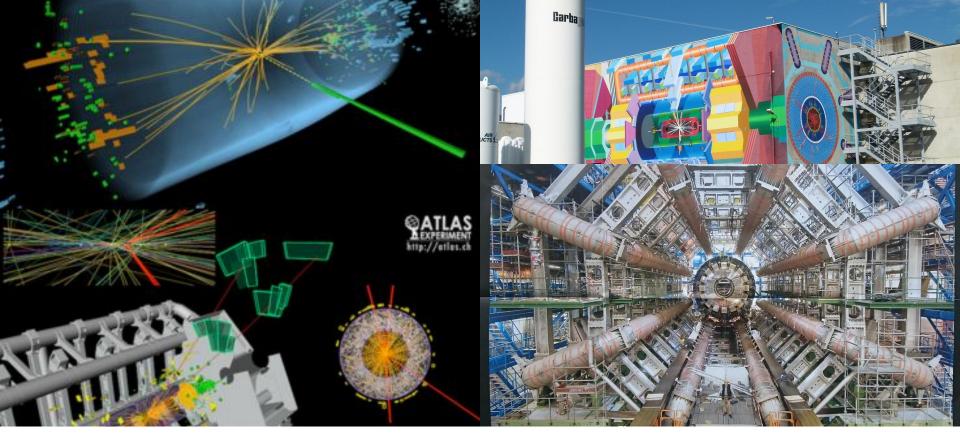
Bachelor & Master Thesis Subjects in ATLAS

Alessandra, Arnau, Craig, Daniel, Janni, Jørgen, Mogens, and Troels

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High Energy Physics (2) ATLAS

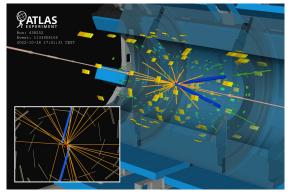




Why High Energy Physics? Why ATLAS?

- Want to probe the basic building blocks of reality
- New particles are either very heavy or very subtle
- We need collisions at high energy (to find heavy particles) and *many* collisions (to find subtle particles)
- The Large Hadron Collider accelerates particles at ten thousand times the energy of everyday life

... match this?



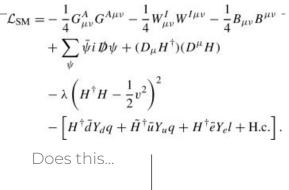




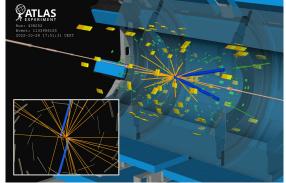
Why High Energy Physics? Why ATLAS?

By **analyzing** collisions **simulated** according to the Standard Model with collisions **recorded** and **reconstructed** in the real ATLAS detector, we can see if there is a mismatch

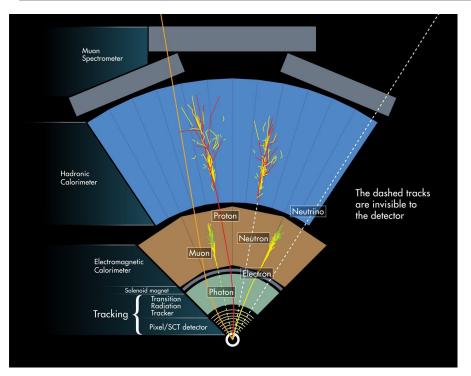
 This mismatch might be "New Physics": extra dimensions, dark matter, extra symmetries (forces), more fundamental particles



```
... match this?
```



What happens in ATLAS hardware, triggering, reconstruction & analysis?



- Hardware: The detector needs to be upgraded, already started!
- **Trigger:** Data needs to be stored or discarded.
- Reconstruction: All the raw measurements (energy deposits in 4D space) need to be converted to "real physics"
- **Analysis:** Prediction and data are compared, and New Physics discovered (or ruled out)





