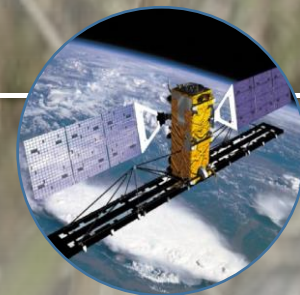
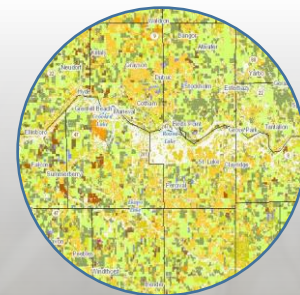




Agriculture and
Agri-Food Canada

Agriculture et
Agroalimentaire Canada

Canada



Application of Remote Sensing in Operational Agrometeorological Advisory Services

Catherine Champagne, Environmental Scientist

National Agroclimate Information Service

Science and Technology Branch

Agriculture and Agri-Food Canada

National Agroclimate Information Service (NAIS)

Science and Innovation
in Agroclimate

Development and
Application of Geospatial
Analytical Tools

Expert Interpretation and
Analysis

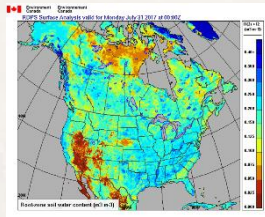
Building Partnerships
and Collaborations

- Applied science & operational delivery of information and tools to monitor impact of climate on agriculture
- Monitoring, early warning using authoritative data and models are key to understanding climate risks and impacts
- NAIS activities focus on four main thematic areas:
 1. **Mitigating and Adapting to Climate Change**
 2. **Increasing the resilience of agro-ecosystems**
 3. **Accelerating Digital Transformation**
 - Building and maintaining cross-cutting agroclimate data sets
- All information made available via **Weather and drought** related to agriculture website:
 - <https://agriculture.canada.ca/en/agricultural-production/weather>

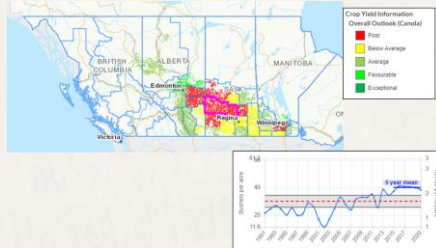
Some key activities...

Research & Development

Modelled Weather Data



Sustainability & Crop Metrics



Near-Real Time Data



Operational

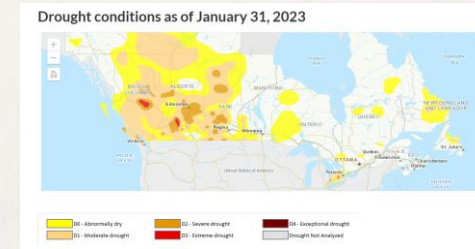
Weather and drought related to agriculture website

Drought Watch and agroclimate
 Maps, reports and data, historical and current weather and climate, drought, yield forecasts and impacts across Canada.

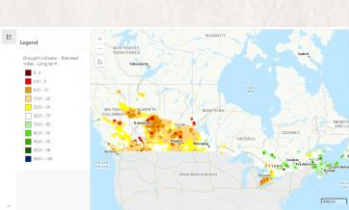
Current and historical conditions
 Canadian Drought Monitor
 Precipitation, temperature and soil moisture affecting agriculture
 Evaporative Stress Index

Quick links
 Current drought conditions
 Drought Watch interactive capacity assessment
 Agroclimate Impact Reporter
 Drought Impacts

Canadian Drought Monitor



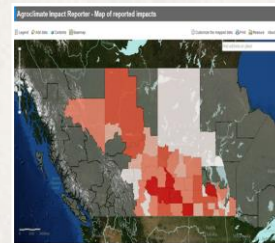
Blended Drought Indices



Drought Outlook



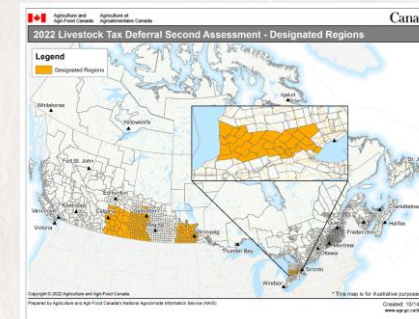
Agroclimate Impact Reporter



Climate-Related Risk Committee

| Seasonal trend | CC | IC | IS | IS2 | IS3 | IS4 | IS5 | IS6 | IS7 | IS8 | IS9 | IS10 |
|----------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| Jul 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |
| Aug 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |
| Sep 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |
| Oct 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |
| Nov 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |
| Dec 1, 2024 | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green | Green |

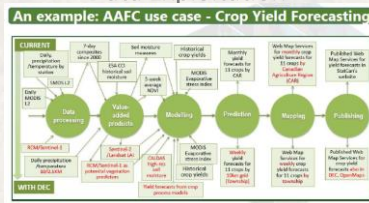
Livestock Tax Deferral



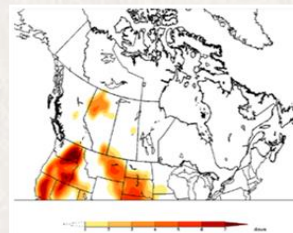
High Resolution Yield Modelling



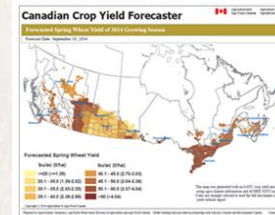
High Performance Computing tools For Big Data Exploitation



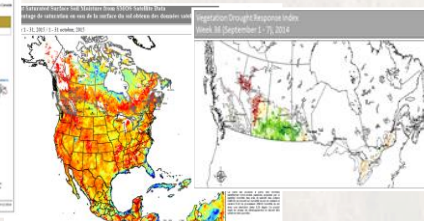
Extreme Weather Indices



Crop Yield Forecaster



Satellite Soil Moisture VegDRI, Evaporative Stress Index



How Do We Collect Geospatial Information about Agricultural Sector?

1. Surveys – Crowd Sourcing

- Questionnaires (census, statistical surveys)
- windshield surveys
- Field data collection

2. Monitoring Stations

- Automated Weather Stations
- Doppler Radar

3. Satellite Remote Sensing

- Optical, thermal, radar, microwave, gravity satellites
- Polar orbiting or geostationary
- Resolution from <1m to >10km

