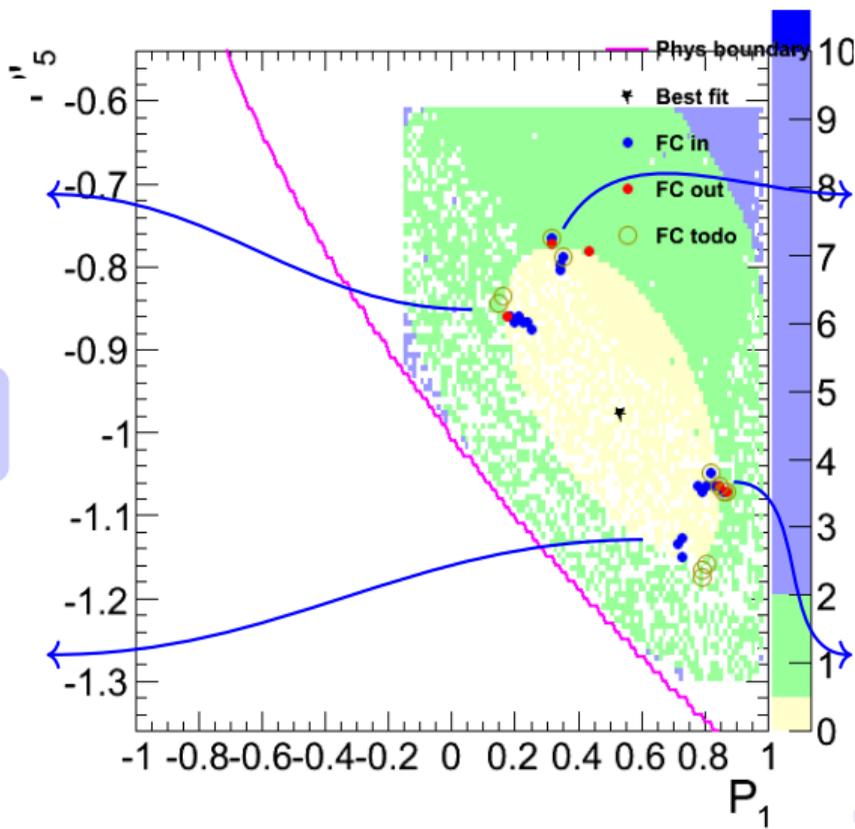
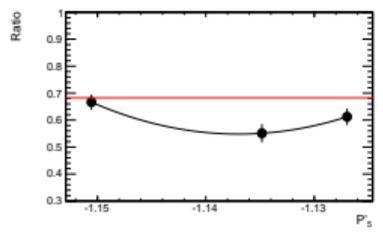


Done  
More GP:  
fluctuations?

