

Use of Machine Learning tools: **plans from IFCA**

COMCHA meeting
5th June 2018

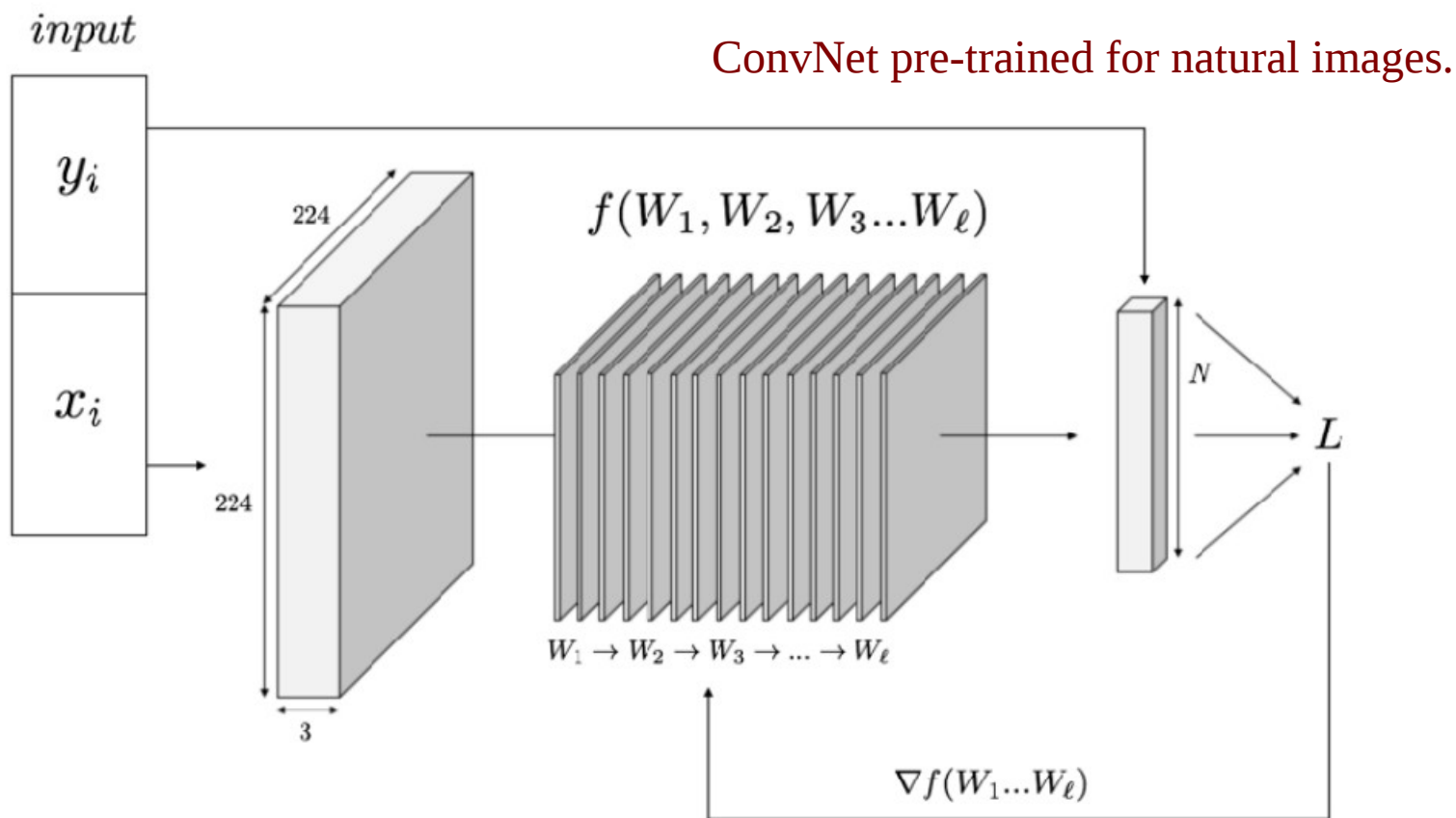
A. Calderón, C. Fernández, P. Fernández, I. Heredia, L. Lloret,
J.Marco de Lucas, P. Martínez, J. Piedra, M. Señas, N. Trevisani