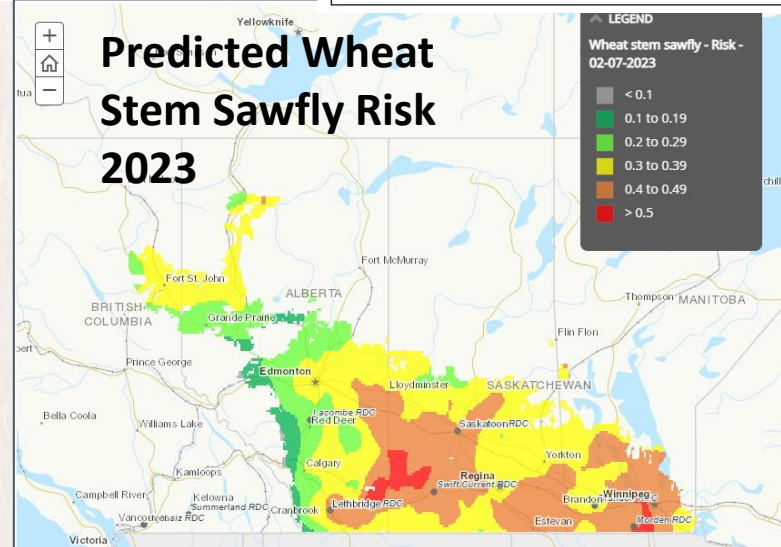
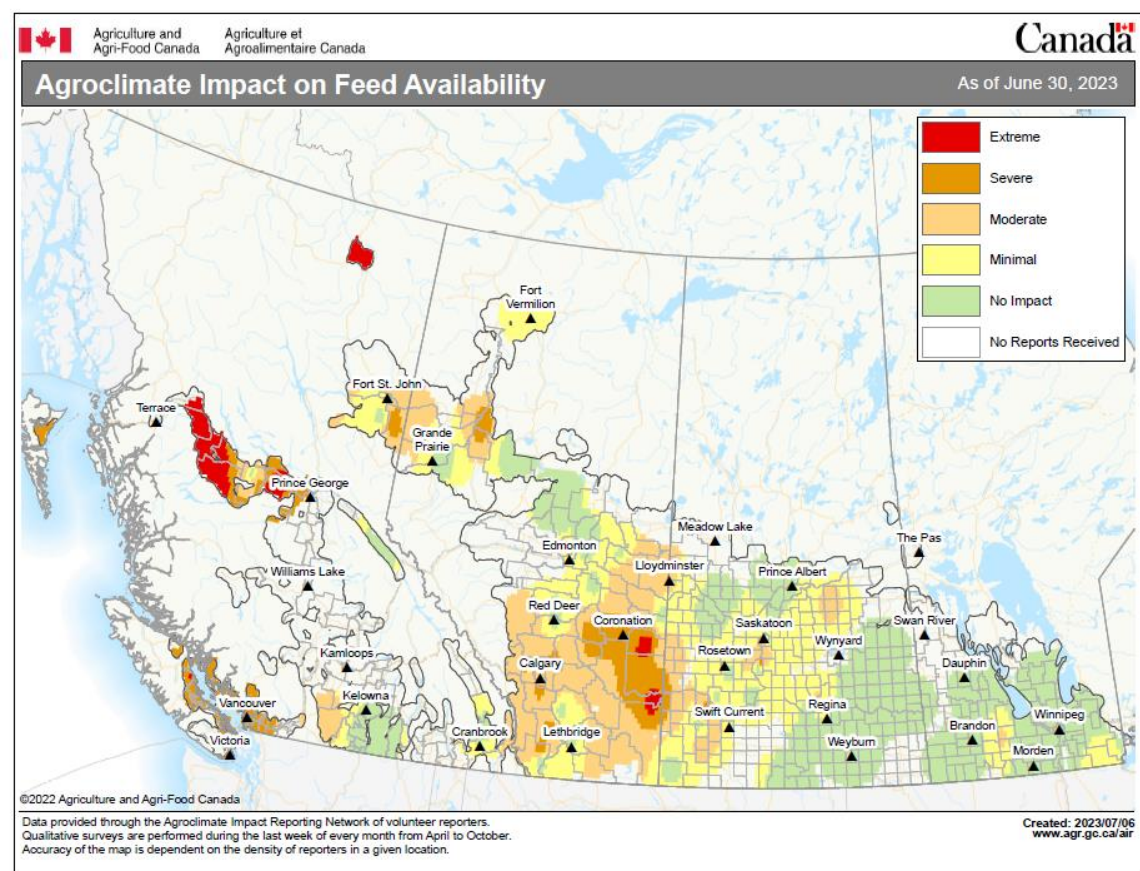
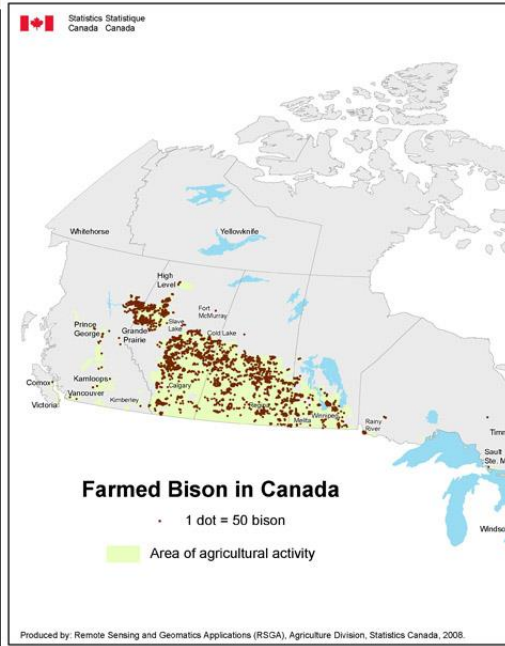
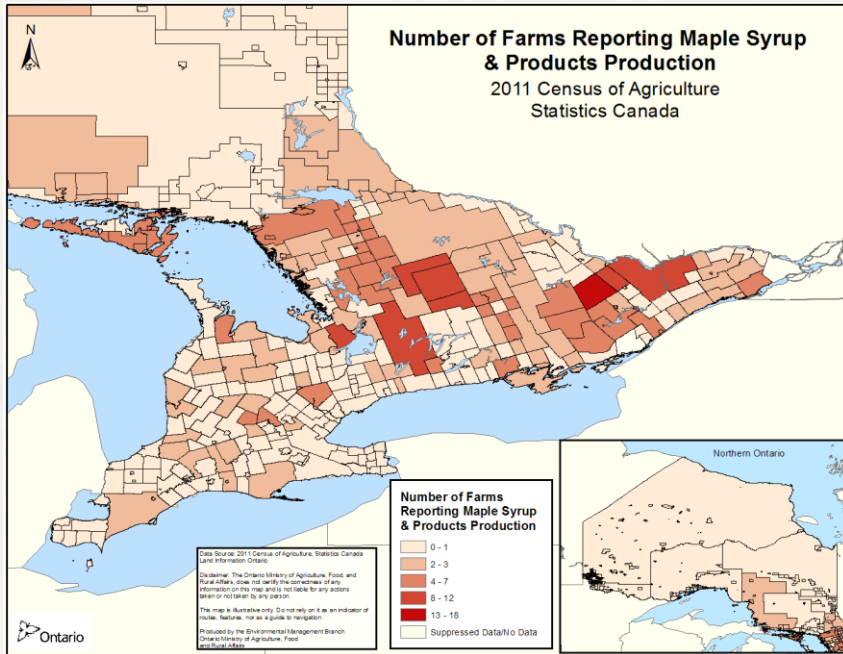
4. Models

