

# RooStats validation results

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# Outline

- Our interface to RooStats
- Validation results
  - Profile Likelihood method
  - Bayesian methods
  - Hybrid methods
- Summary
- A full table with all the numbers is also attached to the agenda

# CMS Higgs interface to RooStats

A binary package wrapping RooStats

- Takes as input either a simple text datacard for counting experiments (same format as L&S) or any RooStats HighLevelFactory file.
- Configures and runs RooStats methods, prints results and saves them to root files.
- Takes care of generating toys for expected limits, or averaging results of multiple runs.

# Validation: code

- Validation done on ROOT 5.27.06 (64bit/gcc434)
- Another set of validation will be done on 5.28 once it's integrated with the CMS software. (but we already started to look at it)
- We tested mainly the following routines:
  - ProfileLikelihoodCalculator
  - BayesianCalculator, MCMCCalculator
  - HybridCalculatorOriginal
  - We had a quick look also at the new HybridCalculator, but not at the new features only in 5.28.

# Validation: observables

The validation was twofold:

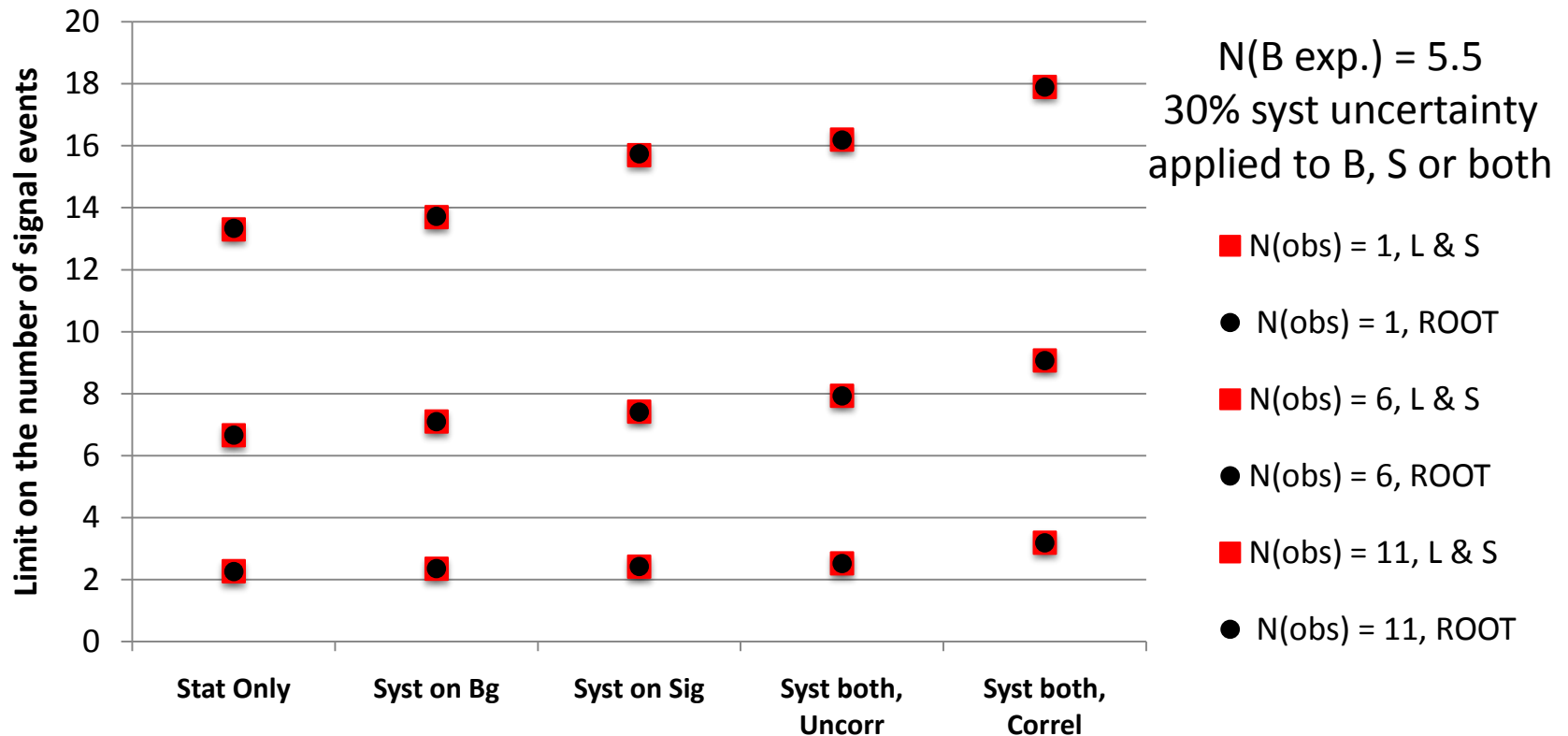
- Check that we get results in agreement with the reference ones from L&S, or from semi-analytical results for the simple cases.
- Check what can be computed with a reasonable amounts of resources per job.
- For the second part, we should probably follow up on the technicalities with roostas developers (not in this meeting).

