

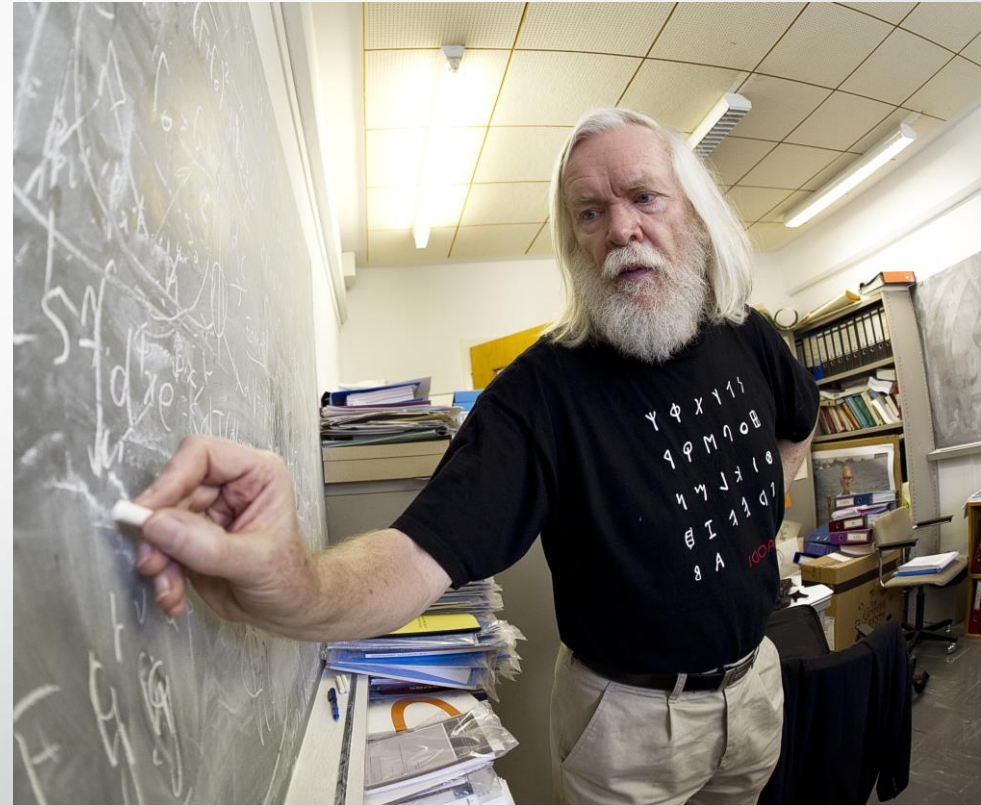
# Machine learning in Higgs physics

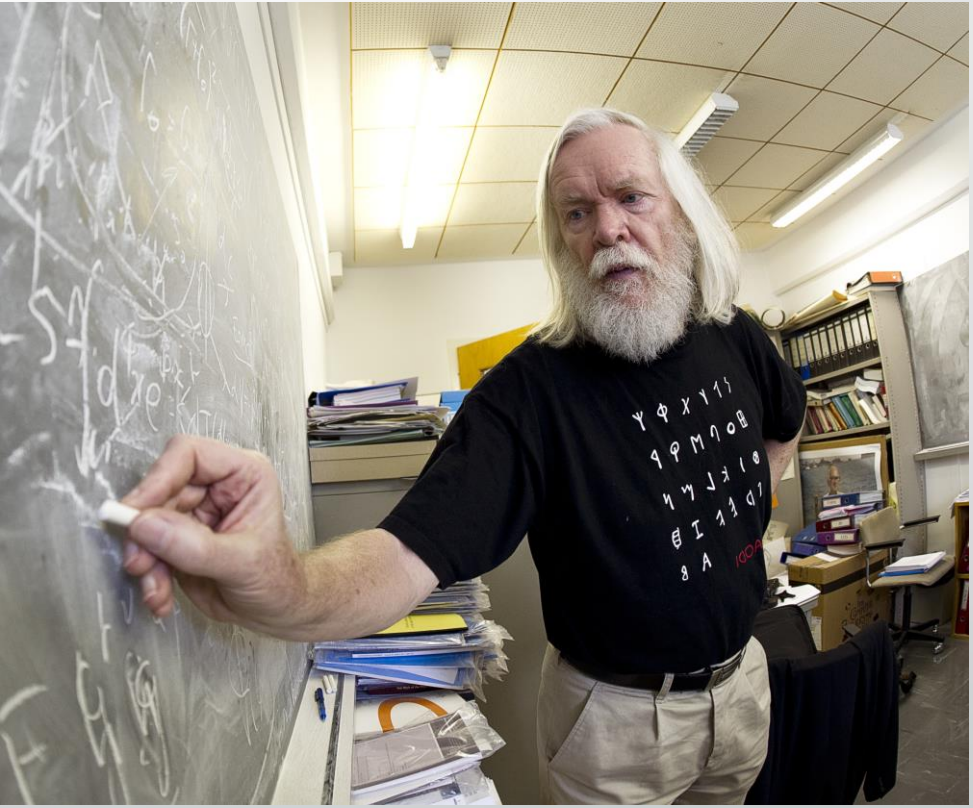
Giles Strong – LIP-Lisbon

Lisbon mini-school on Particle and Astro-particle Physics

08/02/17







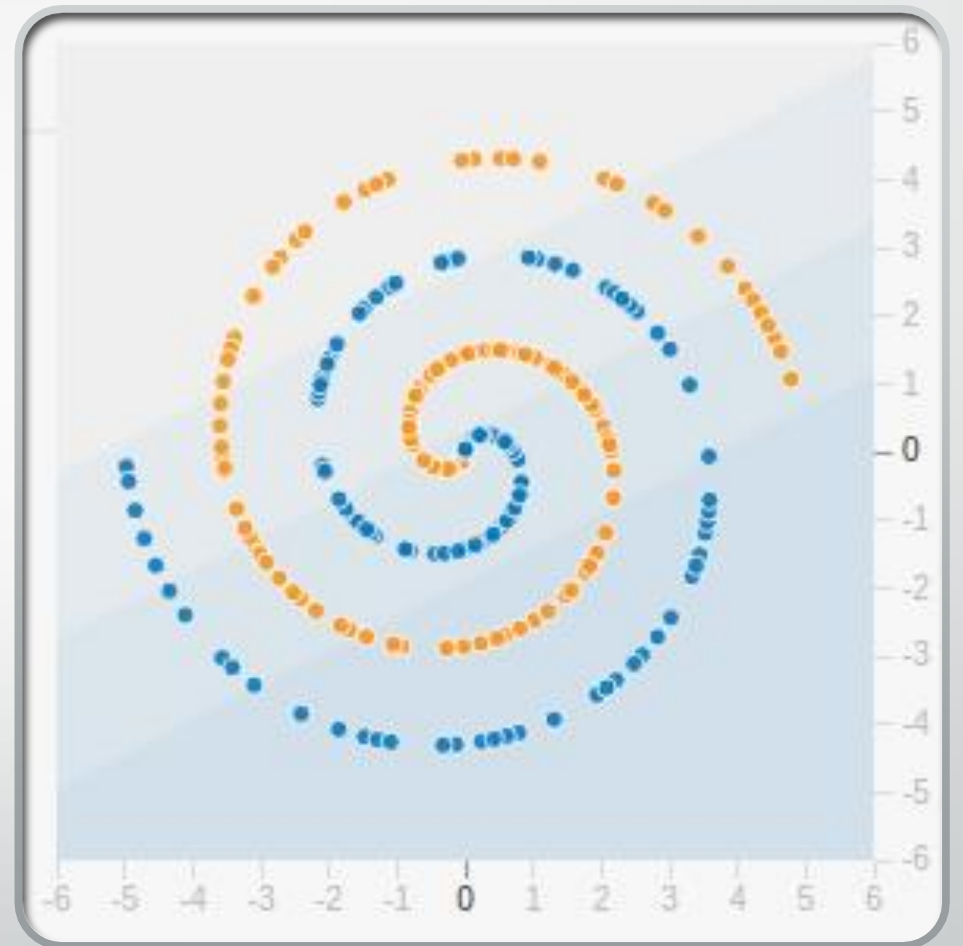
Machine Learning

# What is machine learning?

- In essence, it is the **automation of model building**
- By exposing a machine to a large amount of data, it can build up an understanding of the patterns within the data, and form predictive models
- Several methods exist, but let's examine artificial neural-networks

# Artificial neural- networks

Say we want to **predict the class** (orange or blue) of points according to their position





# Artificial neural-networks

FEATURES

Which properties do you want to feed in?

