# **Promille**afgiftsfonden for landbrug

# YMP re-training on 2016-2021 data

Yield Map Prognosis

Exported on 12/15/2021

## **Table of Contents**

1 Cleaning of yield maps (YMCA)	4
2 Fetch satellite images (YMP)	5
3 Retraining of YMP	6
4 Overall conclusion	9
5 Future work	10

#### **Table of Content**

- Cleaning of yield maps (YMCA)(see page 4)
- Fetch satellite images (YMP)(see page 5)
- Retraining of YMP(see page 6)
- Overall conclusion(see page 9)
- Future work(see page 10)

### 1 Cleaning of yield maps (YMCA)

- It was complicated to use the code from 2019, as it relied on a different data structure and codeenvironment.
- There is no matching polygon/yield map sets in the timespan 2018-2020 (both included) resulting in no useful data for this period, for 2021 only 7 unique yield maps was found.
- Use of field polygons 'Dansk Mark Database' (DMDB) instead of 'Internet Mark Kort' (IMK) resulted in a vastly different amount of matching polygon/yield map sets, even when adding the 7 new yield maps (100 unique sets in 2019, and 77 now)
- The location of the fields in the old and new set of yield maps also differs, as the new yield maps also covers the area of Randers and Ringkøbing, as seen the these plots:



Old yield maps:

- Conclusion: For this pipeline to be automated, and to ensure an easy and reliable way to retrain and expand the data, this solution needs a lot of work. We strongly recommend a full restructure on a more automated pipeline.
  - Ensure all needed field polygons exists in DMDB else add the field polygons from IMK to DMDB, such that we can still rely on our SEGES database, instead of an external data source.
  - Heighten the consistency of the yield data, a way to achieve this is by using the ISOXML files from CropManager instead of the current shape files (manually received from Danish farmers).
    - The developer time spend on normalizing the inconsistencies in the data, makes the current setup unfeasible for reuse in coming years.
  - Our opinion is that YMCA should become a data product, owned and maintained by a dedicated datateam, thus avoiding the technical debt which prevented improvements in the current rerun.

## 2 Fetch satellite images (YMP)

- It was (again) complicated to use the code from 2019, as it relied on a different data structure and codeenvironment.
- Instead of downloading the satellite images for each rerun, we should rely on all ready downloaded images, stored in an easily accessible way e.g. the 'sat-lake'-solution, which pre-cashes all sentinel 2 images in .zarr format on Azure Data Lake.

## 3 Retraining of YMP

- It was complicated to use the code from 2019, as it relied on a different data structure and codeenvironment.
- Model score results:

Model score results:													
	OLD - trained on 2016-2017 yield maps 2016-2021 yield maps						Difference						
Field overall scores	April 10 <sup>1</sup>	May 10 <sup>2</sup>	Jun e 5 <sup>3</sup>	Aug ust 1 <sup>4</sup>	Apri l 10 <sup>5</sup>	May 10 <sup>6</sup>	Jun e 5 <sup>7</sup>	Aug ust 1 <sup>8</sup>		Apri l 10	May 10	Jun e 5	Aug ust 1
train_fiel d_mean_ mae_kg/ ha	327. 9	333. 9	290 .8	283. 1	354. 9	362. 7	331	223. 7		27,0 0	28,8	40,4 0	-59, 40
train_fiel d_mean_ std_kg/ha	451. 9	454. 3	419 .7	381. 4	472. 4	472. 6	458 .4	301		20,5 0	18,3 0	38,7 0	-80, 40
train_fiel d_mean_ samples	79	79	79	79	61	61	61	61		-18	-18	-18	-18
val_field_ mean_ma e_kg/ha	835. 7	926. 3	962 .8	717. 5	648. 2	747	608	674. 7		-187 ,50	-179 ,30	-354 ,20	-42, 80
val_field_ mean_std _kg/ha	1127	122 0.7	124 0.1	863. 7	879	969. 5	799 .9	859. 4		-248 ,00	-251 ,20	-440 ,20	-4,3 0
val_field_ mean_sa mples	22	22	22	22	15	15	15	15		-7	-7	-7	-7
Position overall scores													

 $<sup>{\</sup>tt 1\,http://localhost:} 5000/\#/experiments/4/runs/36759b7 dea 8045a1ab8ab1ed2b6bc427$ 

 $<sup>{\</sup>tt 2~http://localhost:} 5000/\#/experiments/4/runs/8e24b64ef9194a10aeaa776f1d5cc6ae$ 

<sup>3</sup> http://localhost:5000/#/experiments/4/runs/a16a72324a5849819ae37f4e5e7e04ee

<sup>4</sup> http://localhost:5000/#/experiments/4/runs/8616813e09c64e44a6e8e3ea3a566f56

<sup>5</sup> http://localhost:5000/#/experiments/4/runs/642826c95b66489db8199587bd920d94

<sup>6</sup> http://localhost:5000/#/experiments/4/runs/b840288d70a34d8c8e8299dc40d0c130

<sup>7</sup> http://localhost:5000/#/experiments/4/runs/cf7a340a973e4accb98a3f1951f6aa7b