# Profile Likelihood method

- **The results agree to high precision with what computed with the independent tool L&S.**
- **It's fast, below 1s even for complex models.**
- On rare occasions we've found numerical issues that cause the computation to fail. Solved switching between Minuit2 and Minuit, or to making very tiny changes to the variable ranges or inputs.

# Profile Likelihood method

Comparison for simple counting experiments with different  $n(\text{obs})$  and syst. uncertainties



# Bayesian Methods: BayesianCalculator

- **Allows only very few nuisance parameters.**
- **In the simple counting experiment tests, the results in good agreement with L&S, both for flat prior and prior prop. to  $1/\sqrt{\sigma}$**
- Whenever it works, the performance is good.
- Anyway, due to the limitation, it won't be of much use for any real higgs combination that includes systematical uncertainties

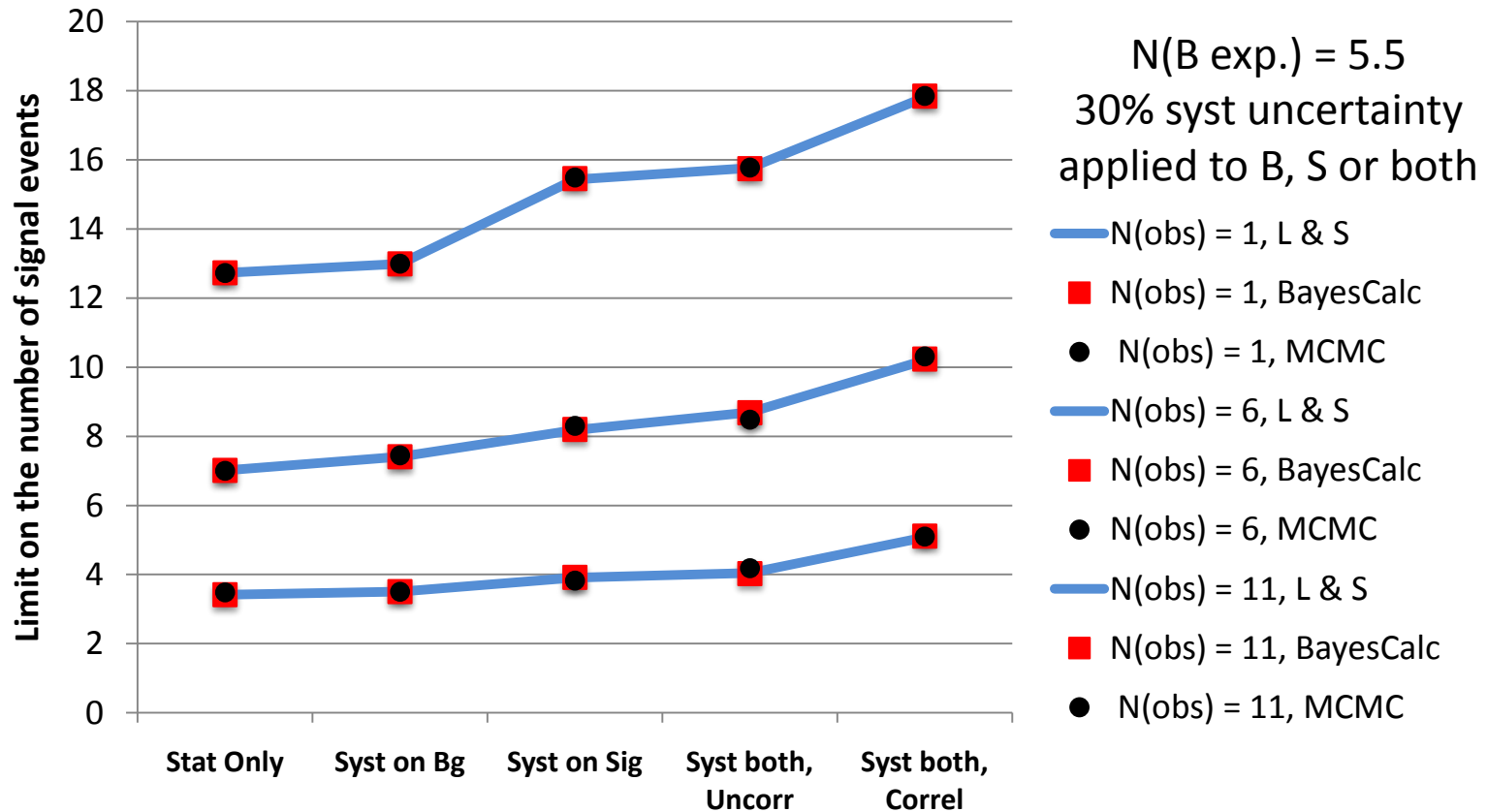


# Bayesian Methods: MCMC

- The MCMCCalculator is what we normally use for Bayesian limits for complex models.