$X_1$

$X_2$

$X_1^2$

$X_2^2$

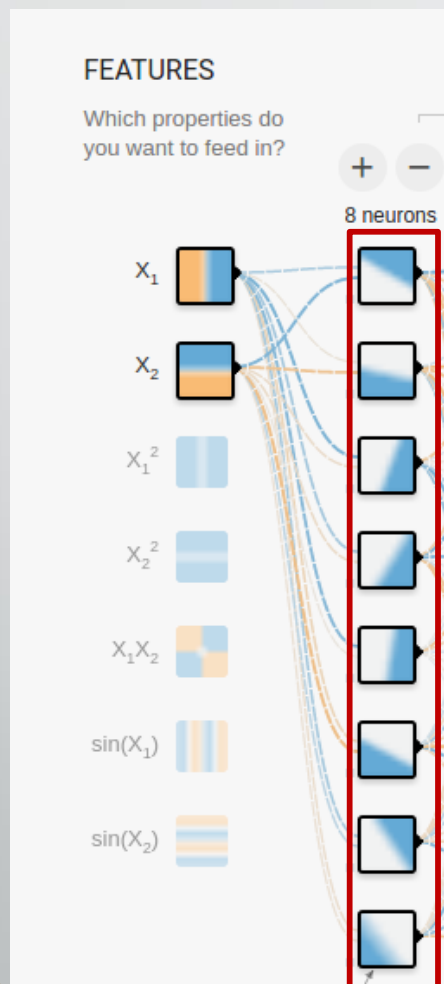
$X_1 X_2$

$\sin(X_1)$

$\sin(X_2)$

2 input features:  
X and Y coordinates

# Artificial neural-networks

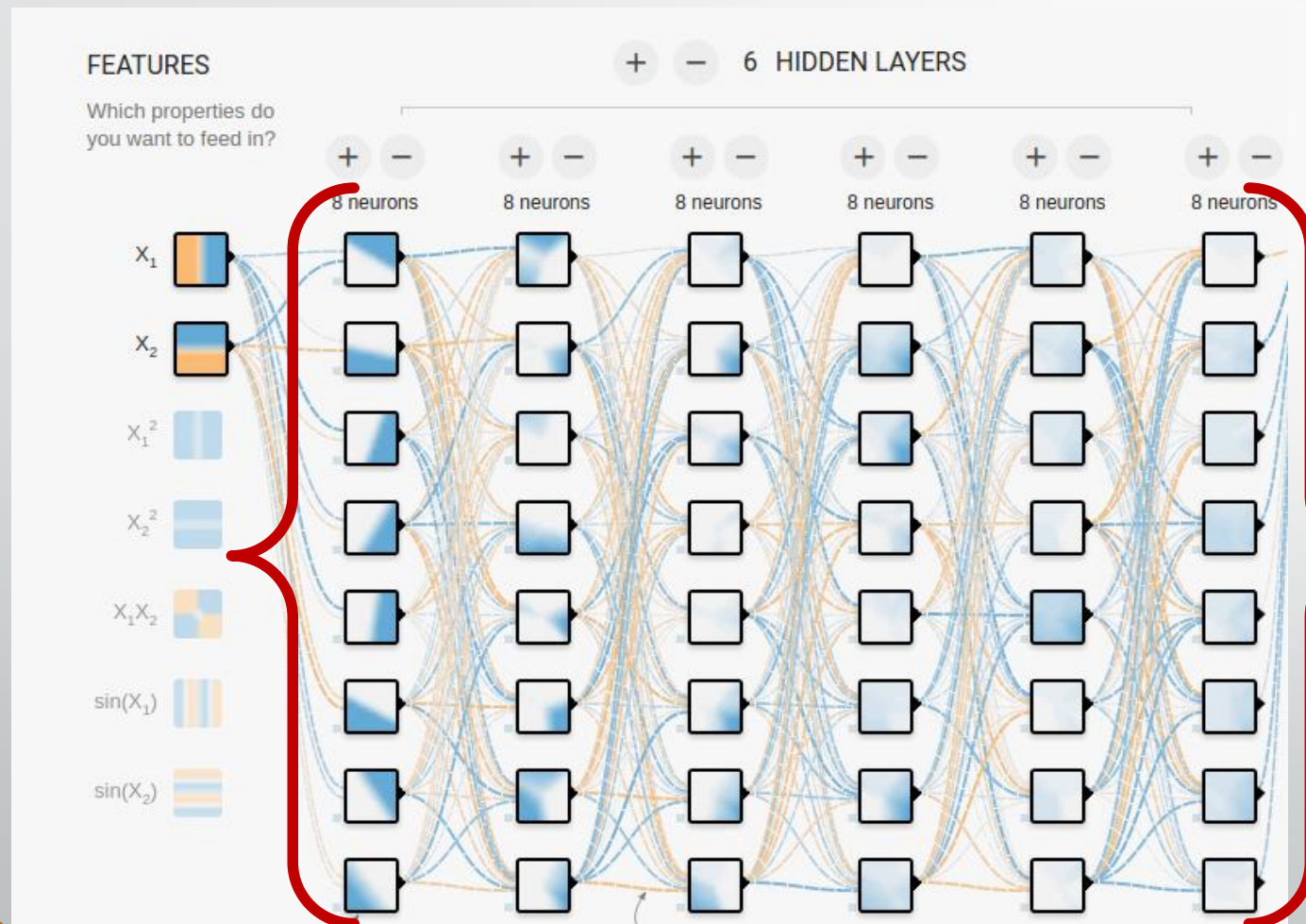


Neuron – applies mathematical transformation

Layer of neurons



# Artificial neural-networks



Add additional layers

# Artificial neural-networks



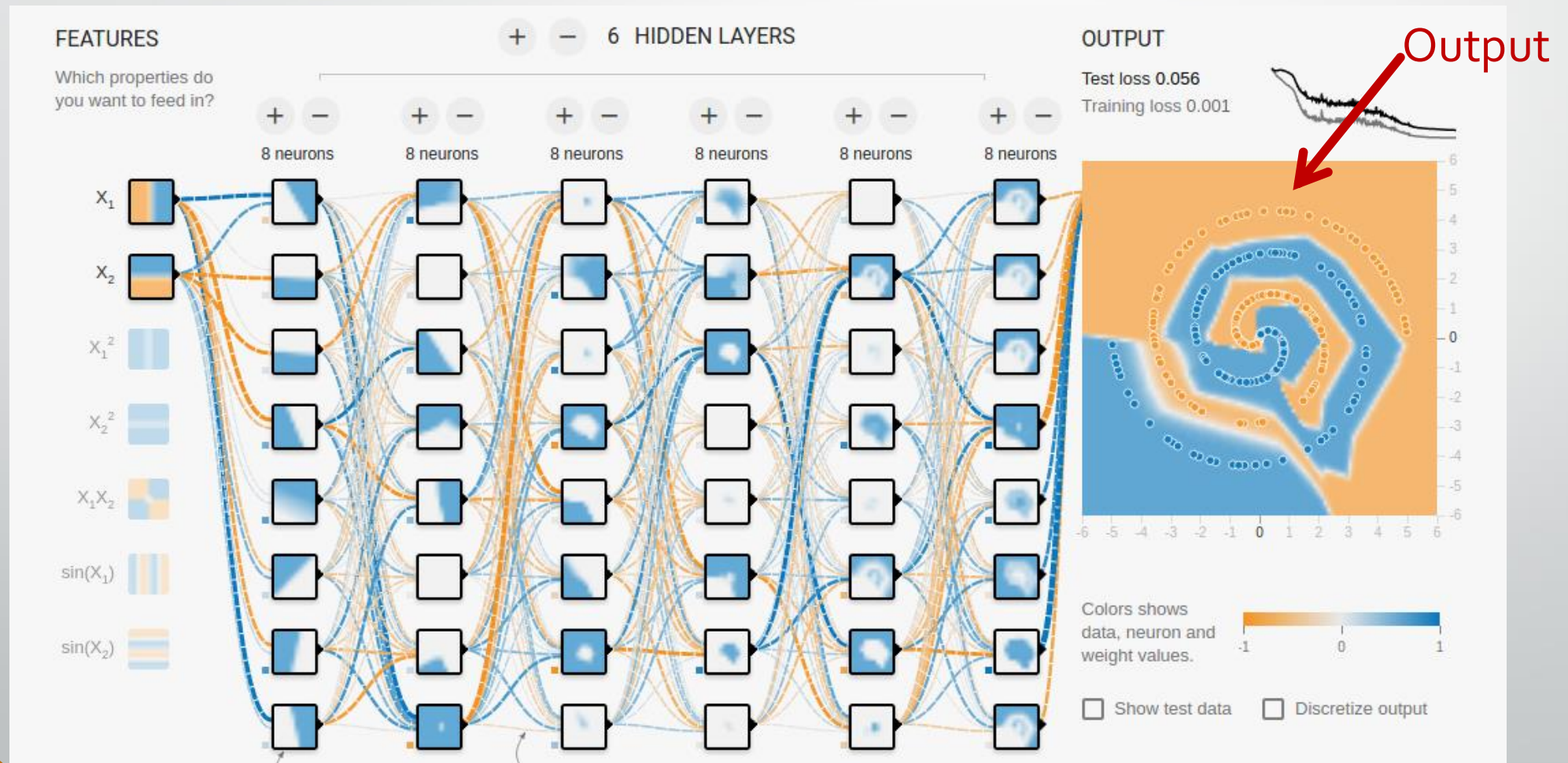


# Artificial neural-networks



Iterative training process

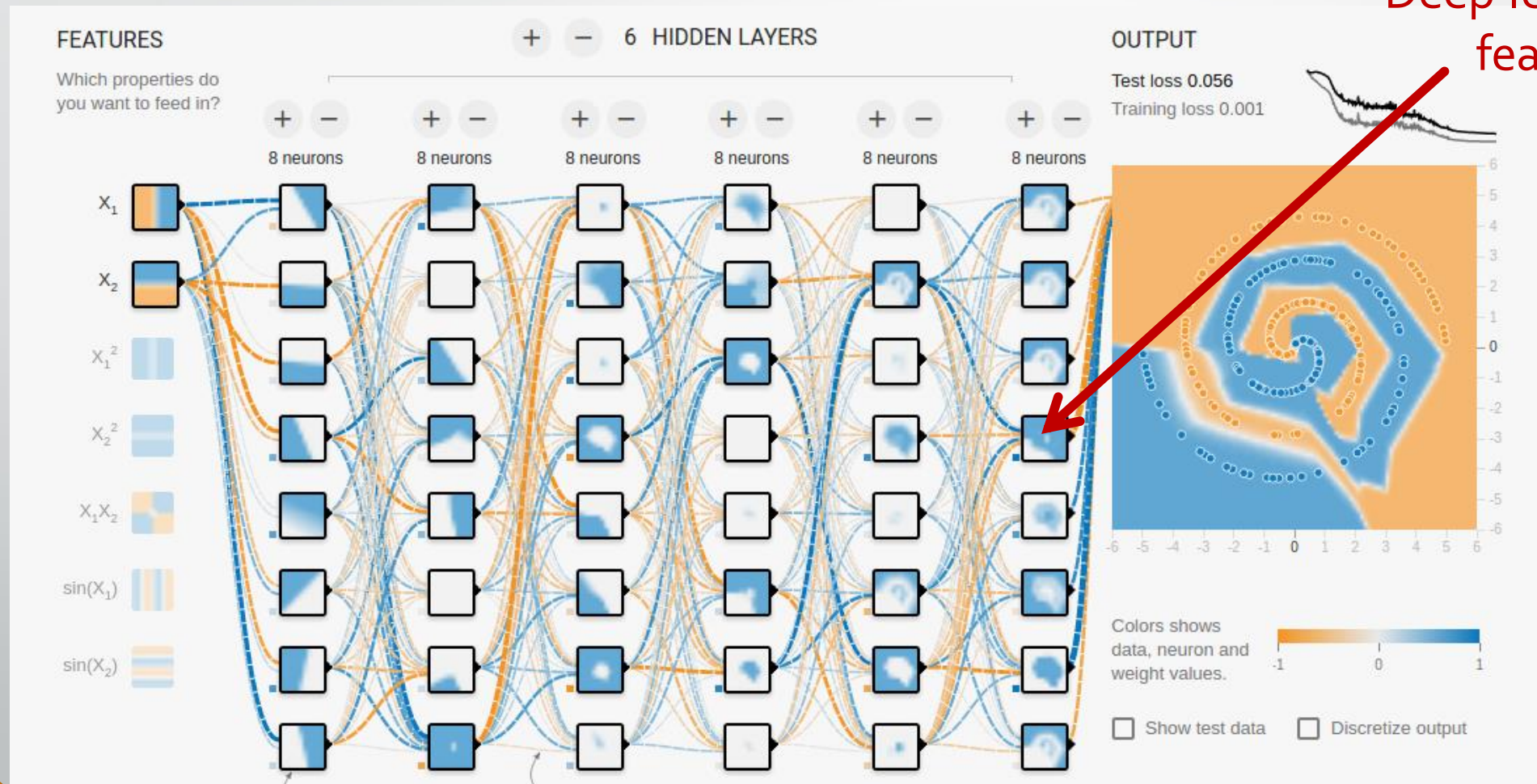
# Artificial neural-networks





# Artificial neural-networks

Deep learning of features



# Machine learning in particle physics

- We just classified points by their positions
- Imagine if instead the inputs were observables you could measure in a particle detector like CMS

# Machine learning in particle physics

- We just classified points by their positions
- Imagine if instead the inputs were observables you could measure in a particle detector like CMS
- Now you could **classify** collisions according to the particle process which occurred
- Neural-networks can also be used for **regression**; estimating numbers like particle masses from data



