

# Computer Vision and Machine Learning Integration for real-time Straw Level Monitoring

David Ari Ostenfeldt, Dennis Chenxi Zhuang  
MSc. Students, DTU Compute  
{s194237, s194247}@student.dtu.dk

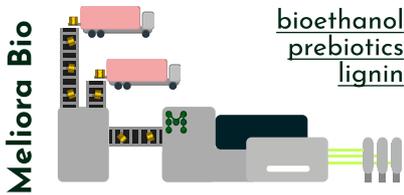
Marjan Mansourvar  
Supervisor, DTU BioEng  
marjma@dtu.dk

Steffen L. Poulsen, Allan Hedegaard  
Supervisors, Meliora Bio  
{sap, alh}@meliora-bio.com



## I. Introduction

- Meliora Bio upcycles straw into *bioethanol*, *prebiotics* and *lignin*.
- During a typical production cycle, **4 tonnes an hour** of straw is processed.
- All straw must pass through a chute during production, which aims to be **50% full** as much as possible.
- Currently, the fullness of the chute is measured with **sensors**.
- Problem:** The sensors are prone to reporting incorrect straw level, and easy to dislodge during maintenance.



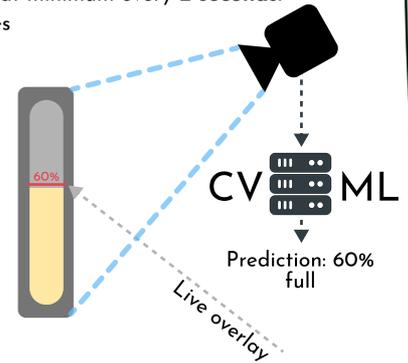
State of the art biorefinery of straw. Something out of Nothing...

## II. The Solution?

- An integrated **computer-vision** and **machine-learning** system, taking input from a mounted CCTV camera pointed at the chute, and outputting a **straw level prediction**.

### What is required?

- Straw level predictions that are within a **±5% threshold** of the *actual* level.
- Needs to report straw level at minimum every **2 seconds**.
- Should be **robust** to changes in the environment, e.g. brightness and murky straw conditions.

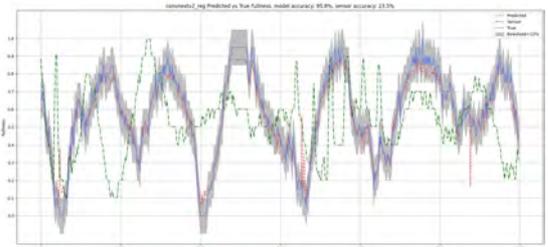


### Our method requires advanced machine learning and AprilTags.

- Requires a PC with GPU
- Can predict and overlay the scores live.
- Deficiencies can be solved on the backend.

## IV. Outcome

- 95.8% test accuracy
  - Production tests to come...
- Still many improvements to make...
- 4-19 predictions per second (depending on model)
- Complete YOLO frame-works seems to work the best currently.



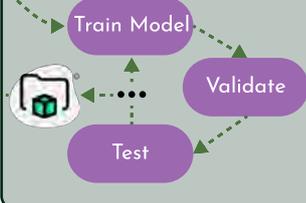
## III. Methodologies

- We have a series of different pipelines for the various steps in this project.

### 1. Data collection



### 2. Model Training



### 3. Inference



### Models

- Goal:** Predict straw level, with **invariance** to chute position.
- 2 Model Methods:
  - YOLO for straw bbox (pixel value), AprilTags for straw level (pixel --> %).
  - YOLO to detect chute, crop+rotate, ConvNeXtv2 or ViT to detect straw level (regression).
- Manual feature additions for input, e.g. Canny Edge Detection and image difference between current and previous frame.

### Training

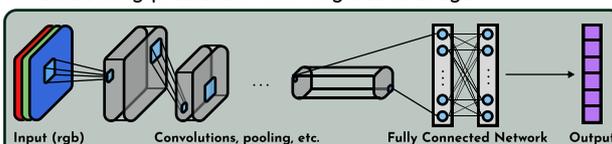
- Utilise data augmentation to **increase robustness**
- Transfer learning
- Cross-validation
- Hyper-parameter tuning

### Live overlay

- AprilTags for higher level resolution
- Camera distortion makes the different levels non-equidistant.
- Smoothing prediction with weighted average

### Reporting level to control

- Convert digital signal to analog, using Arduino
- 0-1 fullness -> 4-20mA



## V. Perspective

- Computer Vision and Machine Learning can solve many problems.
- Exciting opportunities at Meliora Bio to apply these techniques, and might be relevant for other industry stakeholders.
- Can automate many tedious tasks.
  - But also challenging tasks!