Introduction and context

- IFCA has a strong group of advanced computing with focus on ML techniques.
 - About 10 people including physicists and software engineers.
 - This has to be added up to the High Energy Physics group.
- Administration of the IFCA Tier2 + HEP Tier3 and other resources:
 - Altamira super computer.
 - Currently 10 new GPUs + 1 old GPU.
 - Cloud resources.
- Also strong involvement in the Open Data group within CMS.
- Several years of expertise in Deep Learning techniques mainly for image classification.
 - Biodiversity. (Mobile application). **arXiv:1706.03736**.
 - High Energy Physics classification. **arXiv: 1708.07034**
 - Working now in medical diagnosis.
- Excellent environment to explore and promote ML tools in the context of HEP.

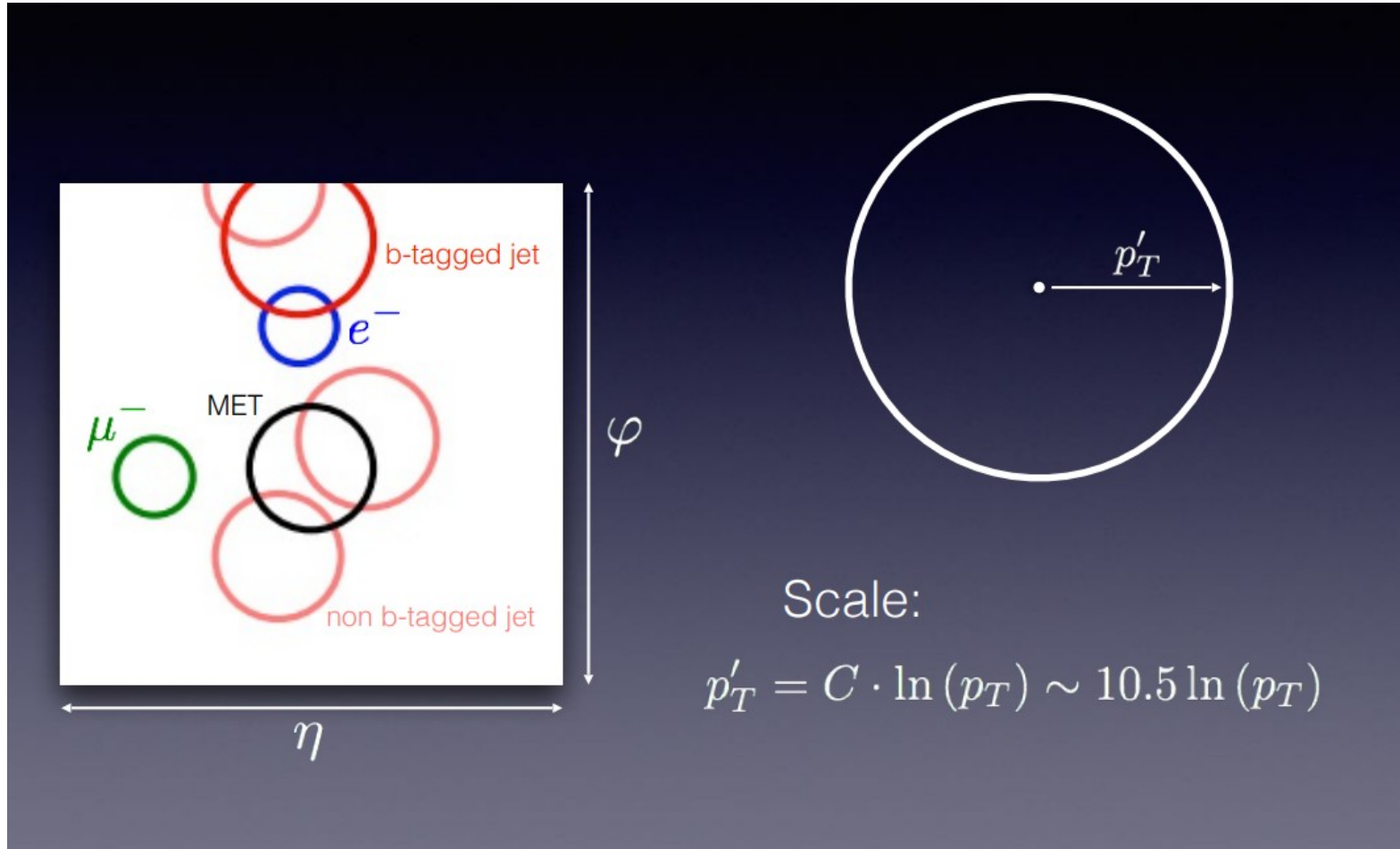
HEP example: event image classification

- Final degree project developed by Celia Fernández. Presented at the CMS ML forum.
- Use of a Convolutional Network to classify event images from different processes.