<sup>8</sup> http://localhost:5000/#/experiments/4/runs/9673b10a2eaf43a9b17a7b576a57039d

train_pos ition_mae _kg/ha	1410 .6	135 9.8	129 5.5	122 3.6	141 0.5	134 9.9	128 3.6	119 9.2	-0,1 0	-9,9 0	-11, 90	-24, 40
train_pos ition_sam ples	3835 94	383 594	383 594	383 594	295 733	295 733	295 733	295 733	-878 61	-878 61	-878 61	-878 61
val_positi on_mae_ kg/ha	1729 .6	171 4.4	170 7.5	147 3.7	146 2.2	155 2.2	132 7.5	127 1.9	-267 ,40	-162 ,20	-380 ,00	-201 ,80
val_positi on_sampl es	8188 0	818 80	818 80	818 80	774 11	774 11	774 11	774 11	-446 9	-446 9	-446 9	-446 9

- We do not see significance changes in any (i.e. field-level or position-level) of the training performance scores. The small changes are likely to be a coincidence, given the changes in the dataset.
- However, both field-level and position-level validation Mean Absolute Error (MAE) scores are improved significantly (~200kg/ha) for all prediction dates.
- But due to the smaller sample size we decided to perform a ad hoc cross-validation to conclude if other train/val splits results in the same model improvements.
- 5-fold cross-validations results of val\_position\_mae\_kg/ha (see notebook YMP-678\_crossvalidation\_of\_2016-2021\_yield\_data.ipynb<sup>9</sup>):

	Supp ort (# pixel s)	04-10_ MAE	04-10 _STD	05-10_ MAE	05-10_ STD	06-05 _MAE	06-05 _STD	08-01_ MAE	08-01 _STD
fold_num_1	1080 92	1715,8 1	2169.0 7	1745.2 7	2298.6 7	1617.6 3	2143.2 4	1369.40	1907. 88
fold_num_2	1782 74	1558.8 5	2063.9 6	1493.9 0	2028.1 6	1578.4 1	2035.5 1	1505.94	1991. 16
fold_num_3	1554 36	1726.9 6	2337.2 7	1722.4 6	2322.1 8	1727.6 6	2307.6 2	1461.70	1989. 68
fold_num_4	1904 02	2901.5 3	3368.2 2	2791.3 0	3317.6 4	2762.4 6	3348.7 4	2449.48	3166. 34
fold_num_5	1140 84	1742.1 7	2303.1 3	1708.3 8	2178.5 7	1594.6 8	2100.7 5	1458.95	1902. 38

<sup>9</sup>https://bitbucket.seges.dk/projects/DDS/repos/yield\_map\_prognosis/browse/data\_analyses/ YMP-678\_crossvalidation\_of\_2016-2021\_yield\_data/YMP-678\_crossvalidation\_of\_2016-2021\_yield\_data.ipynb? at=refs%2Fheads%2FYMP-678-cross-validation-notebook-on-new-ymp-models-with-2021-data

Mean over folds	 1929.0 6	2448.3 3	1892.2 6	2429.0 4	1856. 17	2387.1 7	1649.0 9	2191. 49
Scores over total pixels	 1987.1 8	2678.4 9	1941.7 1	2647.0 4	1919. 75	2623.2 7	1710.4 9	2397. 65
Scores over yield maps	 1023.6 9	1555.2 4	1138.5 0	1607.3 6	1177. 33	1705.1 9	902.73	1359. 97

- We see the MAE for each fold is higher than for the previous trained model, thus the mean over the folds are also significant higher (300-500 kg/ha higher). Thus the specific train/val split used previously benefitted the model, but its performance is not generalizable.
- The standard deviation (STD) of each prediction error, shows that 68% of all validation samples has less than +/-2300kg/ha in prediction error.
- However, our cross validation still shows that more data (closer to harvest) improves the model performance.

#### • Conclusion:

- As a larger, and more consistent, pool of data greatly benefits the model, we think this should be the main focus going forward.
  - A way to achieve this by ensuring all data fetching to the YMP pipeline comes from data products used across SEGES, i.e. yield data from CropManager.

#### 4 Overall conclusion

- Our rerun of the YMCA pipeline resulted in fewer yield maps, i.e. 100 unique fields in 2019, and 77 now. However, the new fields also covers the area of Randers and Ringkøbing, i.e. more divergent soil/climate types.
- Our retrained YMP models on the new 77 yield maps, show models improvements of ~200kg/ha. However, following cross validations, the results were worse than our train/val model. This shows the dataset is not generalizable, thus resulting in better model-performance than expected in real life application.
- The smaller sample size is troubling, but due to the slightly improved train/val model results, we would recommend deploying the new models to production.

#### 5 Future work

- 1. Implement automated pipeline for YMCA based on ISOXML yield maps.
- 2. Implement YMP training pipeline to fetch all geo-data from SEGES data cache store (Spatio-Temporal Data Store Definition<sup>10</sup> like Sat-Lake)

  3. Move YMP training/deployment pipeline to Azure Machine learning.
- 4. Implement cross-validation in to the YMP training pipeline.

<sup>10</sup> https://confluence.seges.dk/display/YMCA/Spatio-Temporal+Data+Store+Definition