# Machine Learning in particle physics

Jet tagging

Regression

Event triggering

Many applications  
exist

Classification

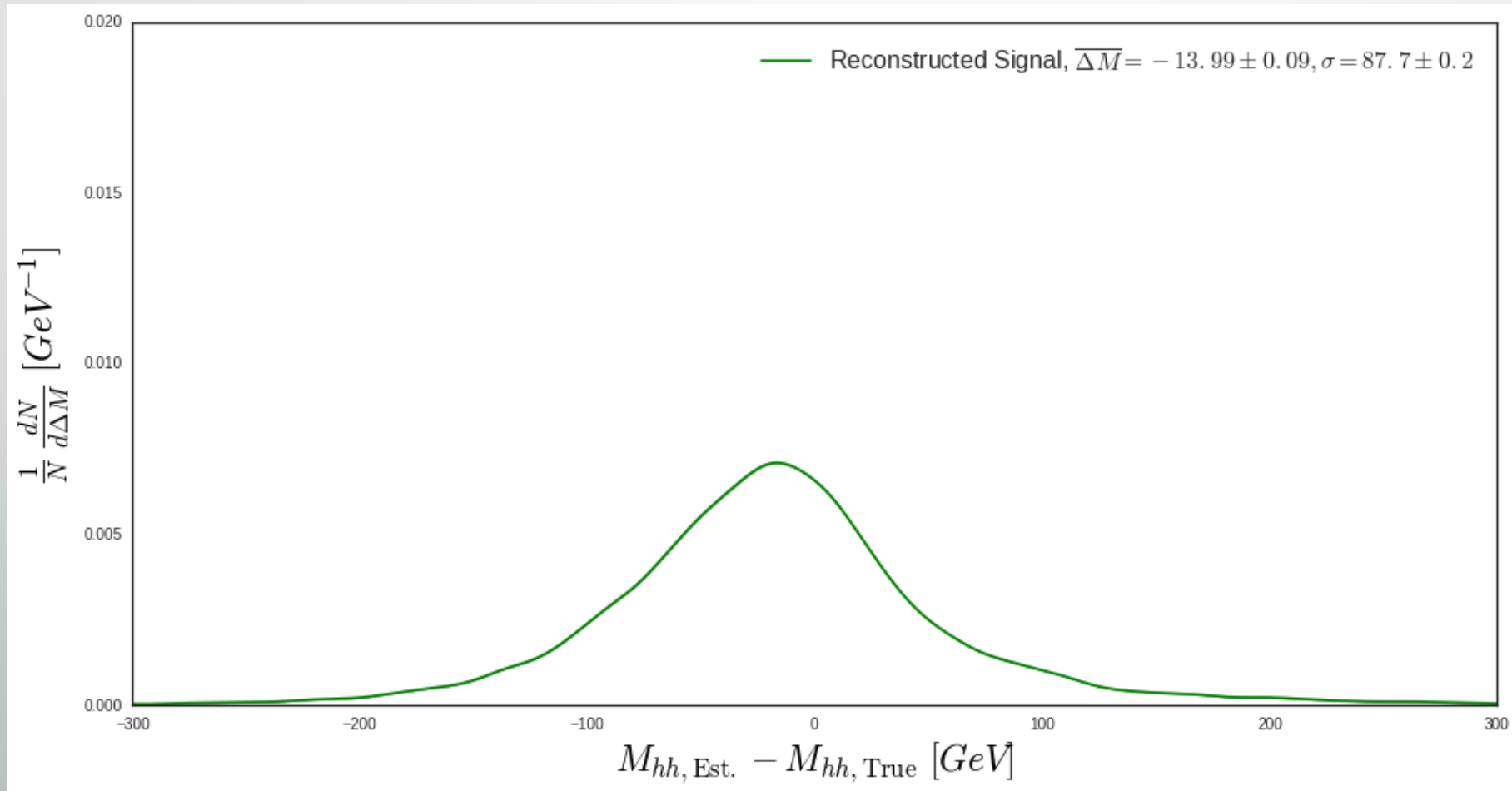
Simulation

Model evaluation

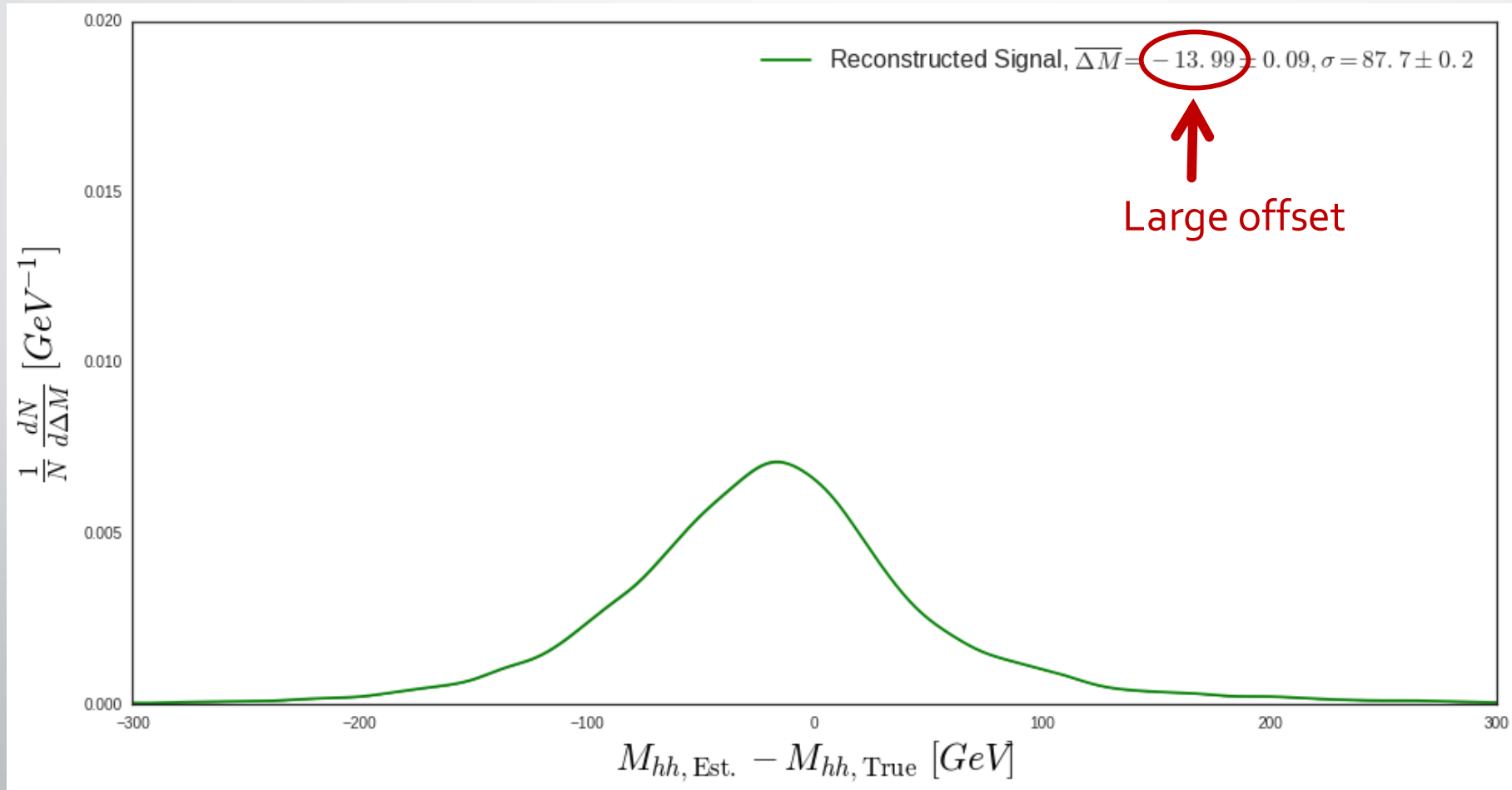
# ML in di-Higgs physics - an example

- Reconstructing the