- There are lots of configurable options; we're not experts of MCMC so our approach has been "try and see what seems to works best". **The choice of proposal function seems the more critical.**
- **Whenever MCMC works, the results are in good agreement with L&S for the two tested priors.**
- **One point also cross checked with BAT using the same Workspace. Good agreement found.**

# Bayesian

On simple experiments, good agreement between L&S, BayesianCalculator and MCMCCalculator

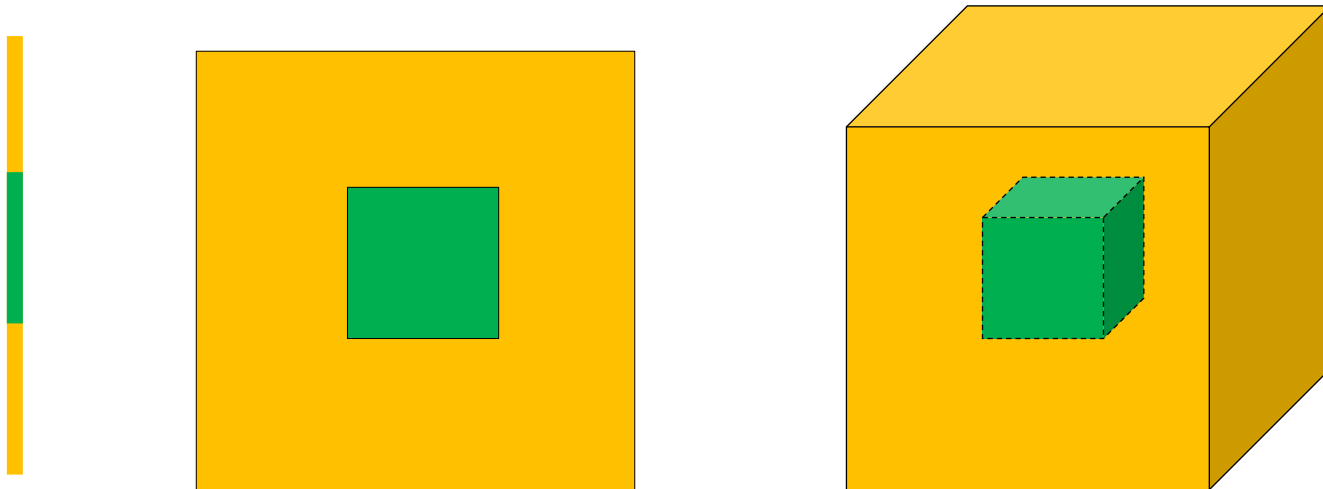


# MCMC Proposals

- Uniform proposal: works very well for very simple models, but fails in complex ones (the acceptance of the chain drops)
- Multi-gaussian proposal from ProposalHelper: our current default, it works fine but for large number of nuisances ( $>20$ ) the acceptance of the chain starts to drop, affecting the performance.
- We didn't test the proposal that starts from a fit covariance matrix; in past tests it worked like the Multi-gaussian but sometimes gave biased results