Research Projects in ATLAS









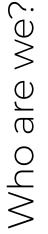
Arnau Analysis



Craig Hardware wigleswt@nbi.ku.dk



Daniel Tracking + ML daniel.murnane@nbi.ku.dk





Janni Analysis



Jørgen FCC beck@nbi.ku.dk



Mogens Hardware + FCC dam@nbi.ku.dk



Troels Analysis + ML petersen@nbi.ku.dk

ATLAS Physics Analysis



- Data analysis in ATLAS is a wide concept: We publish about 100 papers/year across a very varied field. It can broadly be divided into two:
 - **Combined Performance (CP)**: Better performance and understanding of particle identification and reconstruction.
 - **Physics Analysis**: Measurement or Searches based on physics data.
- For CP, our group is working on measuring (and thus optimising) the performance of energy regression on (b-quark) jets. This is a vital input to exploring the Higgs particle (which decays to b-jets 58% of the time).
- For physics, our group is working on a search for the Higgs decay to a Z-boson (decaying to two leptons) and a photon. The decay is rare, which makes it hard. But the final state particles, Z and photon, are not rare, and leaves lots of room for improving the performance (machine learning) and understanding (calibration).



Higgs to jets in ATLAS

Science Question: Does the Higgs couple to itself?

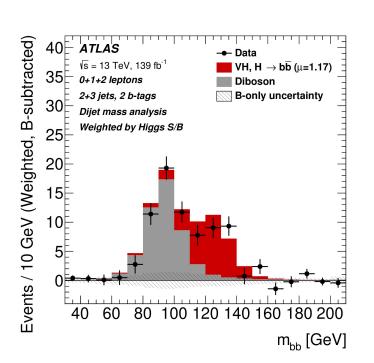
Observation to make: Test if there are events with two Higgs particles.

Challenge: (Di-)Higgs events are very rare (if they exist) and Higgs particles decays most to two b-jets, yielding a very wide mass distribution (below).

Master thesis project:

Combine machine learning with good calibration, to minimise the Higgs mass resolution in real data.

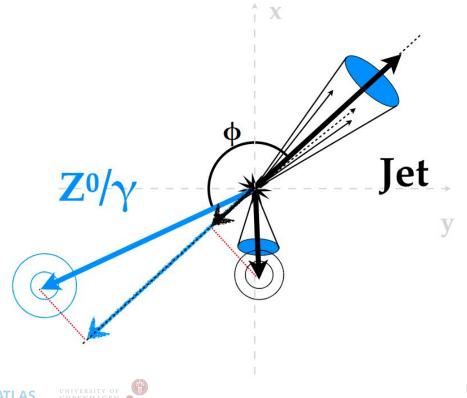
Part of the challenge lies in measuring the performance in data.





Jet energy calibration

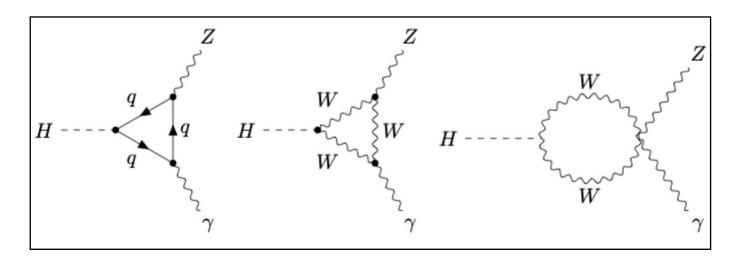




The jet energy calibration can be done based on events like these, or alternatively semi-leptonic ttbar events.



The Higgs decay to a Z and a photon has yet to be observed.



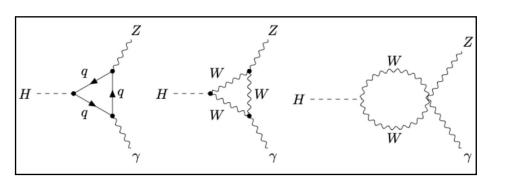


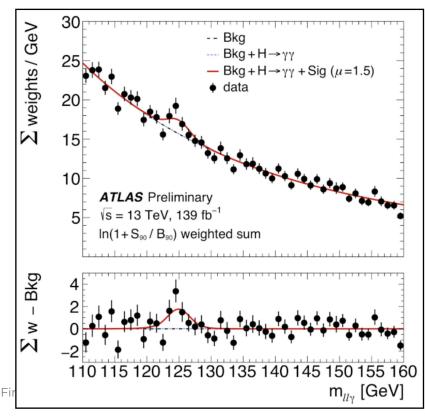
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Searching for a Higgs decay



The Higgs decay to a Z and a photon has yet to be observed.