- Use statistical or physics based model to simulate information

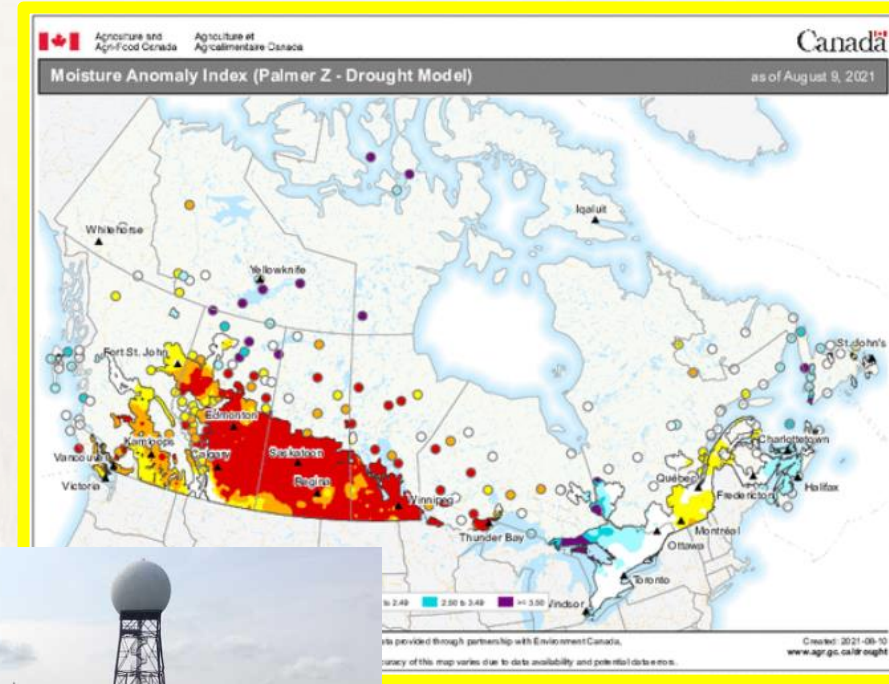
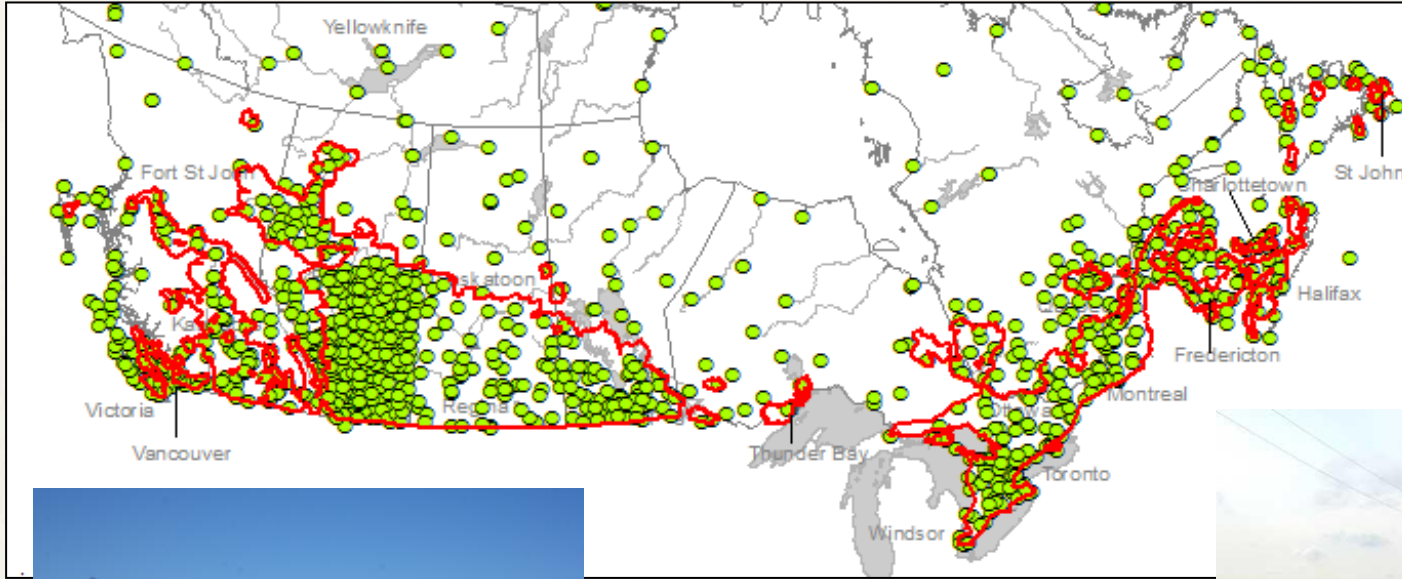
Many agroclimate services rely on using current data in combination with historical normals for determining risk

Long term, consistent data records are key for observing change

Surveys



“Network of Networks” for Agroclimate Data: Station-Based Monitoring



Agroclimate Data uses raw weather inputs to create indicators of heat and water accumulation relevant to crop growth

>2500 climate station locations

- ≈ 1000 ECCC
- ≈ 420 CoCoRHaS
- ≈ 450 Alberta Environment
- ≈ 100 Manitoba Ag.
- ≈ 80 PQ Mesonet
- ≈ 10 New Brunswick DNR

Near Real Time Monitoring System

- Daily quality control of weather stations fully automated system)

- Use current data, station quality rankings and historical gridded data to produce gap-filled time series for estimation of large number of agroclimate measures:

- **Precipitation**

- Percentiles (30, 60, 90, 180 day, Growing Season, Winter)

- Accumulated (up to 1825 days)