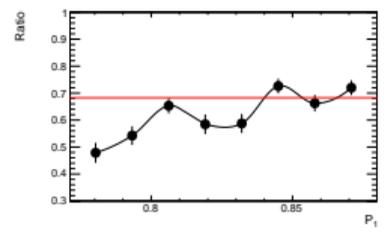


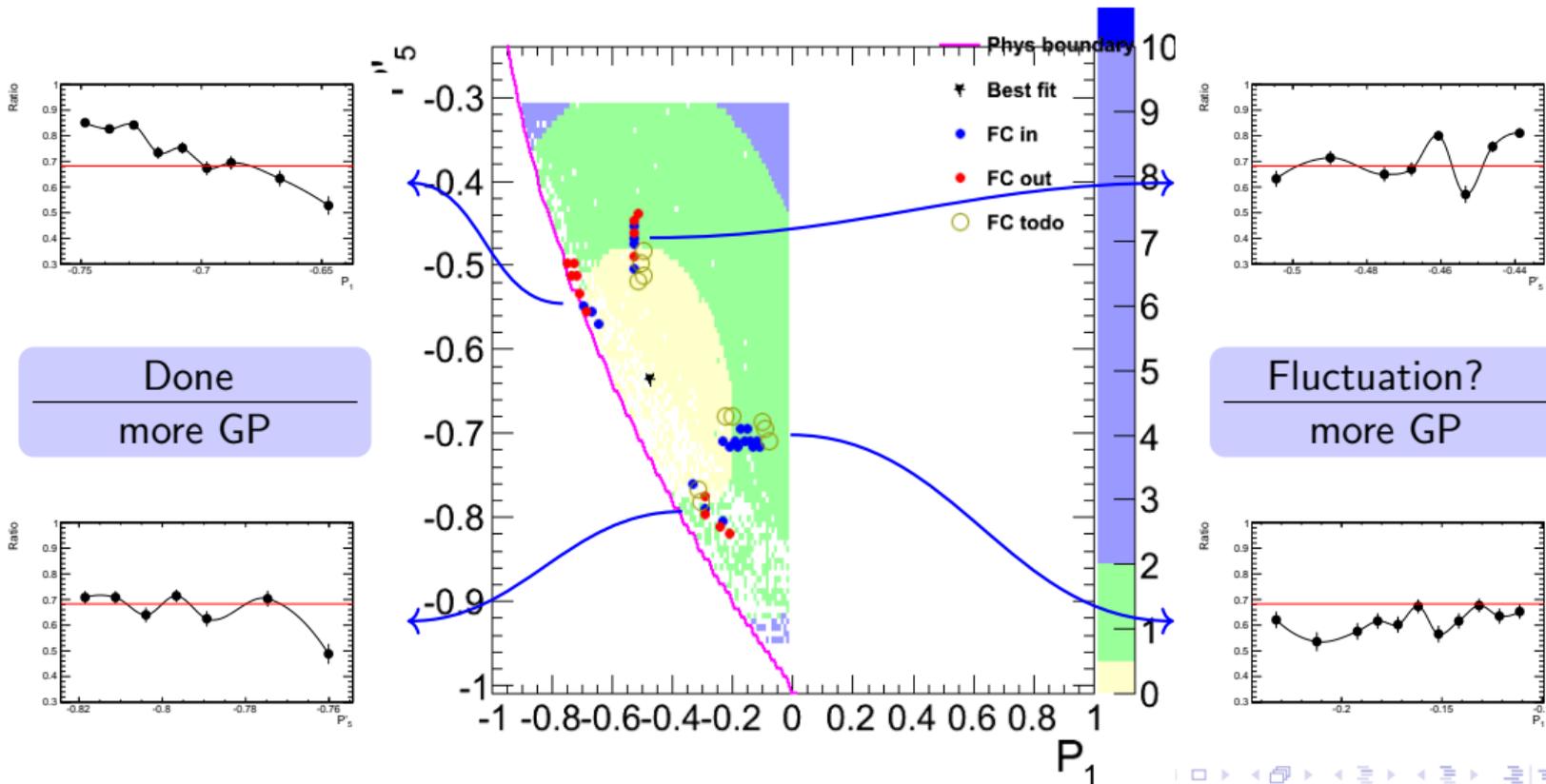


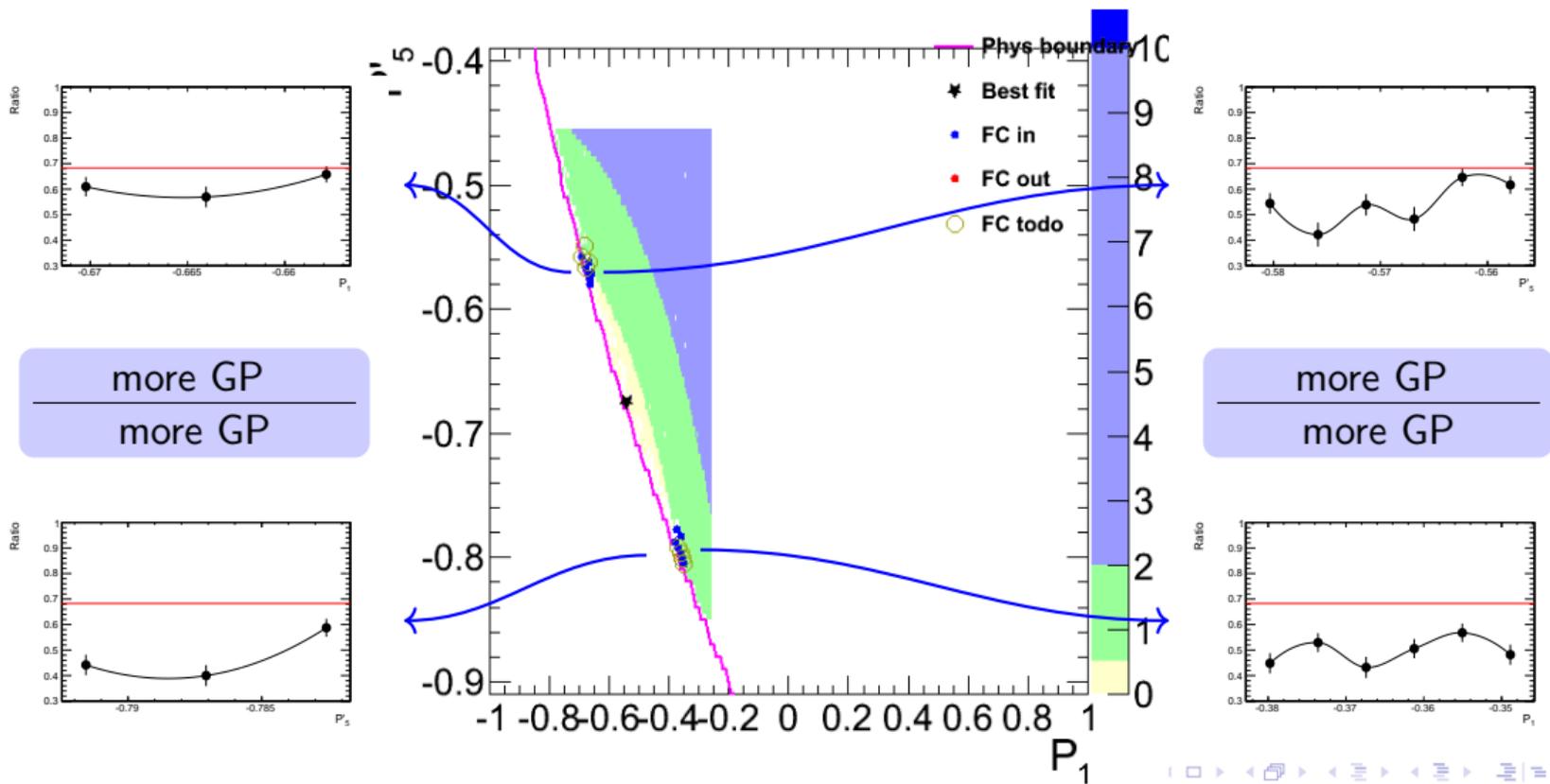
More GP  
-----  
more GP



More GP:  
fluctuations?  
-----  
Almost done

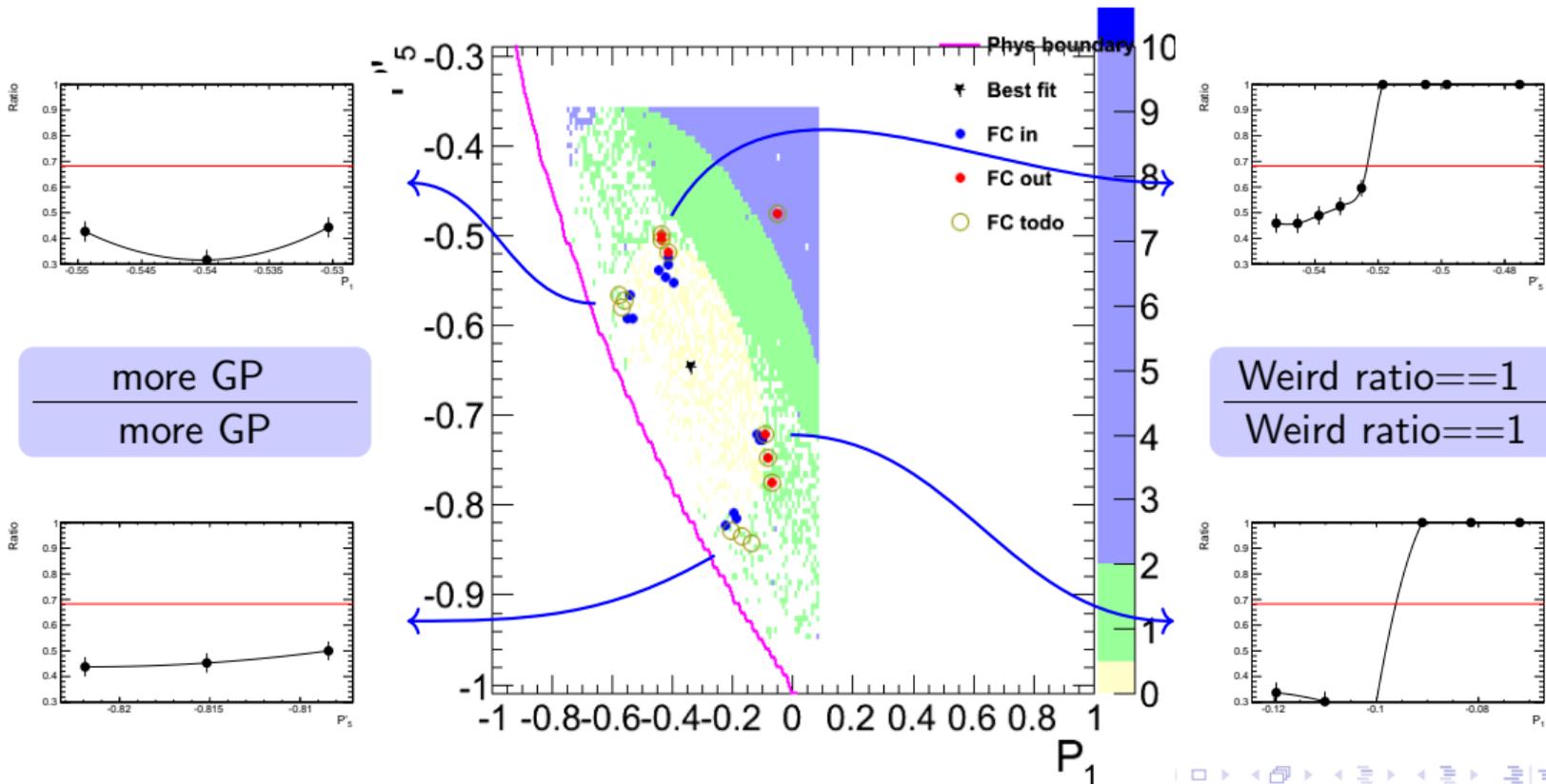


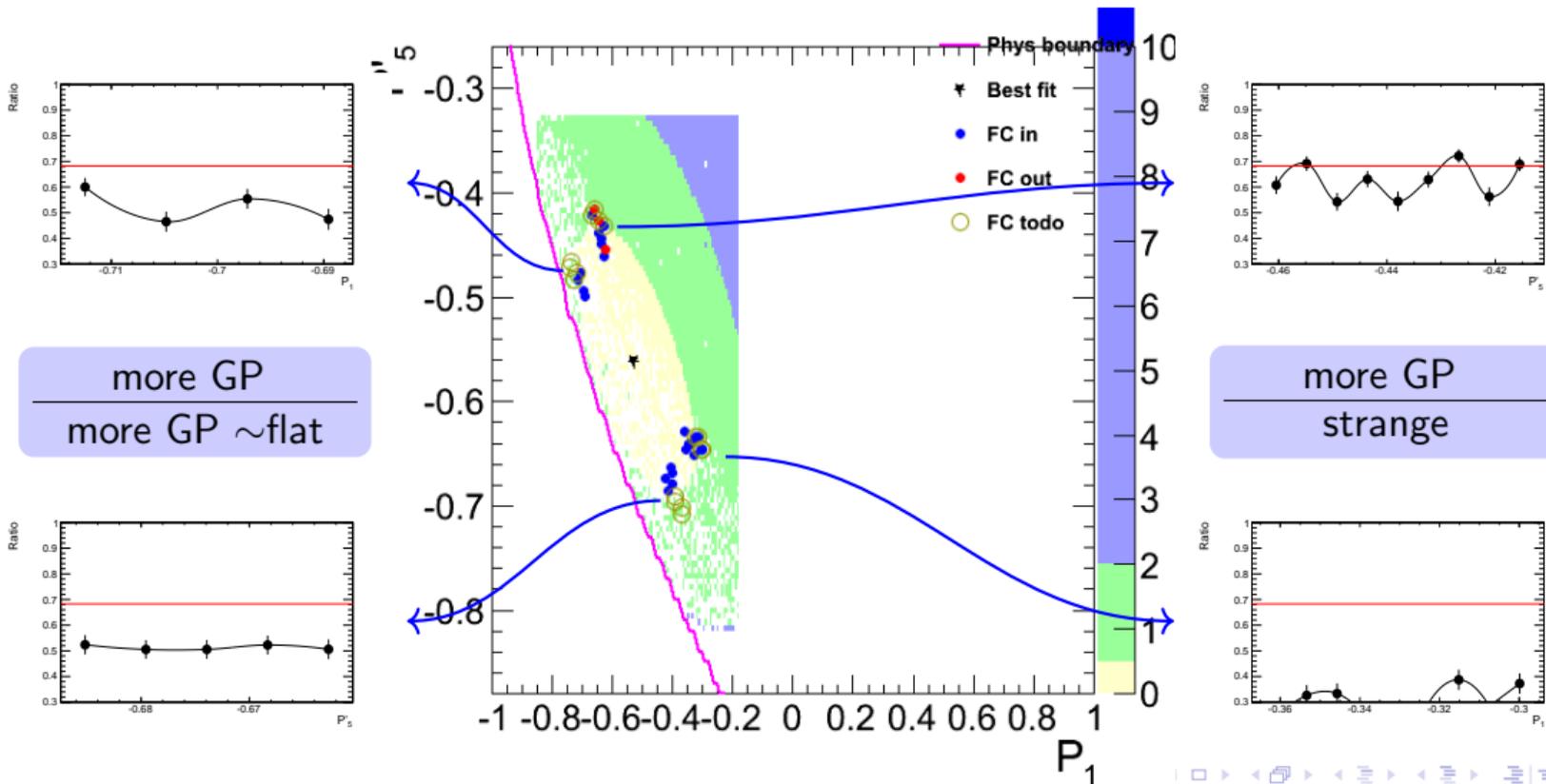




more GP  
more GP

more GP  
more GP





more GP  
 more GP  $\sim$  flat

more GP  
 strange

## Statistical uncertainties

- We have implemented the 1D likelihood profiling FC
- Running it takes an awful lot of CPU and **human** time in babysitting;
- Preliminary results have been shown;
  - ▶ The crossing from inside to outside  $1\sigma$  is not as clear as expected
    - ★ More gen point to be evaluated already ran, to be resubmitted
    - ★ Try a linear fit of the ratio;
    - ★ Try to increase the # of toys more CPU and more babysitting, hurrah!
    - ★ Investigate the failed gauss2D fit
  - ▶ When the crossing is clear, the  $1\sigma$  boundaries are quite close to the  $\Delta \log \mathcal{L} = 0.5$  surprise!<sup>a</sup>

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<sup>a</sup>No, wait. We did prove that, no surprise, sorry.

## Systematics uncertainties

- Almost all done;
- only one missing, but we have a plan how to evaluate it;
- it should not take long, just a matter to find the time to run the free fit on large toy.

## Documentation

- We are updating constantly the documentation (PAS and AN) as soon as we have stable results;
- **We still strongly want to go to Moriond!**



Additional or backup slides