# MCMC Proposals

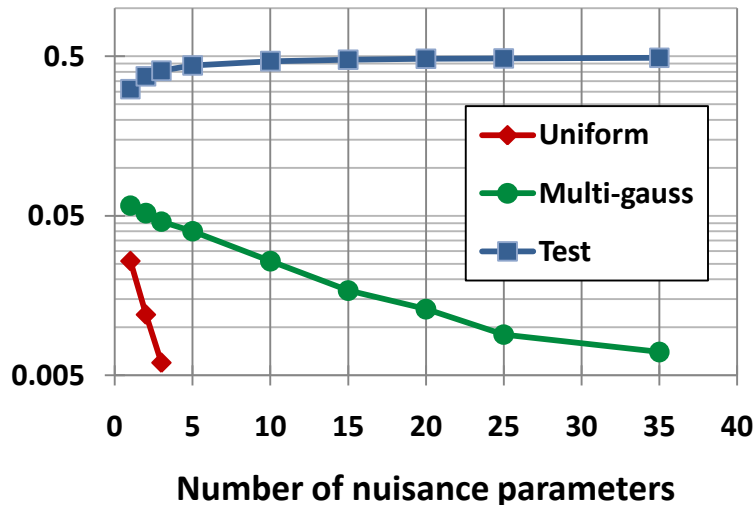
- The issue is that for large  $N$  most of the points of the  $[\pm n\sigma]^N$  hypercube have vanishing probabilities even for small  $n$  (4-5).



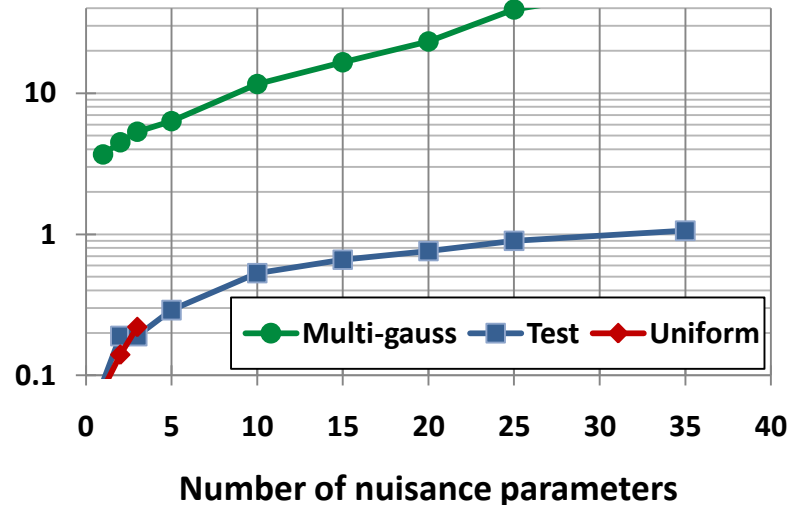
# MCMC Proposals

- Evaluated the performance of the various proposal with H->WW model but including only N nuisance parameters.

Acceptance of the chain



Time for 20x10k toys (minutes)



- Due to the higher correlation between points with the test proposal, the uncertainty increases with the number of nuisances (20x10k toys: 0.09%->1.4% for 1->35 nuisances)