HEP example: event image classification

- HEP events are translated into pictures according to a simple geometrical mapping.

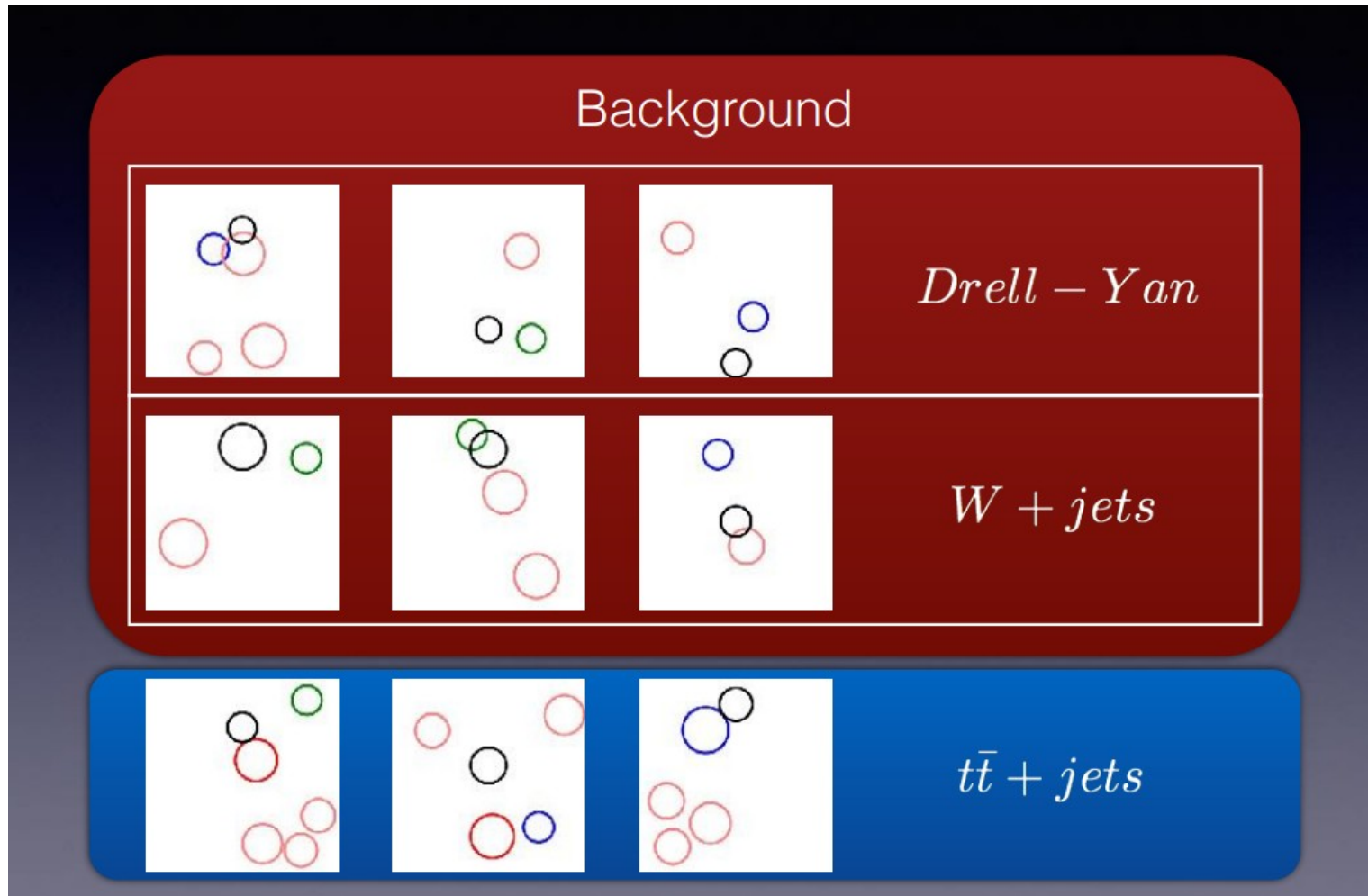


<https://indico.cern.ch/event/646782/>

arXiv: 1708.07034

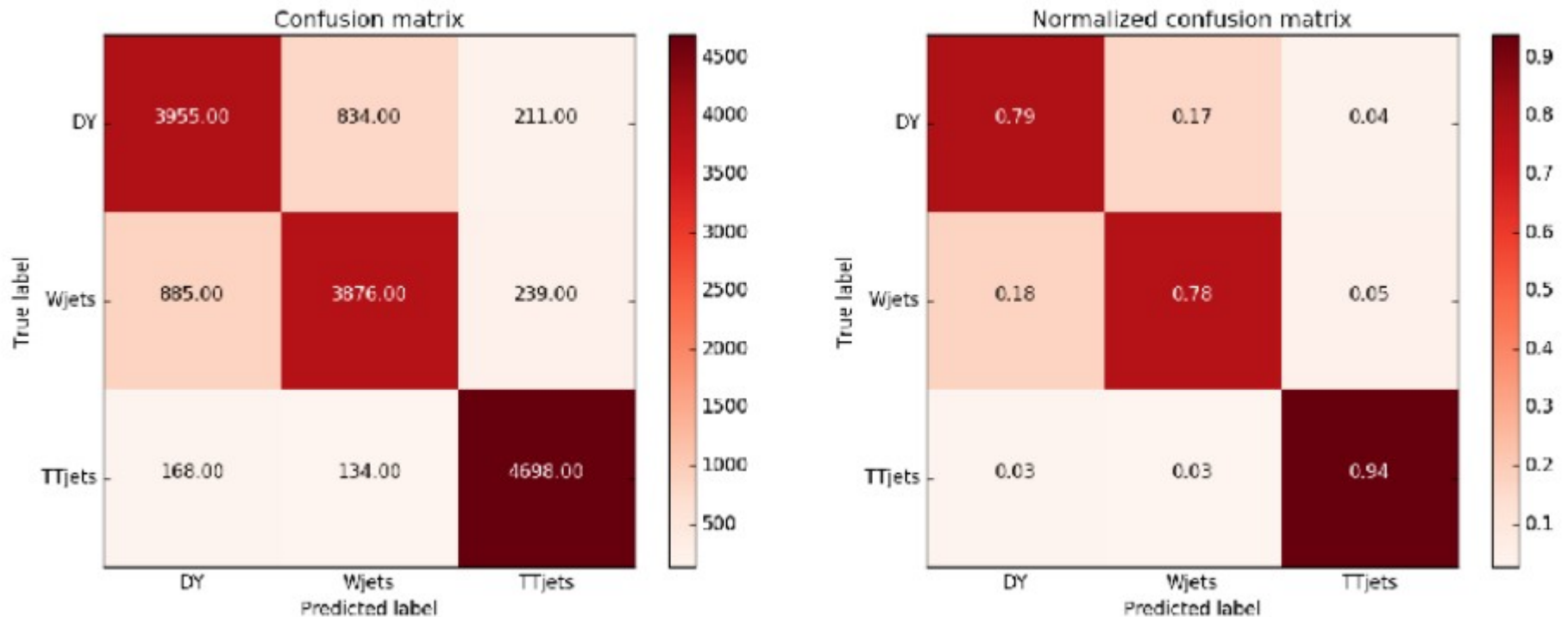
HEP example: event image classification

- Using these images to treat the background discrimination as an image classification.



HEP example: event image classification

- Using these images to treat the background discrimination as an image classification.



- Probability of choosing the right category: 84% (98% within the two first choices)