di-Higgs mass can be imprecise due to missing energy (neutrinos)
- By training a machine to understand the relationships between **observable features** and the **true values** of variables, the precision can be improved

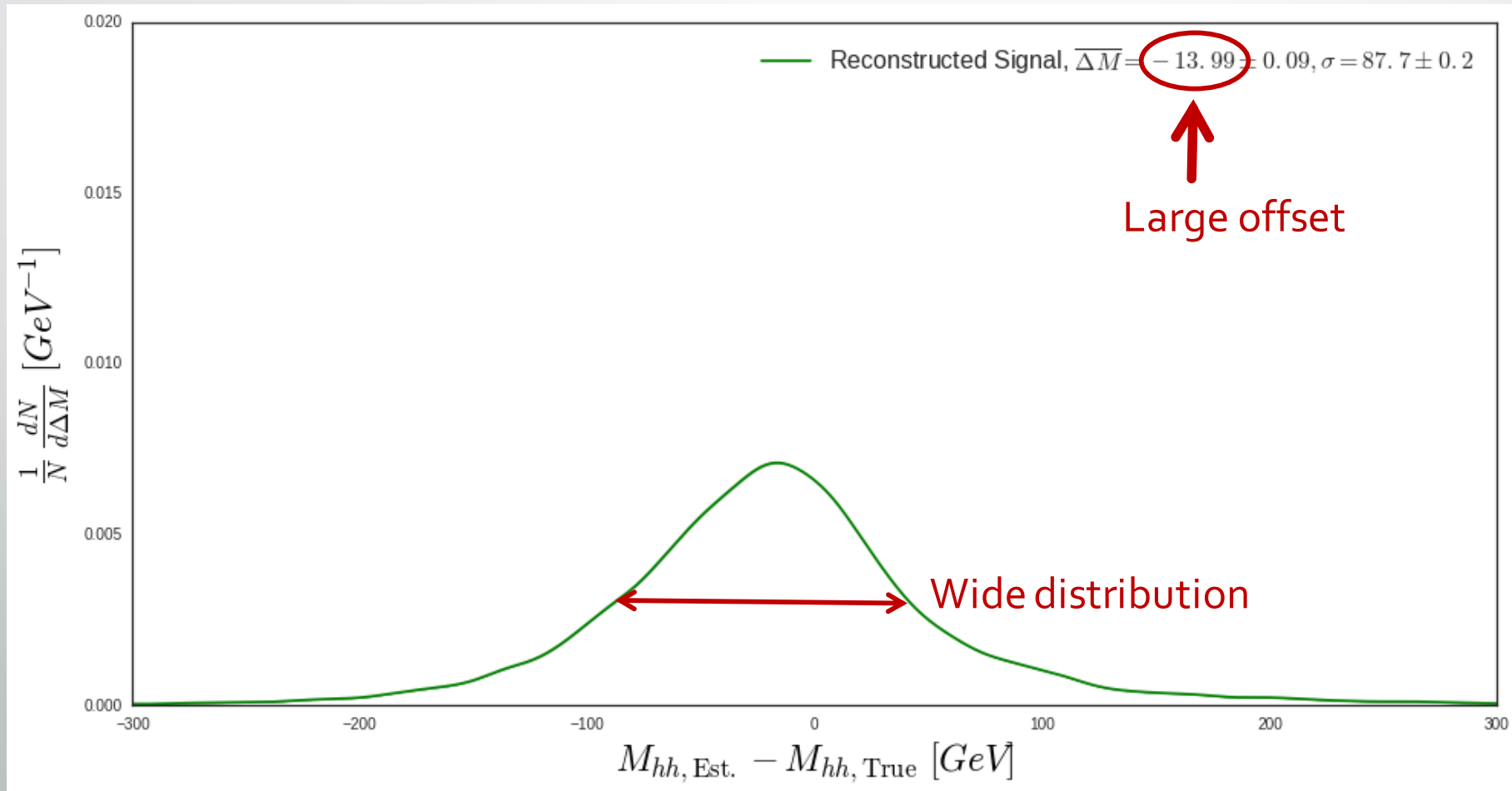
# Di-Higgs mass: reconstruction estimate



