

CGI









Service Pilot - Improved access to credits for smallholder farms in Kenya



- Support Kenyan company Farmdrive, who provide micro endits to farmers
- Integrate EO data into methods for objective assessment of credit risk
- Minimize operational costs while increasing the efficiency and operational capacity of financial services providers to support meeting the demand for credit by farmers



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Relevant crops & parameters to be delivered

Sorghum

- Scientific name: Sorghum bicolor
- Other names: great millet, durra, jowari, milo
- First planting season: April-August (Long rains)
- Second planting season: September December (short rains)

Maize

- Scientific name: Zea mays
- Other name: Corn
- Planting seasons: 1. Late October, November, December and January – Right season. 2. June – August- wrong season, intercropping of both maize and beans.

Three different EO variables were delivered to FarmDrive in Kenya:

- SMOS Soil Moisture Products
- LAI time-series derived from Sentinel-2
- Yield estimates for the two chosen counties

For all variables, a time-frame of November 2018 to November 2019 was considered.



Relevant Counties and Available In-Situ Data







Sample farm_1 - Sorghum

Location: Siboti ward, Bungoma county Crop: Sorghum Planting date: 3rd October 2019 Crop stage: Weeding Main crops: Maize, groundnuts, Tobacco, beans, sorghum Size of the farm: 0.5 Acres Data collection date: 28th October 2019.



Kipsonoi ward sample farm_1 Location: Kipsonoi ward, Bomet county Crop: Maize Planting date: 14th October 2019 Crop stage: Weeding Main crops: Maize, beans Farm size: 1 Acre Data collection date: 30th October 2019.



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Crop Parameters to be delivered





Leaf Area Index for whole counties and past year Example: 16/07/19

Yield Estimates for maize (Bomet) and sorghum (Bungoma)

Challenges for yield estimation:

- Very small fields
- Crop types of fields not known
- Only very coarse information available on planting / harvesting dates
- Very few in-situ data for calibration / validation



SMOS Soil Moisture time series for whole counties and past year

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From Sentinel-2 Data to equidistant LAI time series without prior land use information

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Step 1:

Random selection of pixels on masked remote sensing image stacks (agricultural land only)

Step 2:

Detection of plant phenological parameters

- Vegetation onset
- · Vegetation maximum
- Vegetation offset

Step 3:

Selection of LAI time series matching the following criteria:

- All parameters have been derived (no missing value)
- $Min1_{pixel} = Min1_{samplefarm}$ ± 15 DOYs

Step 4:

Outlier removal, linear interpolation, moving window smoothing







TIME

± 15 DOYs



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Example for resulting LAI time series as input into crop growth model





Interpolated and smoothed LAI time series of maize pixels in Kenya

Time range: October 2018- February 2019

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Pixel selection for maize in Bungom





- Only few commercial farms with large fields in county
- Selected pixels are all on smallholder fields
- Image from 04/10/2019: Second harvest phase of sorghum, planting phase for maize



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Pixel selection for sorghum in Bom





- Only few commercial farms with large fields in county
- Selected pixels are all but one on smallholder fields
- One S2 tile (East) had better data availability, so more pixels could be found
- ▲ Image from 04/10/2019

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Pixel for sorghum

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40 km

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Crop Growth Model PROMET

Simulation of plant growth and water, carbon and nutrient household

PROMET attributes:

- physically based
- spatially distributed
- raster based (10m 1km)
- hourly calculations



Atmosphere

Vegetation

So:

Capillary Rise



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Assimilation of Satellite Data into Crop Growth Model





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Bungoma wards - Model Data Sorghum

| Yield class | Flag | Average t/ha | Minimum t/ha | Maximum t/ha |
|---------------|------|-----------------|-----------------|-----------------|
| Failing | 1 | 0.13 | 0.00 | 0.30 |
| Weak | 2 | 0.47 | 0.30 | 0.65 |
| Below Average | 3 | 0.82 | 0.65 | 1.00 |
| Average | 4 | 1.17 | 1.00 | 1.34 |
| Favourable | 5 | 1.52 | 1.34 | 1.69 |
| Strong | 6 | 1.87 | 1.69 | 2.04 |
| Superior | 7 | 2.21 | 2.04 | 2.39 |



Kabuchai

0%

Bumula



Kanduvi

Kimilili

Distribution of simulated sorghum yield classes in 11 different Constituencies in Bungoma County, Kenya

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Lugari



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