HEP example: realistic MC generation

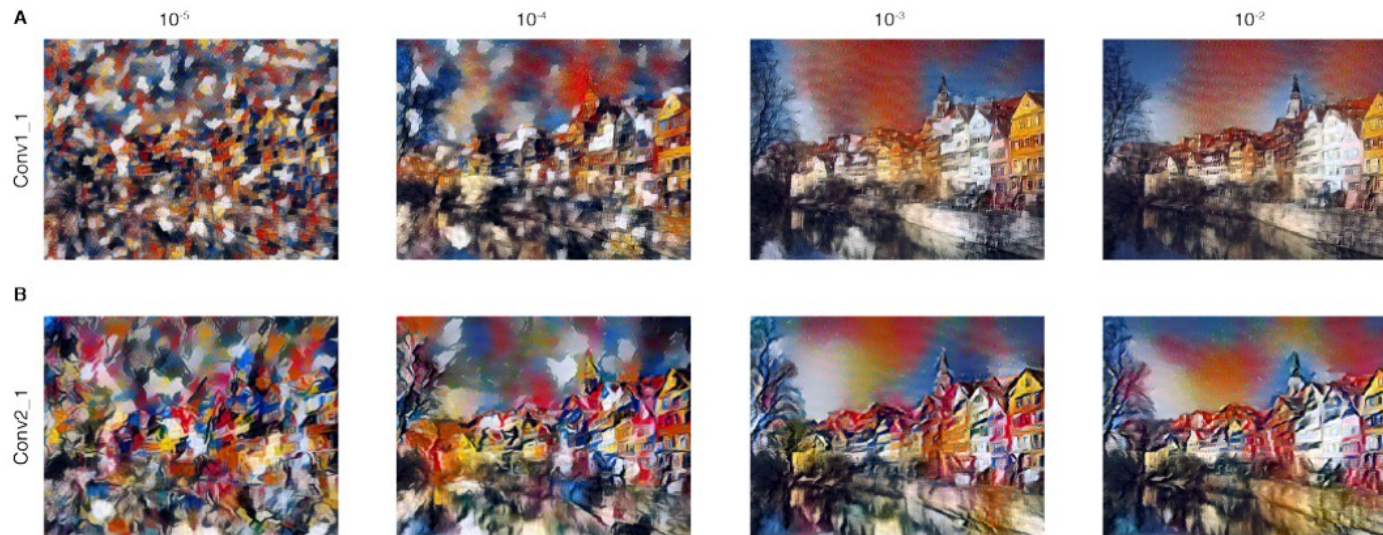
- This is currently a Final Degree Project (Mario Señas) to be presented in September.
- Learning local correlation patterns to produce realistic MonteCarlo simulations.



Based on arXiv:1508.06576

HEP example: realistic MC generation

- The “style” of an artist is encapsulated in the local correlations among pixels.
- The content produces larger-scale correlations among distant pixels.
- Convolutional networks learn local correlations in the first layers/filters.
- Idea is to consider MC simulation as a real image and data as an “artist/detector” vision.
 - Detector effects in real data introduce local correlations in the observables.
 - We train with data to learn these local features and then apply to real MC.



$$\mathcal{L}_{total}(\vec{p}, \vec{a}, \vec{x}) = \alpha \mathcal{L}_{content}(\vec{p}, \vec{x}) + \beta \mathcal{L}_{style}(\vec{a}, \vec{x})$$

Based on arXiv:1508.06576

HEP example: realistic MC generation

- In order to map processes into pixel maps where correlations make sense:
 - We make two categories of observables: simple and composed.
 - Both of them are used in a 2D table. Size of each pair map into a pixel colour.

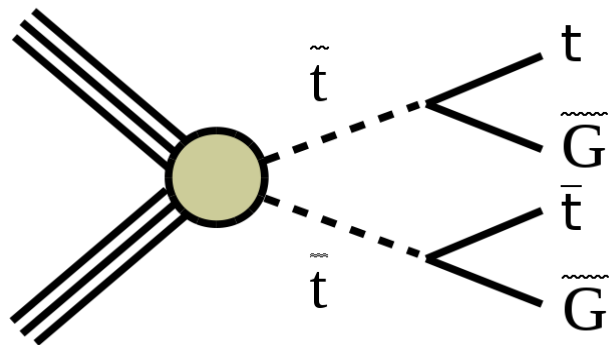
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Work ongoing...