- Dry Spells

- **Temperature**

- Max Temp

- Min Temp

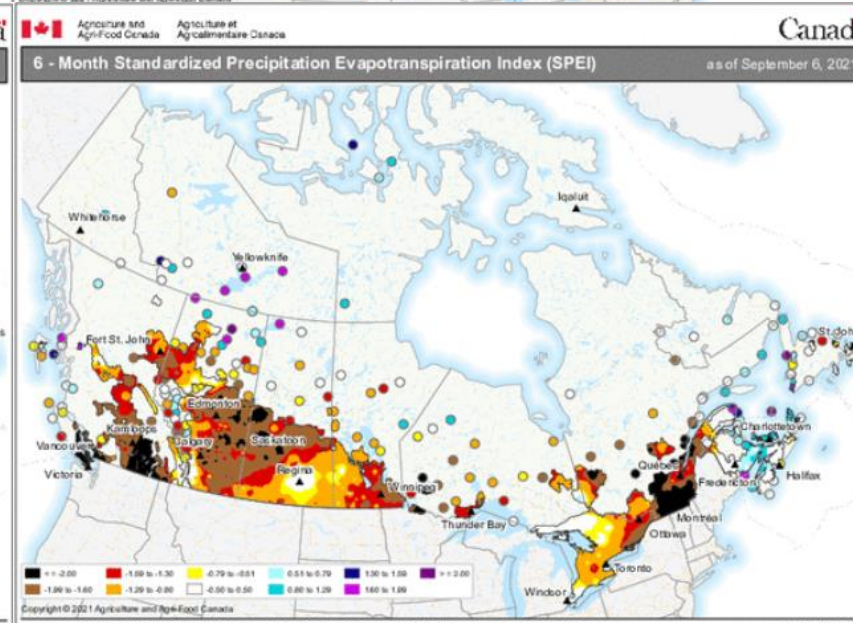
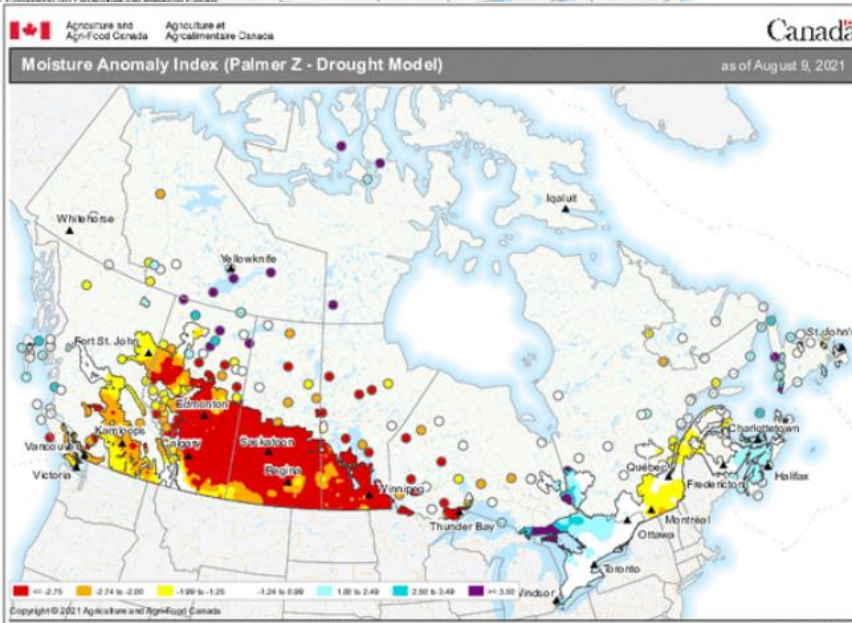
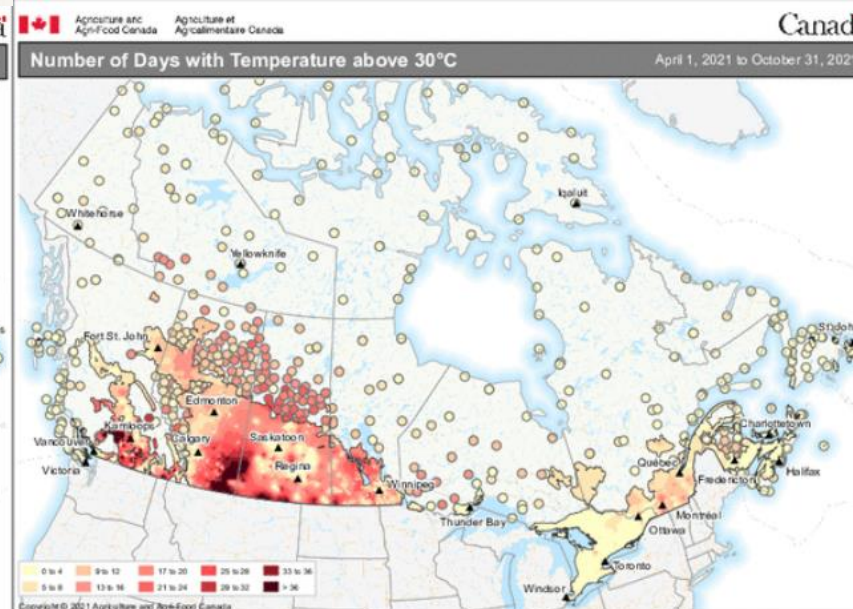
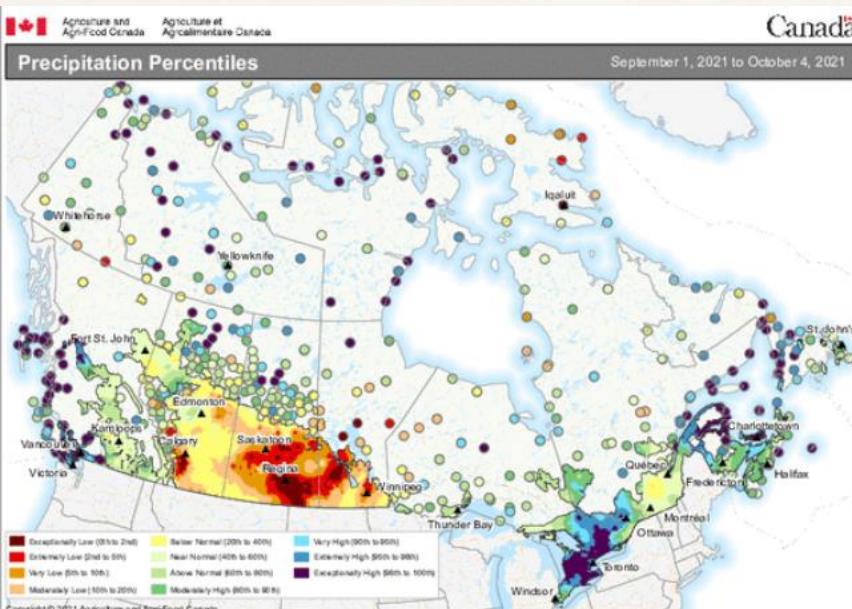
- Heat Wave

- Crop Heat Units

- Temperature Anomalies

- **Drought Indicators:**

- Palmer Z, Hydrological, Soil Moisture, SPEI, SPI



Satellite Remote Sensing

Different frequencies for different applications:

Optical Data

- Crop Health
- Leaf Area Index
- Biomass
- Crop Residue Cover
- Crop Types
- Crop Growth Stage

Radar Data (Active)