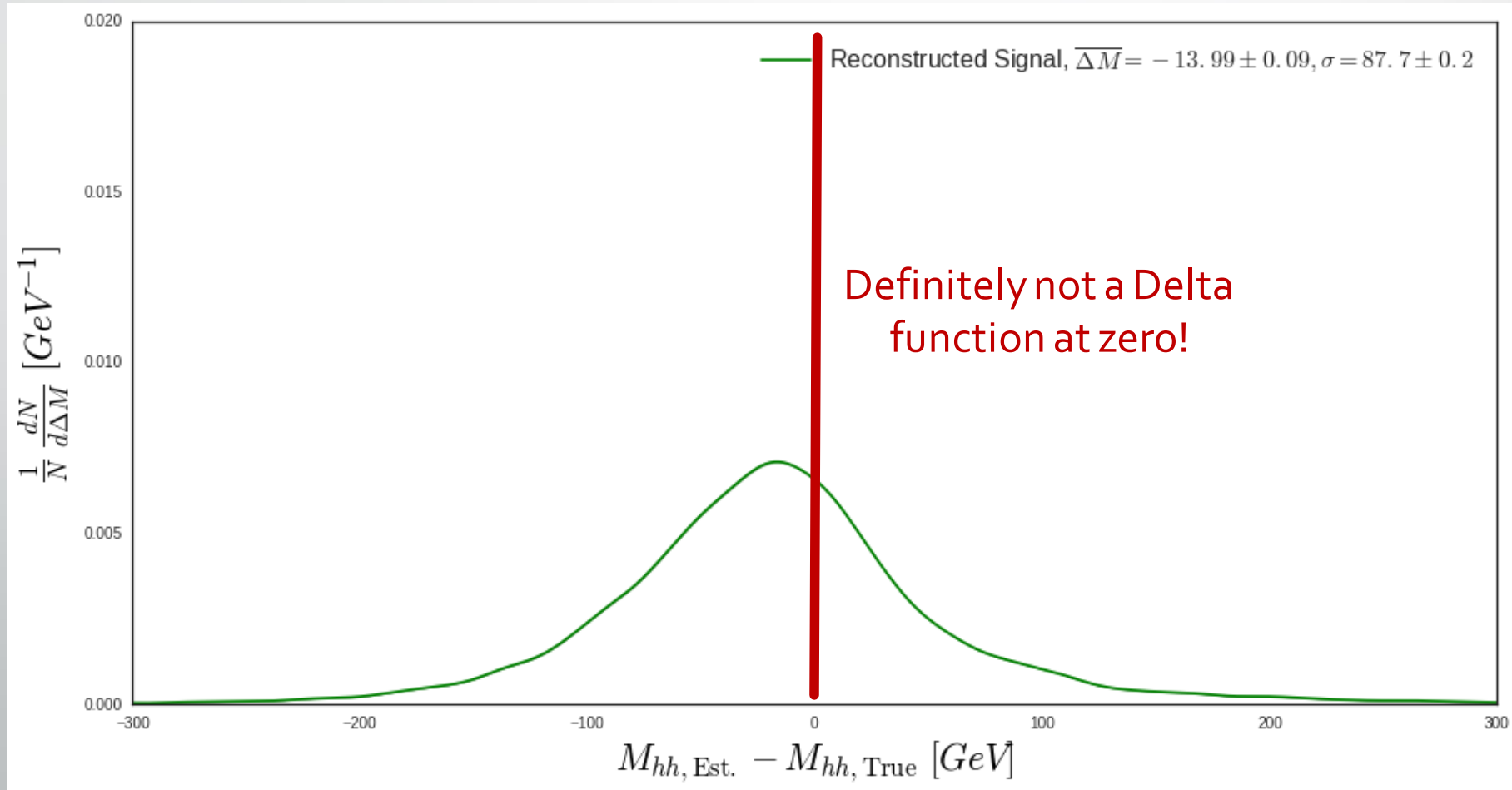
# Di-Higgs mass: reconstruction estimate



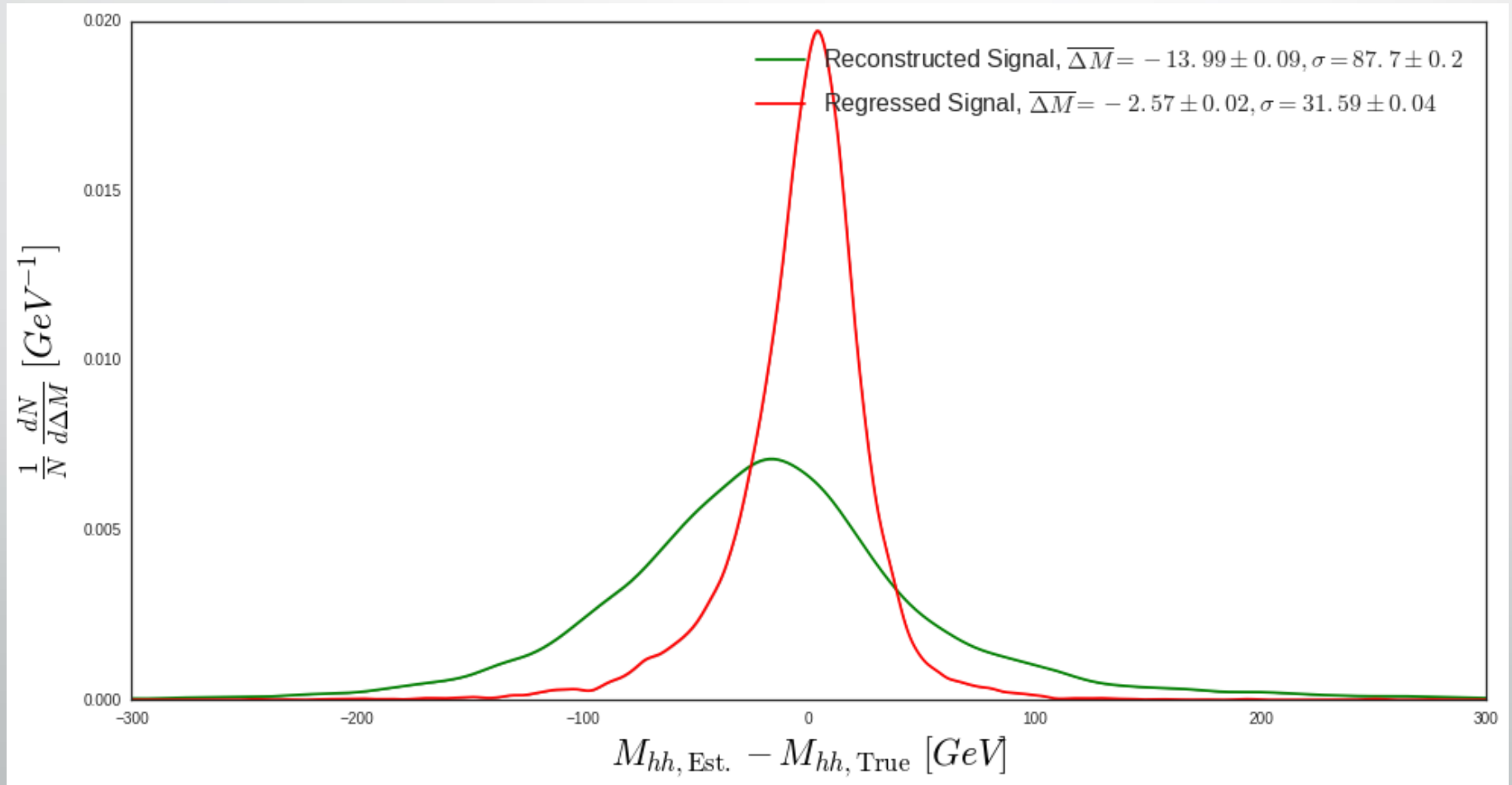
# Di-Higgs mass: reconstruction estimate