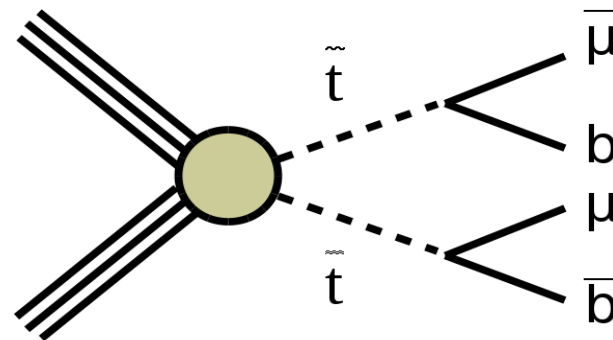
Based on arXiv:1508.06576

Plans for displaced vertex/track reconstruction

- › IFCA group has an increasing interest in searches including displaced tracks/leptons.
 - › Expertise both in SUSY and EXO searches in conventional prompt-signatures.
 - › Also expertise in b-tagging algorithms.
- › Starting now involvement within the new timing detectors for the CMS upgrade.
- › Already contributing in the context of the Yellow Report for the European Strategy.
- › Two PhD students (Celia Fernández, Pedro Fernández) with focus on this activity.



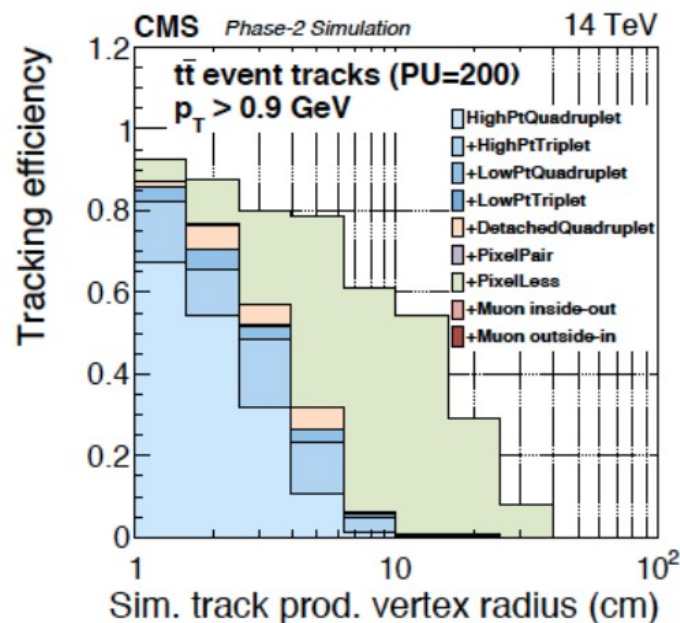
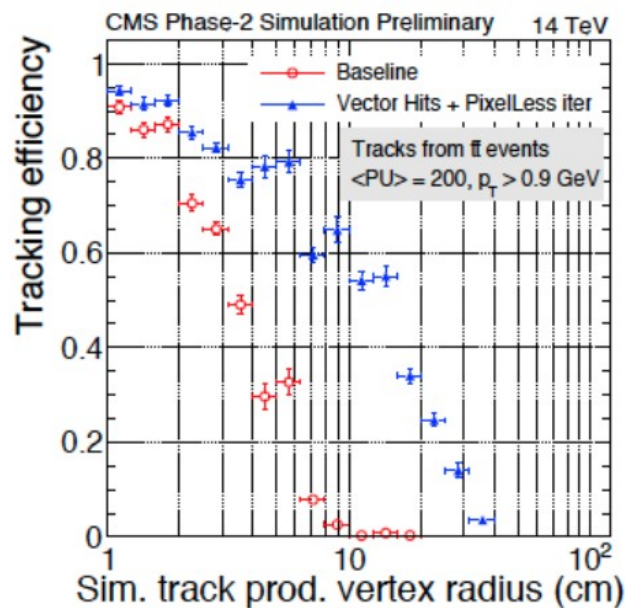
GMSB stop



RPV stop

Plans for displaced vertex/track reconstruction

- › Vertex and displaced tracks have very low efficiency (even lower for HL-LHC)
- › Plan to work in the application of Deep Algorithms to reconstruct/identify these tracks.
 - › Both including the timing and only with spatial information.
- › Two PhD students (Celia Fernández, Pedro Fernández) with focus on this activity.
- › Using Deep CSV (convolutional network for b-tag identification) as a starting point.
- › **Exploration and first steps being carried out right now.**



Conclusions

- IFCA HEP group is increasingly getting into the use of Machine Learning tools.
- Great synergies with the Advanced Computing group with expertise on the subject.
- Computing resources also available at IFCA to perform studies/work.
- Several exploration examples have been already developed.
- Facing now the first large project to reconstruct displaced tracks w/wo timing information.