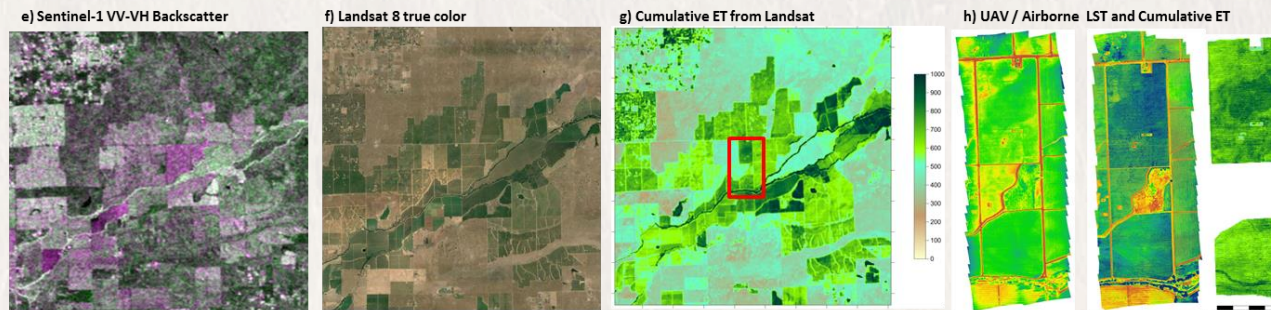
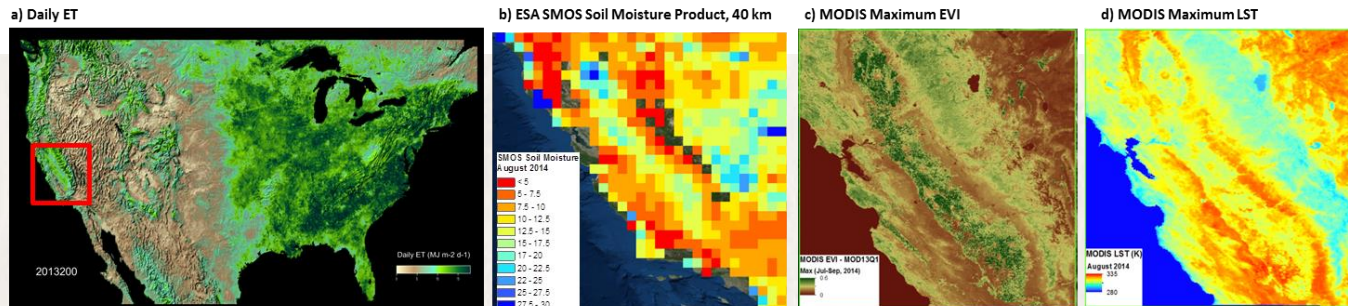
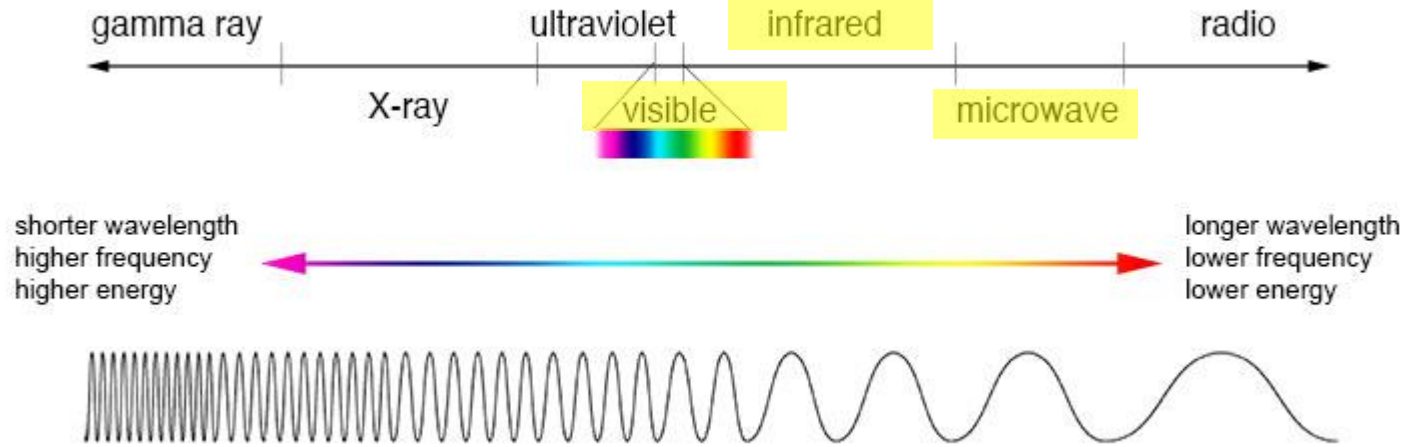
- Soil Moisture
- Surface Hydrology
- Plant Structure
- Crop Biomass
- Surface Roughness/Tillage

Passive Microwave Data

- Soil Moisture
- Vegetation Water Content

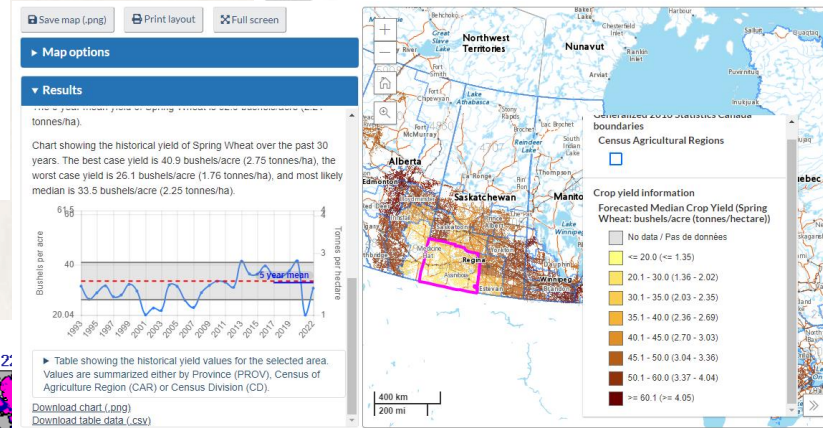
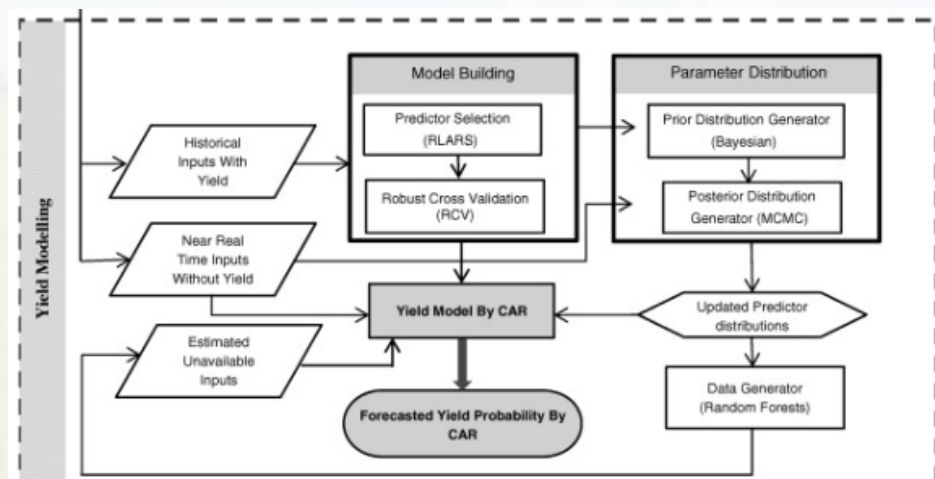
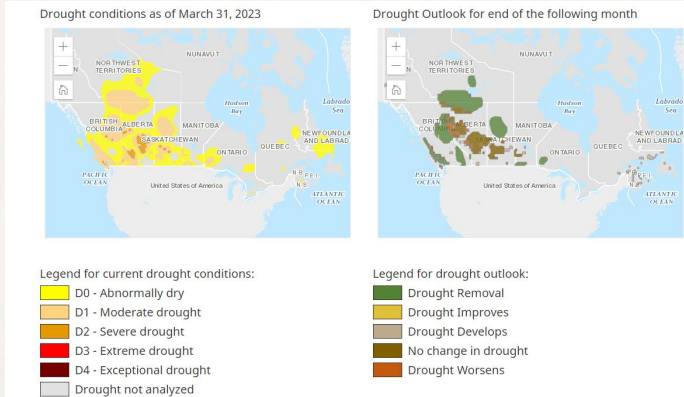
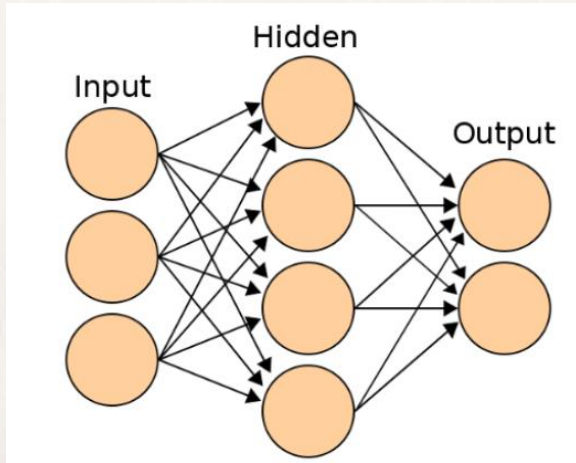
Thermal