Generally on ML models

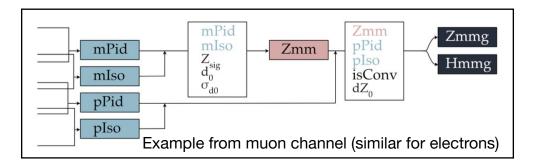


PID models (Pid): Trained for muons, electrons, and photons. Note: A separate Pid model was trained for forward electrons.

ISO models (Iso): Trained for muons, electrons, and photons. Note: The Iso models takes different kinematic variables as input.

Di-lepton models (ZII): Trained for $Z \rightarrow \mu\mu$ and $Z \rightarrow ee$ decays. Combines the Pid and Iso models with track and vertex variables.

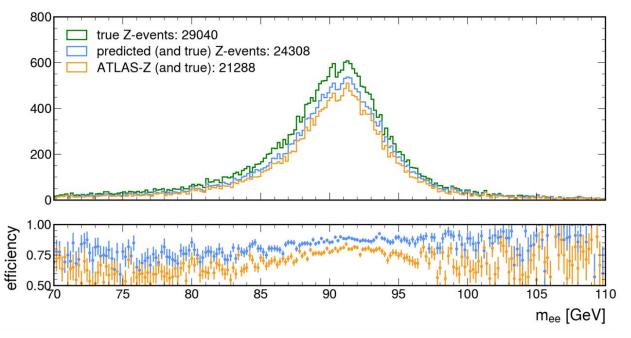
Zllg/Hllg: Models for the full Z \rightarrow ll γ and H \rightarrow Z(\rightarrow ll) γ decays. Combines the di-lepton models with Pid and Iso models for the photon.





Using PID and isolation variables for each lepton, along with their distance in z0, we obtain the following selection improvement in (here Zee) signal.

This can be improved further and expanded to include inv. mass estimate.



The gains for ATLAS are:

- Better Z selection (by 10-20%, already proven)
- Better Z mass determination (by 5-15%, partially proven)
- Likely smaller systematic uncertainties
- If this works well, proceed to $Z/H \rightarrow$ tau tau All of the above gains comes with less work for analysis groups, as the systematics will reduce to a single number.

Finally, our field may gain from spurring research into domain shifts:

Researchers outside particle physics do not have access to such imminent, huge, complex and (with this QT project) curated data/MC samples, where the central issue is domain shift (differences between data and MC).

In ATLAS we have about 1/4 billion Z \rightarrow II signal events in each channel (e,µ) and source (data, MC) along with background, which could become a (re)known public dataset in both size and quality.



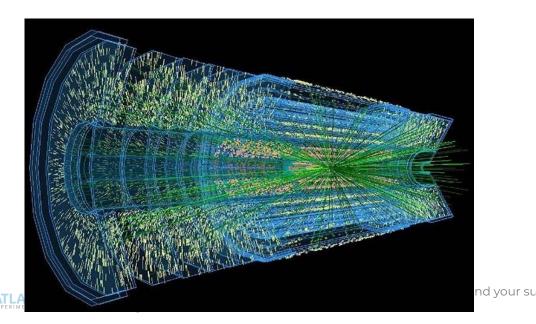


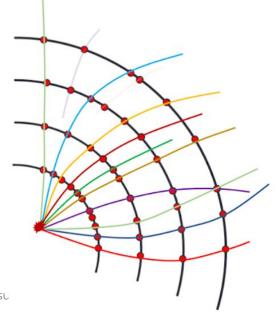
Troels



31

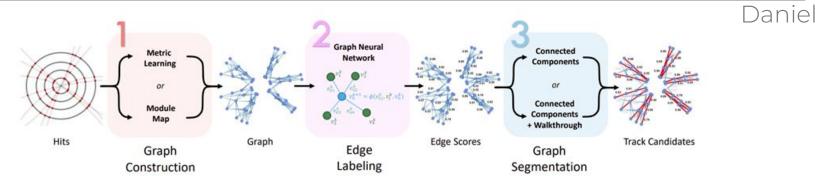
Goal: Take 300,000 points in space, and "connect the dots" to reconstruct the 1,000 interesting particles (and do this 10-100 times per second)





Offline Tracking: Graph Neural Networks & Transformers





Many ongoing projects (~12 students across EU + US) to solve tracking with GNNs and transformers: <u>see link</u>. Both bachelor and masters versions of:

Project A: Improved graph construction with neural module map

Project B: End-to-end tracking with transformers, object condensation and learned clustering

