

# Overview over the Capabilities (and missing pieces?) of the Fittino Code

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- 1 What Fittino is
- 2 What do we have?
  - Theory
  - Observables
  - Technologies
- 3 What's missing?

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# Aims of Fittino

- Fit the (N)MSSM parameters to the observables from all possible sources: LHC, ILC, Tevatron, B-factories, etc.
- Fit high-scale (mSUGRA, GMSB, AMSB) and low-scale models (MSSM, NMSSM)
- Bottom-up approach
- If the user wishes to test unbiased measurement: **no prior knowledge of the parameters at any step**
- Provide easy user interface for measurements, parameter definitions and output
- **Goals:**
  - Show that unambiguous parameter determination without human bias is possible
  - Determine precision of parameter measurements
  - Test the necessary **experimental** and **theoretical** precision
  - Study comparisons of models: MSSM vs. NMSSM etc.



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# Models

- mSUGRA, GMSB
- MSSM24
- MSSM24 with arbitrary other assumptions
- NMSSM
- **No** CMSSM, no models with partial unification (e.g. fit  $\tan \beta$ ,  $m_0$ ,  $A_0$ ,  $\mu$ ,  $M_1$ ,  $M_2$ ,  $M_3$ )  $\rightarrow$  need codes
- everything can be fitted with hand without an arbitrary number of SM parameters

# Calculators for Predictions

- Everything which works with SLHA
- Experience with  
SPheno, micromegas, “mastercode”, SuSpect, SoftSUSY
- Some requirements:
  - Code should not write any SLHA output in case of internal error
  - Code should, at best, always exit with return value  $\neq 0$  in case of internal error
  - For spectrum calculators: Please return every relevant particle mass, also the likes of  $m_W$  etc.
- can use different calculators in parallel for different observables, hard-coded...
- Missing: Select different calculators in the input file for each observable



# Observables and other Inputs

## Observables:

- Fittino can fit to any combination of the following observables:
  - Masses of SM and MSSM particles
  - Edges in mass spectra
  - Particle widths
  - Branching fractions
  - Cross-sections
  - Any product of cross-sections and/or branching fractions
  - Any ratio of the above
  - LE-Observables:  $b \rightarrow s\gamma$ ,  $(g - 2)_\mu$ , relic density,  $\Delta m_s/\Delta m_d$ , etc.
- Correlations among observables can be specified
- Limits on masses and BR of unobserved particles/decays can be specified
- HiggsBounds for model-independent limits on Higgs bosons from Tevatron and LEP





# Very Short Example Input File

```
#####
###          Fittino input file          ###
#####
massh0                115.237 GeV +- 0.05 GeV +- 0.5 GeV # comment
massTop                178.0   GeV +- 0.3   GeV
correlationCoefficient  massh0 massTop 0.05                # quatsch
# etc
edge 1 massNeutralino1 massNeutralino2      263.5 GeV +- 1.2 GeV alias 1
sigma ( ee -> Neutralino1 Neutralino2, 1000.,0.8,0.6 ) 7.678 fb +- 2.0 fb
BR ( h0 -> Bottom Bottom~ )    0.8033 +- 0.01 +-(lumiErr) 0.05
BR ( h0 -> Charm Charm~   )    0.05   +- 0.02 +-(lumiErr) 0.01
nofit cos2PhiL                0.62865 +- 0.0005
# etc
fitParameter TanBeta                10.0
fixParameter Mu                    358.6 GeV
universality MSelectronR MSmuR
# etc
LoopCorrections      on
CalcPullDist         off
```



# Technologies

- To find the  $\chi^2$  minimum
  - Tree-level estimate + MIGRAD in MINUIT:  
fast but unreliable, if not started very close to the true minimum
  - Tree-level estimate + Simulated Annealing:  
not too slow, very reliable detection of the global minimum
- To map the parameter space
  - 1D or 2D scans:  
fast, but no correct treatment of the correlations to fixed parameters
  - Markov Chain:  
n-dim. Probability map of the available parameter space
- To determine the uncertainties
  - MINOS in MINUIT:  
Slow and not very reliable
  - Automatic generation of pull fits:  
Very reliable and (at least on a farm) not slower than MINOS
- To get a feeling for the parameters and observables
  - Visualize the effect of variations of each parameter on observable  $\chi^2$ 's

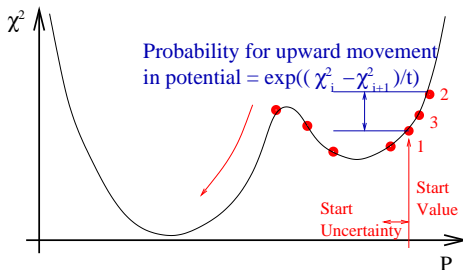


# Markov Chains

- Build a chain of parameter points  $\vec{x}_i$  using tMetropolis-Hastings
- Calculate a likelihood  $\mathcal{L}(\vec{x}_i)$  for each point. Here:  $\mathcal{L}(\vec{x}_i) = e^{-\chi^2}$
- Randomly pick a new point  $\vec{x}_{i+1}$  near  $\vec{x}_i$ , using a proposal distribution  $Q(\vec{x}_{i+1}; \vec{x}_i)$
- Calculate  $\rho = \mathcal{L}(\vec{x}_{i+1})/\mathcal{L}(\vec{x}_i)$  for symmetrical  $Q(\vec{x}_{i+1}; \vec{x}_i)$
- If  $\rho > 1$ , accept the new point and add it to the end of the chain
- If  $\rho < 1$ , accept it with probability  $\rho$ , else, add  $\vec{x}_i$  again to the chain
- For the efficiency of the algorithm: Optimize  $Q(\vec{x}_{i+1}; \vec{x}_i)$ , e.g. gaussian distribution around  $\vec{x}_i$  with widths  $\vec{\sigma}_i$
- Bayesian: Point density is proportional to the probability distribution
- Frequentist: Look for minimal  $\chi^2$  and  $\Delta\chi^2 < 1, 4, 5.99, \dots$
- Extremely effective sampling: **needed number of steps scales with the number of dimensions  $D$  instead of  $n^D$**



# Simulated Annealing



- Works as Markov Chains, but with different potential (Boltzmann) and with reducing the **temperature** with time
- Aim is not sampling of the parameter space, but reliable convergence towards the global minimum
- Use this together with **Toy Fits**: If theory code is stable (not always the case), this is free of any assumptions and renders all (non-linear) correlations and all uncertainties (without assumptions on gaussianities or else)

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# What's missing?

- Observables:
  - Shapes. maybe not necessary at LHC, but who knows
  - A general framework to include asymmetries
  - Systematic Uncertainties: Other shapes than gaussian?
- Models:
  - CMSSM (no fundamental problem, easy to implement if available by spectrum calculator etc)
  - Complex MSSM (too far away?)
  - R-parity violating models
  - No fundamental development necessary for the above, just need/use theory codes
  - More problematic: SM alone  $\rightarrow$  no SLHA!
  - Everything beyond SUSY is difficult simply because of the lack of SLHA. Need an input framework capable of handling SLHA and other formats?
- Minimization:
  - Quite well equipped, but other developments would be great to test, mainly for potential speedup, such as
  - Genetic Algorithms, Swarms, ...