- Temperature
- Evapotranspiration

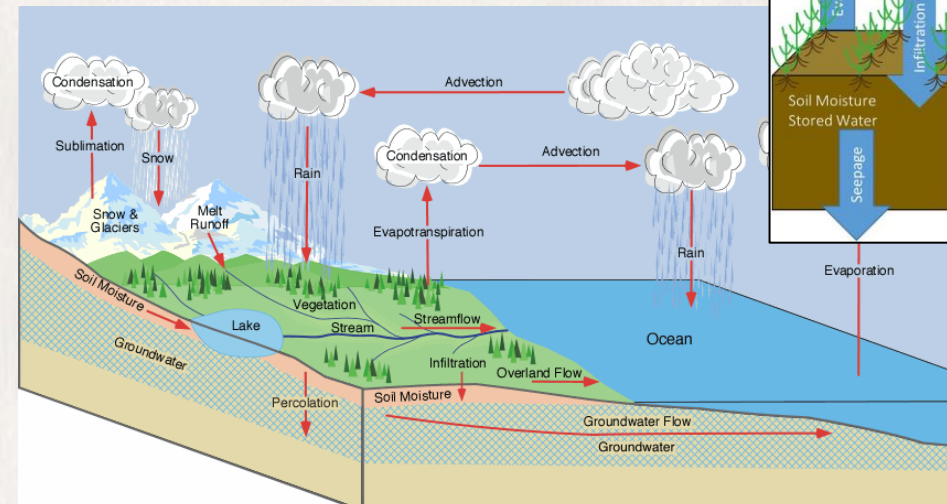
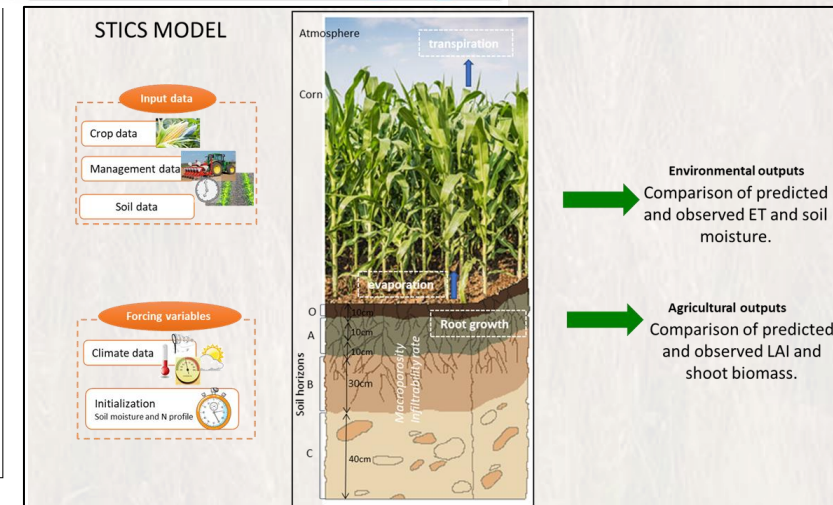
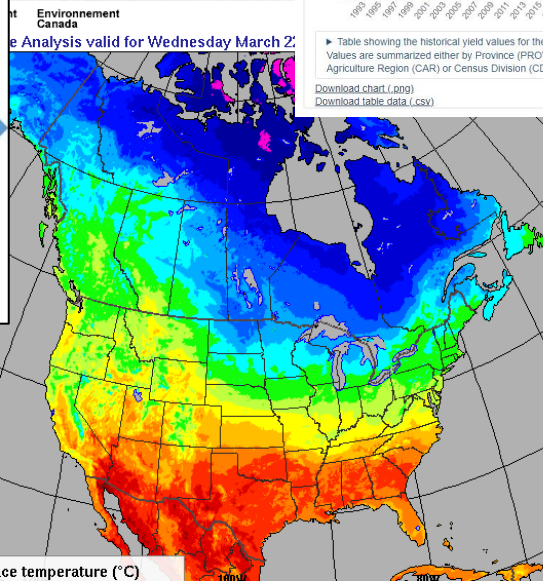
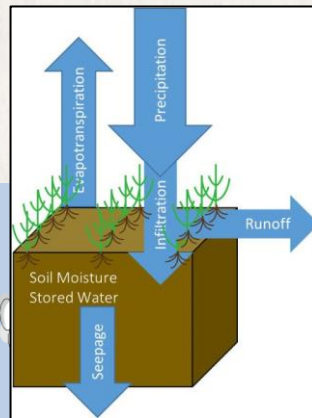


Models

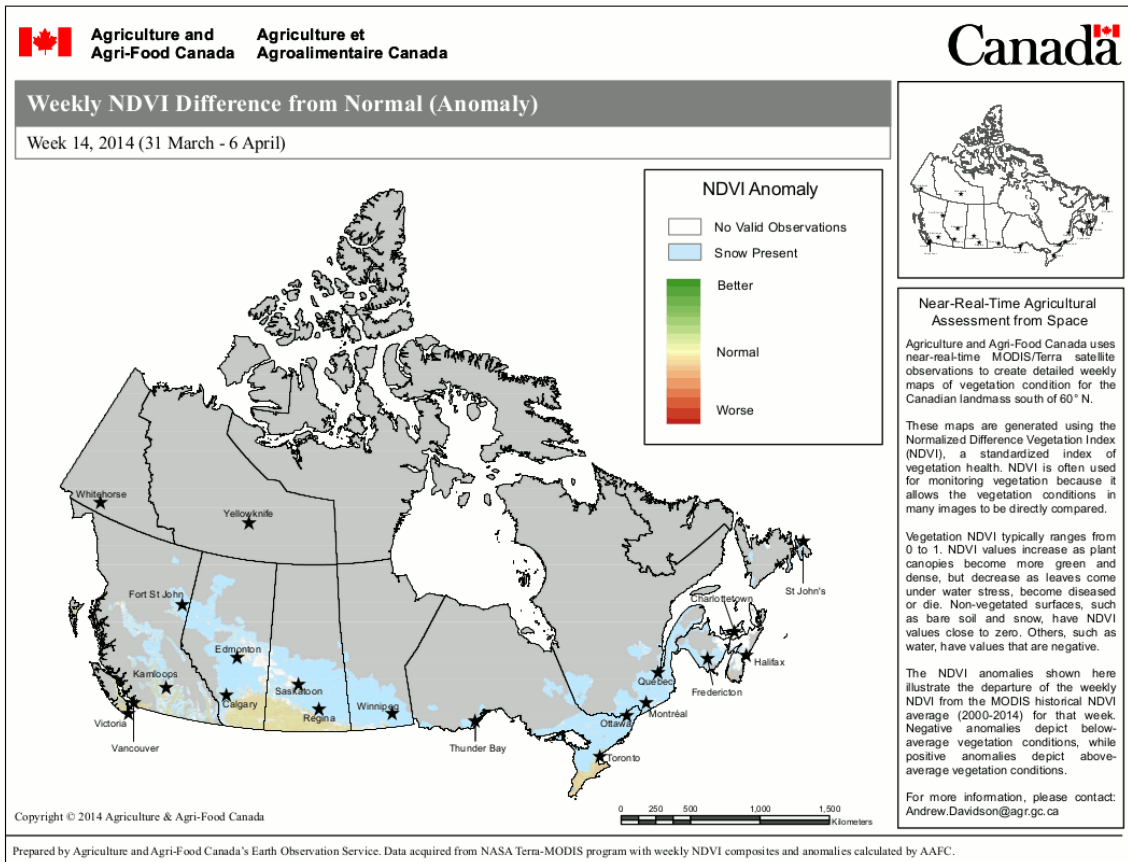
1) Statistical / Empirical / Machine Learning / Artificial Intelligence