Project C: Large-scale, distributed training and inference across GPUs



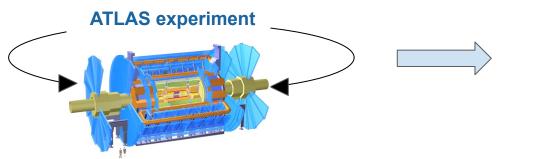


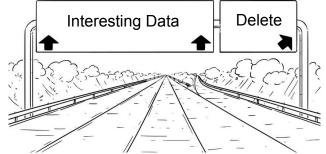
ATLAS Trigger and Data Acquisition system

Alessandra

Amount of data produces in ATLAS: **60 TB per second**.

- Non-feasible and not needed to save everything
- Real-time selection system to select and save only the interesting data (Trigger and Data Acquisition - TDAQ)
 - Critical role- the saved data are used for (ALL) further physics studies





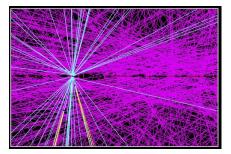


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<u>Alessandra's Signal Processing course</u> in Block 3

Online Tracking in TDAQ

ATLAS TDAQ - Event Filter Tracking



HL- LHC: ten-fold increase in the amount of data to allow observation of rare processes.



Projects

Recent Master Projects:

- GNN for tracking, performance in FPGAs -Sara Schjødt Kjær
- MLP for seeding, performance on GPUs and FPGAs - Mikkel Møller Mødekjær Mikkel accepted as PhD student at CERN!!

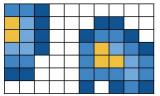
For the future:

- Clustering algorithms

Development and performance studies on GPUs and FPGAs

- Performance studies

On the currently available algorithms





Alessandra

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We're also involved in the production of a new particle detector called the Inner Tracker (ITk)

Both BSc and MSc projects are available

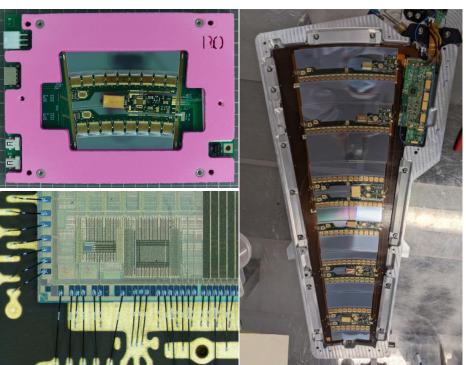
Primary task will be to study the performance of these new detector modules with possible mini tasks in parallel, such as developing GUIs, circuit boards, optimising lab

These projects offer a fantastic and rare opportunity to:

- See very first modules installed at CERN
- Work in an international collaboration
- Be recognised as *THE* local expert in DK
- Visit/work at CERN and shadow experts
- Contribute as author to published paper

The ATLAS ITk Project





Craig



The ATLAS ITk Project

Students who have worked with us in the lab tend to do very well <u>Previous students:</u>

Craig

Jonas - PhD in Uppsala - Now works in nuclear power industry

Catherine - MSc Imperial College

Ashley - PhD in Amsterdam - Now a researcher at Oxford

Oliver - Recently finished BSc - Now working with us!

BTW - we can offer lab tours!



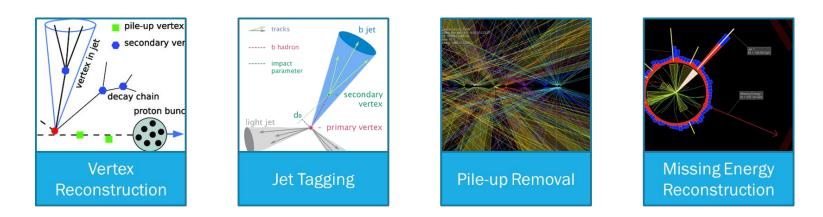






Daniel

Collaborations with colleagues at CERN, Berkeley, Stanford, Adelaide and Yale, to build self-supervised "foundation models" to perform many HEP tasks:







Large A.I. Models for Particle Physics

Several ambitious projects:

Project A: Study a variety of foundation model training techniques (generative, contrastive, etc.) on a new world-class dataset

Project B: Scaling transformers to million-point collision data

Project C: Anomaly detection on encoded physics - can we directly discover "unusual" physics simply by self-supervised training