# Hybrid

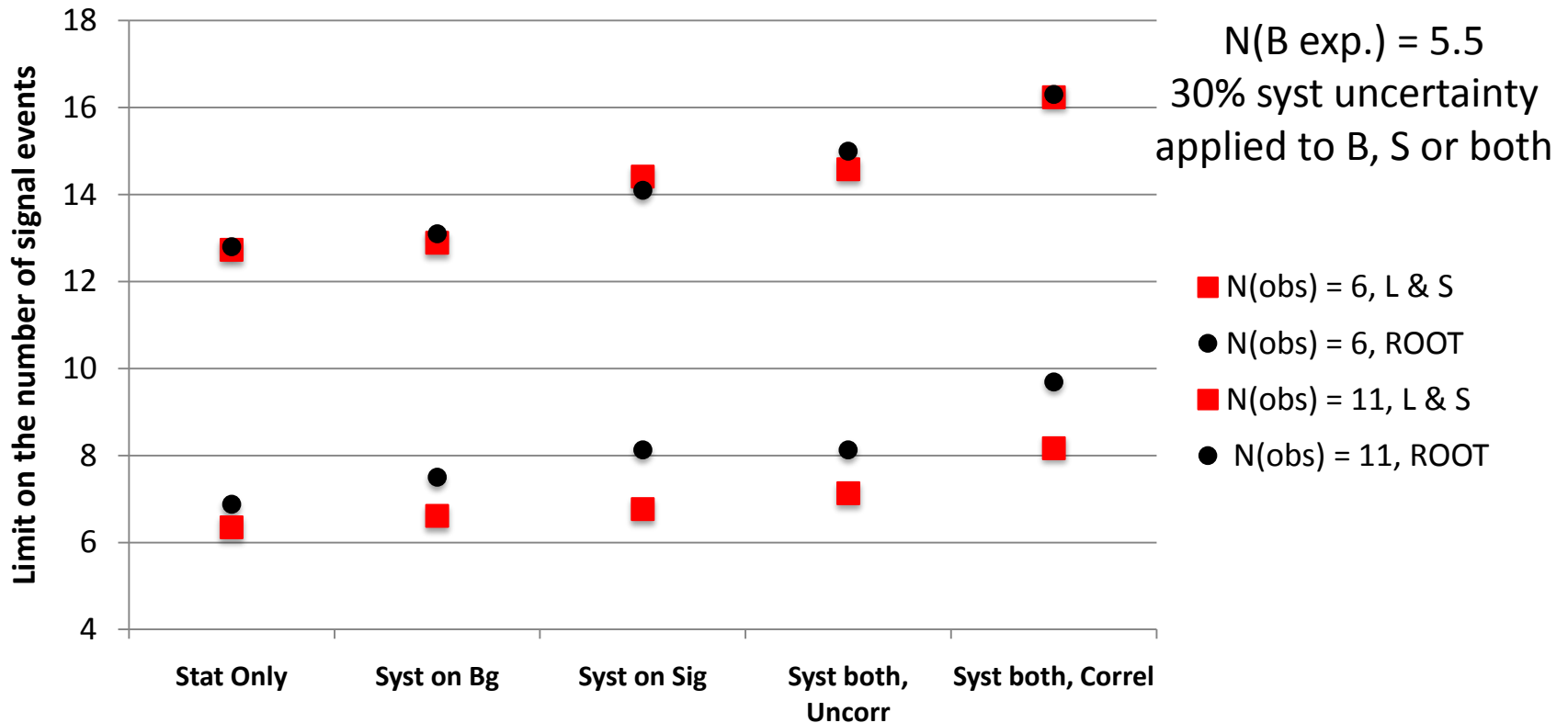
- **This code is very expensive in cpu and memory, limiting significantly the accuracy of what we can compute within “reasonable” bounds per job (e.g. 8hours & 8GB of RAM)**
- Multiple options were tested:
  - Two rules for limits:  $CL(s)$  vs  $CL(s+b)$
  - Two test statistics: LEP (no profiling), Tevatron (profile nuisances). We had some problems trying to implement Atlas’s one (profile all params.)

# Hybrid for limits

- Comparison carried on only with limited accuracy (order 10% for complex models)
- **For the LEP test stat., results appear to be in agreement with expectations**, except perhaps when  $n(\text{obs})$  is significantly below  $n(\text{exp})$ .
- **For the Tevatron test stat., found some issues.** Investigation ongoing to see if it's in our interface, but the poor performance make it painful (limits for simple benchmarks take several hours and sometimes exhaust the available memory)

# Hybrid limits

Comparison for simple counting experiments with different  $n(\text{obs})$  and syst. uncertainties.





# Hybrid for significance

- Again, performance is a problem but in this case the task can be parallelized.
  - **Successfully computed a 3.5 sigma significance for a H->WW model for the LEP test statistics**  
**Used 50 jobs of 10k toys, each taking 1.5h and 5GB.**  
**L&S on the same model does 1M toys in 2 minutes!**
  - Other test statistics tested only in simpler jobs.
- Here the importance sampling should make a huge difference, but we didn't have time to test it yet.