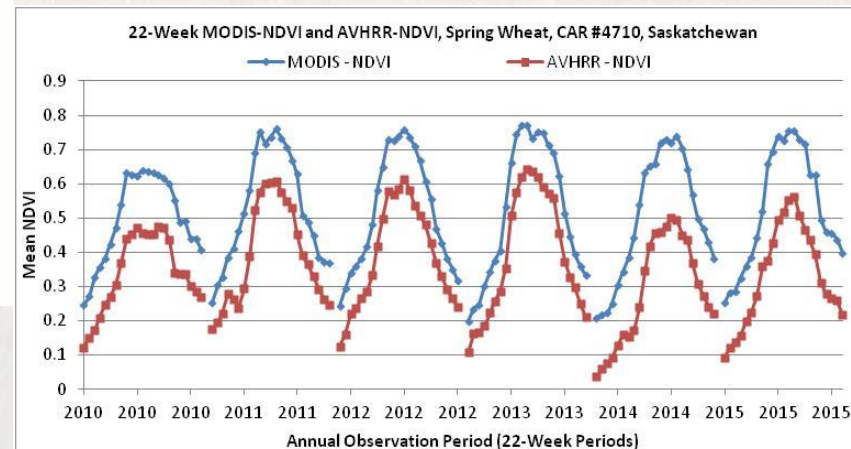
2) Physical/Process-Based



Earth Observation for Direct Monitoring of Agroclimate Conditions: Vegetation Indices

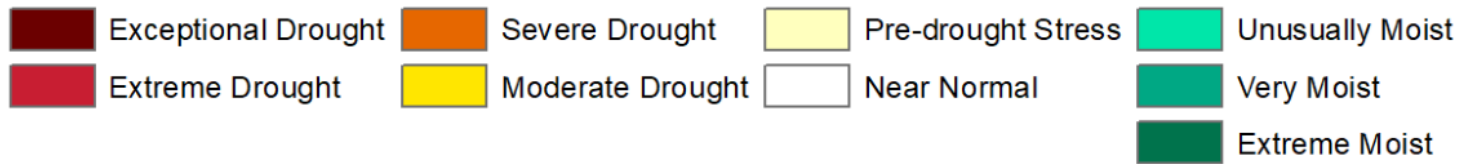
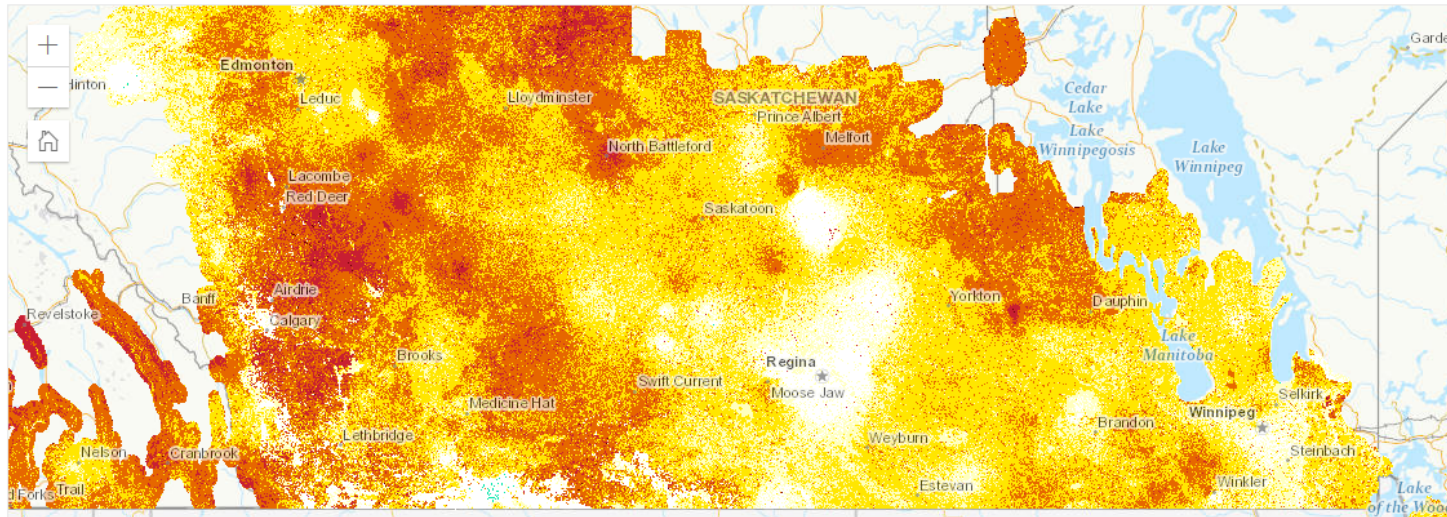


- Normalize Differenced Vegetation Index (NDVI) and other similar indices use band ratios to quantify different between red/near infrared surface reflectance to monitor plant health at many scales
- Long term data records from “moderate” resolution sensors, such as AVHRR, MODIS and now VIIRS are key to these data sets

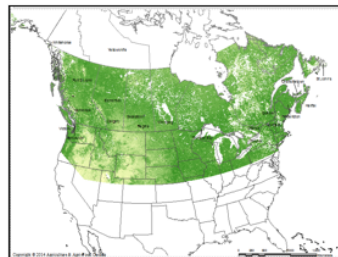


Maximum 7-day NDVI composites from MODIS comparing current conditions to long term average

