# Searches for stop production 8 & 13 TeV

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FUNDAÇÃO PARA A CIÊNCIA E A TECNOLOGIA MINISTÉRIO DA CIÊNCIA, TECNOLOGIA E ENSINO SUPERIOR

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- Motivation for stop production
- "Landscape" for a Susy, stop search
- Innovations we (LIP-CMS) have brought to these searches
- Searches of stop in 2- & 3-body decays
  - > 8 TeV
- Searches of stop in 4-body decays
  - > 8 & 13 TeV
- Conclusions & perspectives



# **Motivation for stop production**

$$M_{\tilde{t}}^{2} = \begin{pmatrix} \tilde{M}_{Q}^{2} + M_{T}^{2} + M_{Z}^{2}(\frac{1}{2} - \frac{2}{3}\sin^{2}\theta_{W})\cos 2\beta & M_{T}(A_{T} + \mu\cot\beta) \\ M_{T}(A_{T} + \mu\cot\beta) & \tilde{M}_{U}^{2} + M_{T}^{2} + \frac{2}{3}M_{Z}^{2}\sin^{2}\theta_{W}\cos 2\beta \end{pmatrix}_{\text{Up squarks}}$$

Mass difference of quark superpartners proportional to  $M_Q = M_t$ : Strong mixing in the stops  $t_{1,2}$  sector ?

#### $\mathbf{t}_1$ might be the lightest squark



### "Landscape" for a Susy/stop search

Cover as much mass hypothesis & decay scenarios as possible: **Be as model-independent as possible** 





# **Innovations brought by LIP-CMS**

- MVA approach for 3<sup>rd</sup> generation Susy searches
- ➤ Variation of signal versus Δm: Main parameter describing kinematically different signals



- → One MVA tool per each  $\Delta m$  of signal: Best selection for each signal hypothesis
- Quantitative rationale for including input parameters for MVA tool: Maximization of FOM: Obtain most reduced & best performing set of input variables

$$FOM = \sqrt{2((S+B)ln(\frac{(S+B)\cdot(B+\sigma_{B}^{2})}{B^{2}+(S+B)\cdot\sigma_{B}^{2}}) - \frac{B^{2}}{\sigma_{B}^{2}}ln(1+\frac{\sigma_{B}^{2}\cdot S}{B\cdot(B+\sigma_{B}^{2})}))}$$

### 2- & 3-body decays: 8 TeV 1 lepton final state



CM?

### 2- & 3-body decays: 8 TeV 1 lepton final state



Pedrame Bargassa - LIP

#### **LIP-CMS manpower involved: 1 researcher + 1 postdoctoral fellow**

- Responsable of signal selection
- Collaboration: 6 people involved through 3 institutes
- **Editorial:** Editor of the 8 TeV legacy paper:
  - \* "Search for direct pair production of scalar top quarks in the singleand dilepton channels in proton-proton collisions at  $\sqrt{s}=8$  TeV" JHEP 07 (2016) 027

# **Motivation for 4-body decays of stop**



b



**LIP-CMS manpower involved: 1 researcher** collaborating with colleagues of Vienna

- Cut & Count approach
- Signal selection:
  - > Introduced new selection variables:  $p_T(j)$ ,  $H_T$ , b-tag,  $\eta(l)$
  - Optimization: Definition of signal regions
- \* "Search for supersymmetry in events with soft leptons, low jet multiplicity, and missing transverse energy in proton-proton collisions at  $\sqrt{s}=8$  TeV" Phys. Lett. B 759 (2016) 9-35

# 4-body decays: 13 TeV 1 lepton final state

- 8 different BDTs trained, 1 per each  $\Delta m$ . Same set of input variables
  - >  $p_{T}(l)$ , Q(l),  $\eta(l)$ ,
  - > MET, M<sub>T</sub>,
  - » N(jet), p<sub>T</sub>(jet1), H<sub>T</sub>,
- > N(b),  $p_T(b)$ ,  $\Delta R(l,b)$
- CSV(jet w high b-disc)



**Premiere:** 1<sup>st</sup> search to make a DD background prediction based on output of MVA

$$N_{\text{DDprompt}}^{\text{SR}}(X) = T_X \cdot N^{\text{CR}} =$$

# 4-body decays: 13 TeV 1 lepton final state



#### LIP-CMS manpower involved: 1 researcher + 1 postdoctoral fellow

- Entire MVA search
- Collaboration: 2 (LIP) + 4 Vienna
- **Editorial:** Editor of the draft paper:
  - \* "Search for top squarks decaying via four-body or chargino-mediated modes in the single-lepton final state at  $\sqrt{s} = 13$  TeV". Under CWR, to be sent to JHEP



- CMS-LIP had/has seminal contributions in searches of stop @ CMS
- > 2015-2018: Two published results, one coming
- Manpower @ LIP: 1 researcher +
  - > 2 postdoctoral fellows: L. Lloret, C. Da Cruz e Silva
  - 2 + 3 summer students in 2016 & 2017
- > Perspectives:
  - Search of stop in 4-body decays with MVA continues: Deep NN where we expect better results than with BDT
    - > D. de Bastos joined PhD programme through IDPASC
      - See poster
  - > Investing our MVA expertise in  $H \rightarrow \tau\tau$  measurement... but that is another story



### Back-up

# **BDT training:** Internal parameters

#### Ntrees = 400: MinNodeSize = 2.5%: MaxDepth = 3: VarTransform = D

- Ntrees: Number of trees in BDT
- > MinNodeSize: Percentage of  $N_{s}$  or  $N_{B}$  @ which splitting of data stops
- MaxDepth: Maximal number of cut levels the BDT can reach
  - 2 last parameters: Stopping conditions of the training
- Optimal use of a BDT: "VarTransform=D"

Diagonalize the n-Dimensional space of input variables before feeding them to the BDT, for BDT to cut on them linearly. i.e. rotates the basis of variables along axis where more 1D discrimination is more present

<u>Internal parameters</u>: Optimized with a base set of variables to have best performance while no overtraining





#### Correlation Matrix (background)

#### Correlation Matrix (signal)



### **Selection for** $\Delta m = 30$ **:** Input variables

#### SetBase1

- ► FOM(300,270)=6.34 @ BDT>0.36
- ► S=50.5
- ► B=17.0
  - Wjets1/2/3/4/5/6/7=0/2.3/3.9/3.8/ 3.6/0.5/0





BDT > x

# **Comparison with CC via** $\sigma^{exp}$ (UL)

#### **CC**

- Expected 50.0%: r < 0.5332</p>
- SetBase0
  - Expected 50.0%: r < 0.3135</p>
- SetBase1
  - Expected 50.0%: r < 0.2412</p>
- SetBase1 + CSV
  - Expected 50.0%: r < 0.2197</p>

1/ The σ<sup>exp</sup>(UL) picture confirms the
FOM picture
2/ Even when taking the most reduced
set of variables (variables of CC
analysis = SetBase0), the BDT
performs better than CC
3/ The improvement across the 3 sets is
clear