# Very first look at ROOT 5.28

- **Profile-likelihood & Bayesian:** exactly identical results for the two releases (for MCMC the same seed was used)
- **HybridCalculatorOriginal:** same as 5.27, memory usage still increases with time.
- **New HybridCalculator,** but not using the new 5.28 features like importance sampling. Performance qualitatively similar to old one, and similar memory problem.

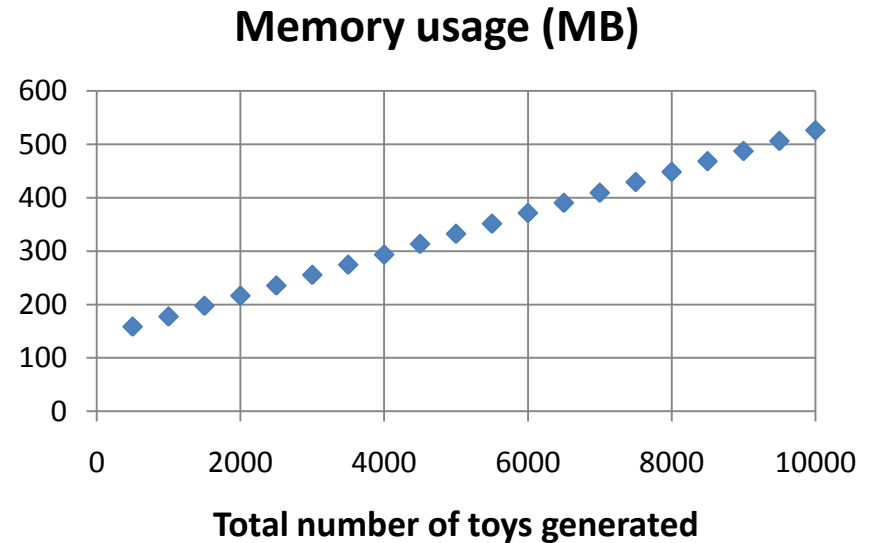
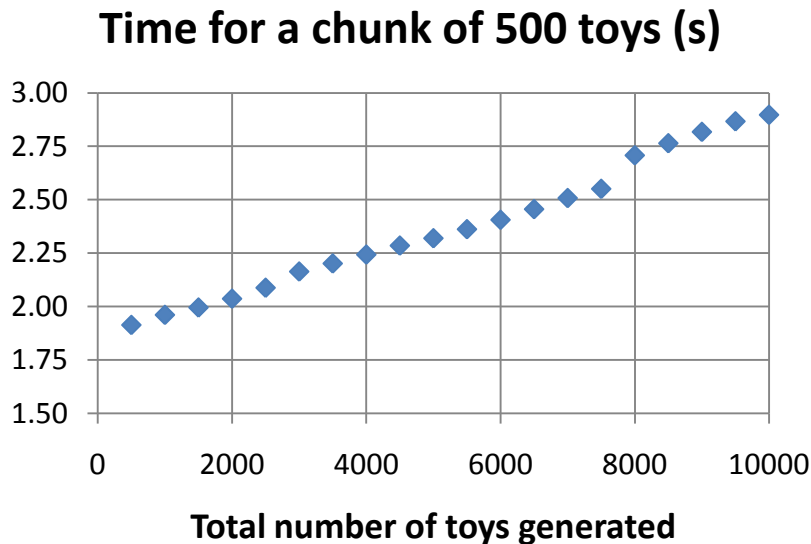
# Summary

- Successful validation of the profile likelihood and the bayesian methods in ROOT 5.27.06.
- For MCMC, testing a new proposal function to improve the acceptance in complex models.
- For Hybrid, only partial validation done due to the performance issues. Will need to work further for the test statistics with profiling.
- First checks on ROOT 5.28 gave similar results as ROOT 5.27, and similar performance issues, but we didn't test yet the new features.



# Getting Slower and Bigger

- HybridCalculatorOriginal toy generation becomes slower as the job goes on.



(simple counting experiment with a gaussian uncertainty on the background; LEP test stat.; **ROOT 5.28**)