# Di-Higgs mass: reconstruction estimate

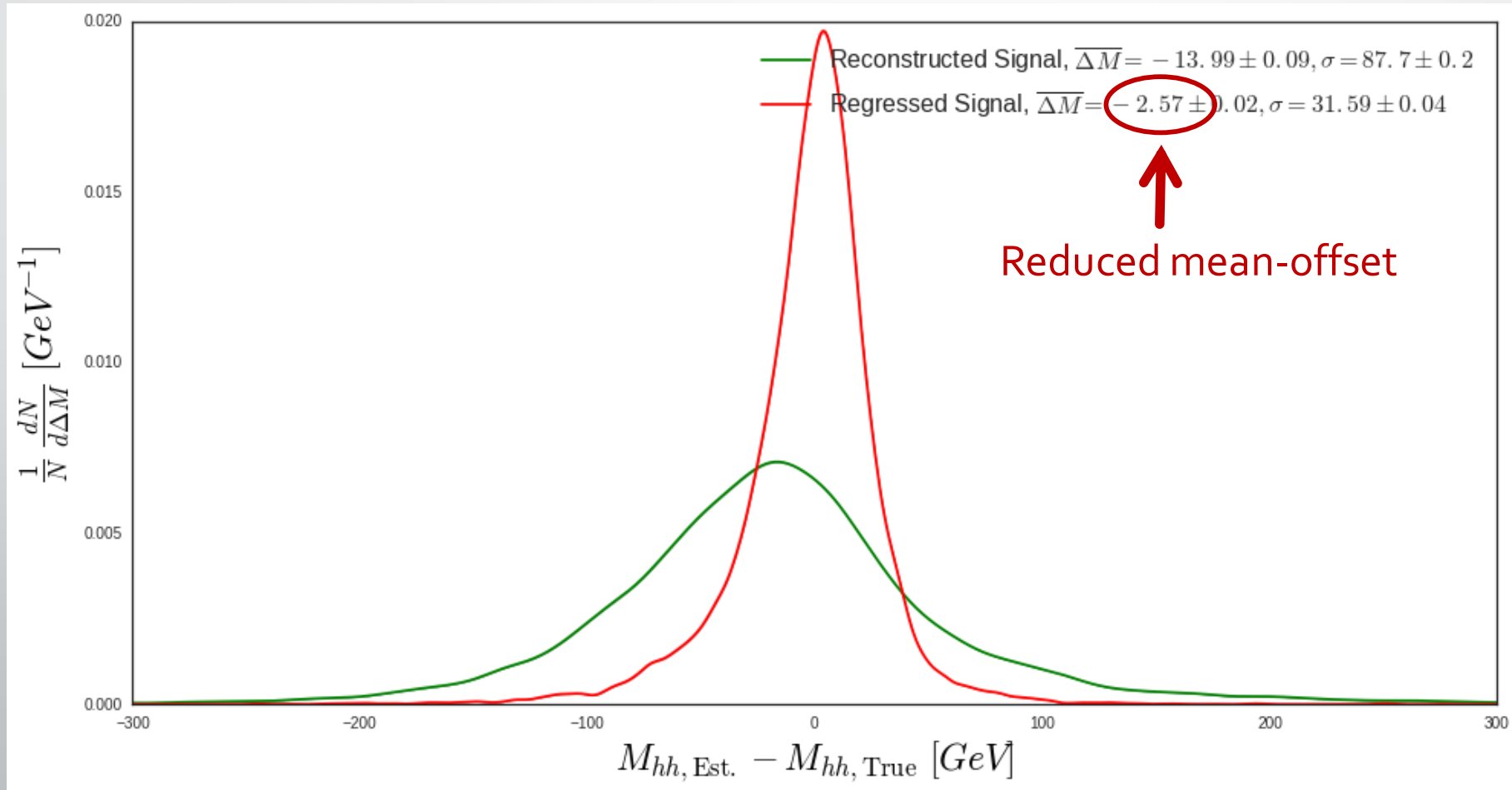


# Di-Higgs mass: regressed estimate

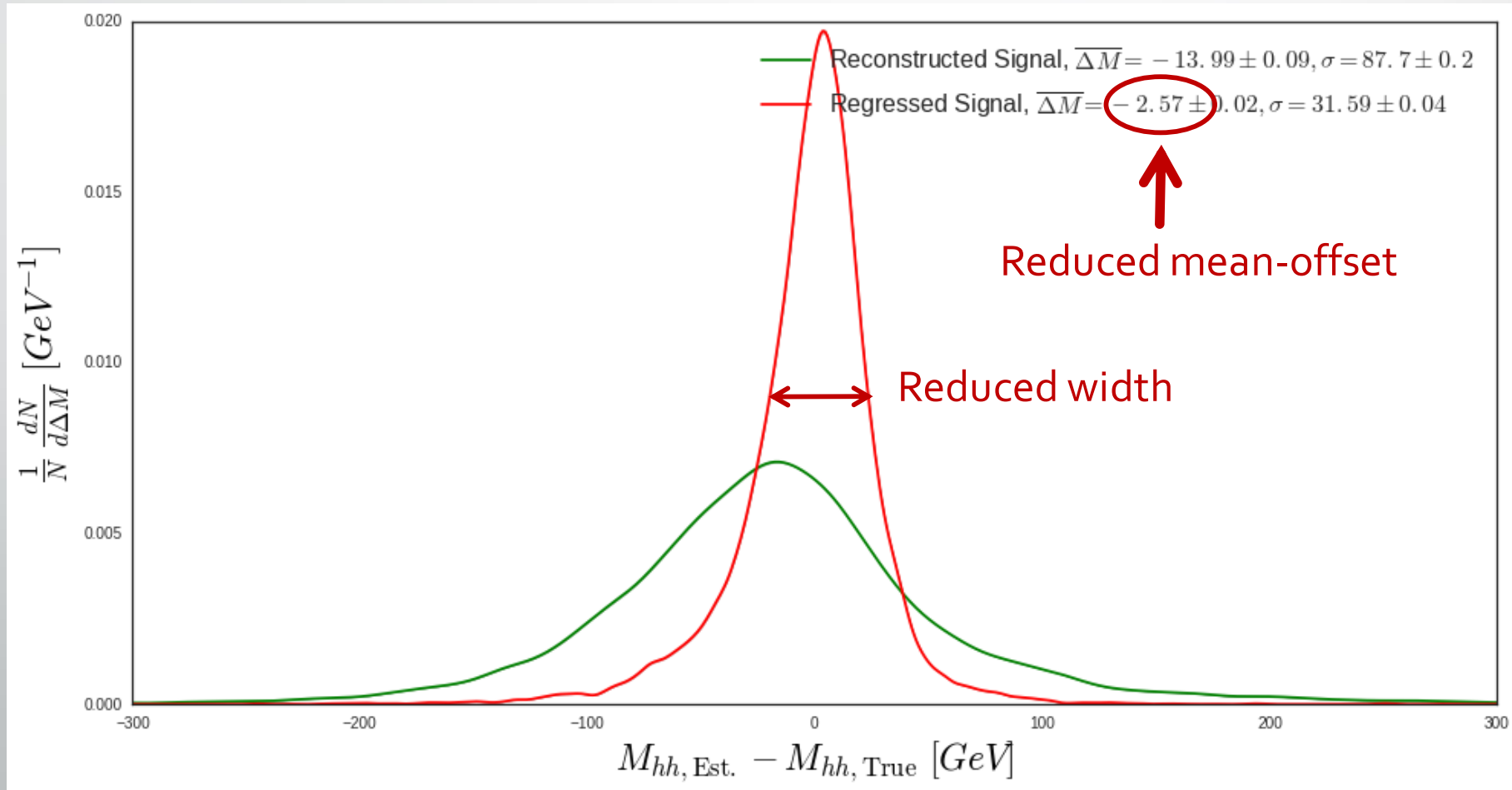




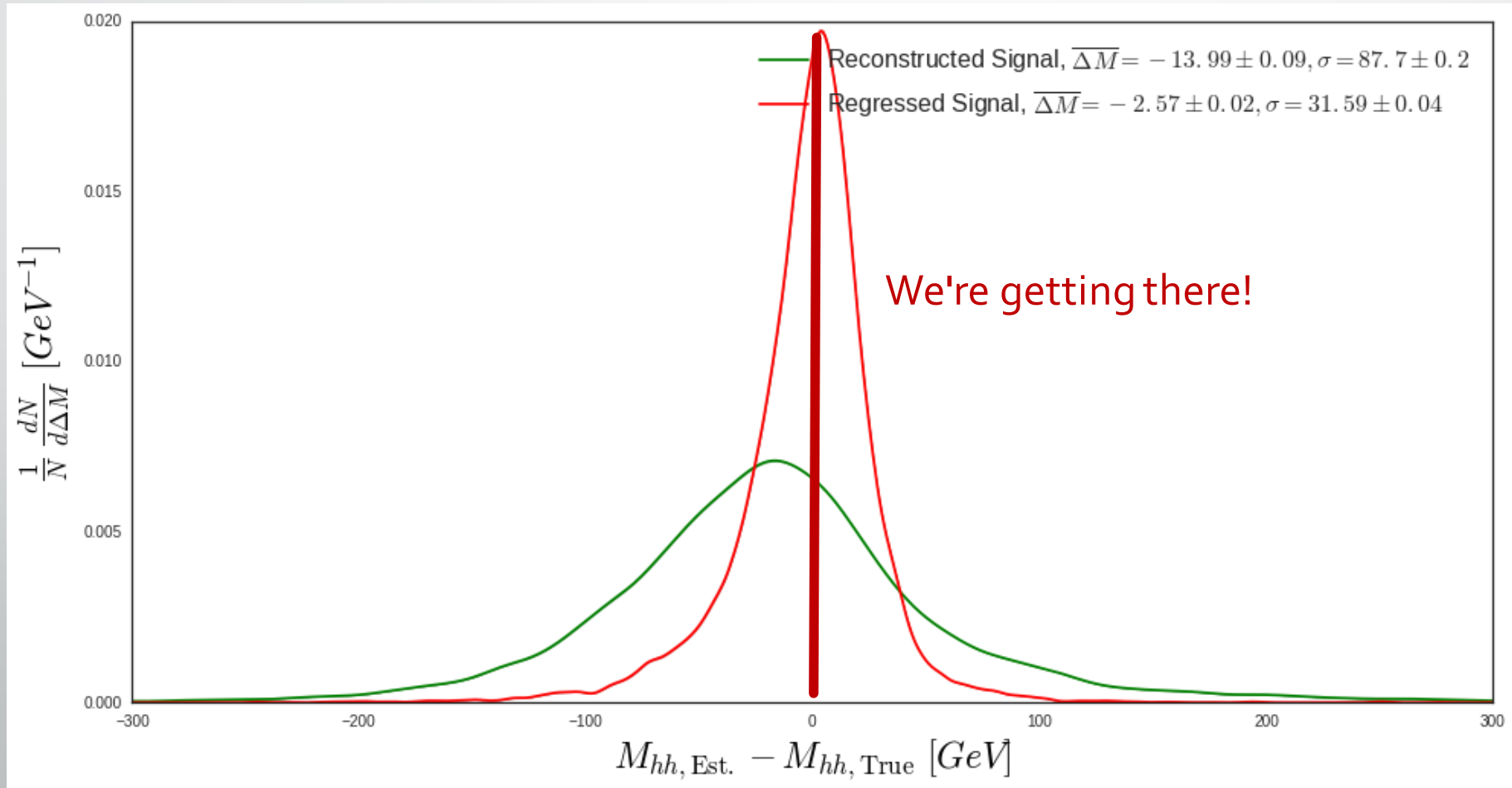
# Di-Higgs mass: regressed estimate



# Di-Higgs mass: regressed estimate



# Di-Higgs mass: regressed estimate





# Opportunities for *you* at LIP

- Master's theses
- Summer internships
- Lecture series

# Conclusion

- Several opportunities for you to get involved!
- ML is increasingly becoming the standard approach in particle physics
- It is also heavily used in industry and other areas of research
- Under heavy research itself

# Further reading – Searches terms

- Browser-based ML playgrounds:
  - [Tensorflow playground](#)
  - [Gradient boosting playground](#)
- Introductory course on ML: [mlhep2016 github](#)
- Network focussing on ML in physics: [amva4np](#)

# Further reading - Links

- Browser-based ML playgrounds:
  - <http://playground.tensorflow.org/>
  - [https://arogozhnikov.github.io/2016/07/05/gradient\\_boosting\\_playground.html](https://arogozhnikov.github.io/2016/07/05/gradient_boosting_playground.html)
- Introductory course on ML: <https://github.com/yandexdataschool/mlhep2016>
- Parameterised classifiers: <https://arxiv.org/abs/1601.07913> <https://arxiv.org/abs/1506.02169>
- Example investigation: <https://cds.cern.ch/record/2204934/files/HIG-16-028-pas.pdf>
- Network focussing on ML in physics: <https://amva4newphysics.wordpress.com/>