Vegetation Drought Response Index

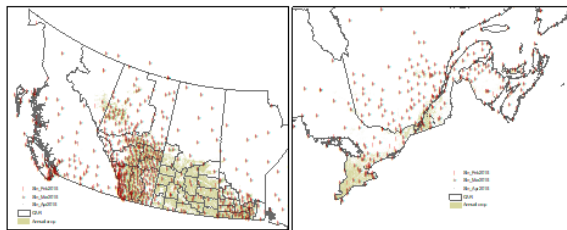


Remote Sensing Component



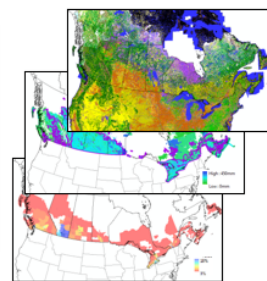
Satellite Normalized Difference Vegetation Index (NDVI)

+ Climate Component



Standardized Precipitation Index (SPI)
Palmer Drought Severity Index (PDSI)

+ Biophysical Component



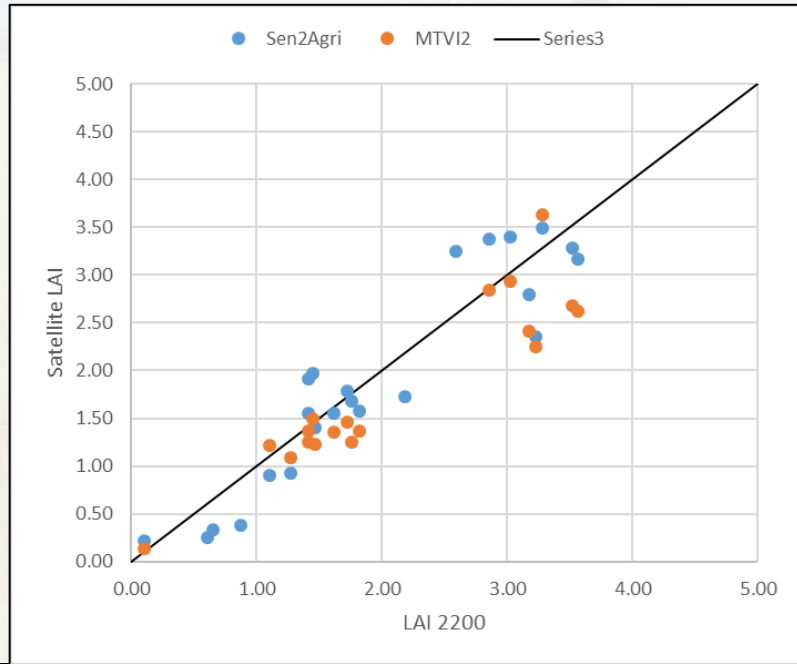
Land Cover
Irrigation
EcoZones
Soil Water Holding Capacity



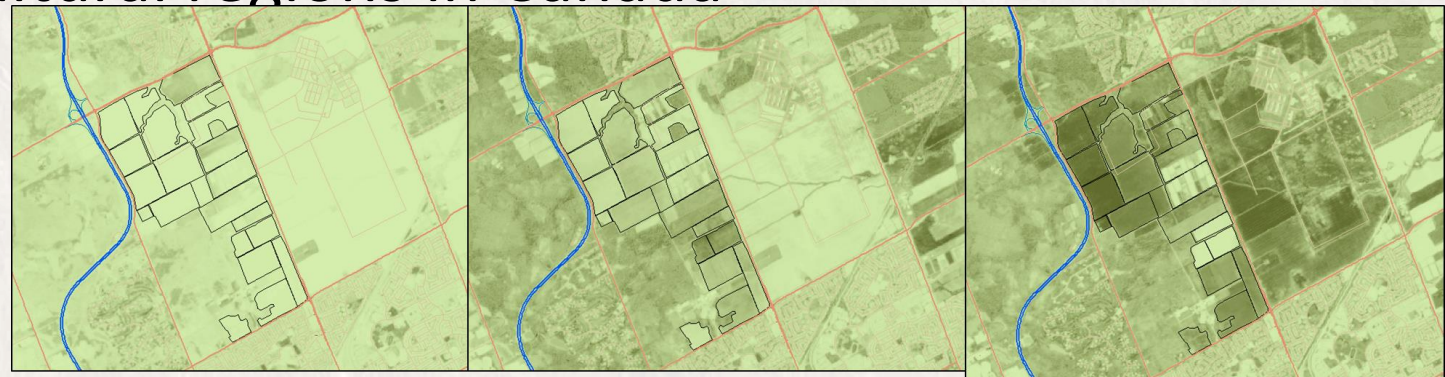
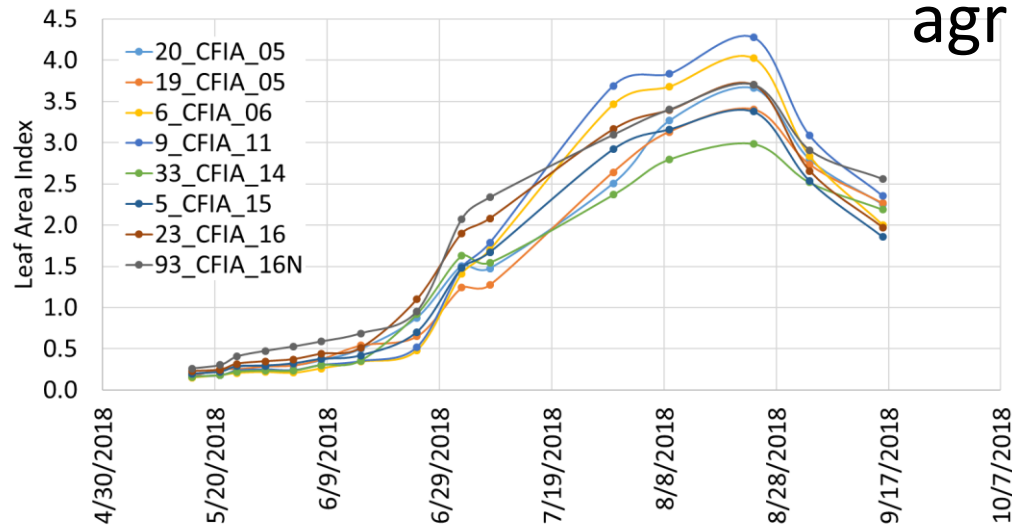
- Combines time series satellite NDVI (MODIS) and station-based drought indicators (Standardized Precipitation Index) and biophysical data (land cover, irrigation, ecozones) to model drought severity at pixel-scale

- Tadesse, T., Champagne, C., Wardlow, B.D., Hadwen, T.A., Brown, J.F., Demisse, G.B., Bayissa, Y.A., Davidson, A.M. (2017). Building the vegetation drought response index for Canada (VegDRI-Canada) to monitor agricultural drought: first results. *GIScience and Remote Sensing*, [online] 54(2), 230-257. <http://dx.doi.org/10.1080/15481603.2017.1286728>

Leaf Area Index



- LAI retrieval algorithm in Sentinel-2 Toolbox (SNAP) shows good correspondence with ground measurements (based on Neural Network Model) at 20m resolution
- Tool has been translated for Google Earth Engine into the LEAF Toolbox (<https://github.com/rfernand387/LEAF-Toolbox>) – can be run both as Java Script or using Python API
- Pilot is currently under way to run this for all agricultural regions in Canada



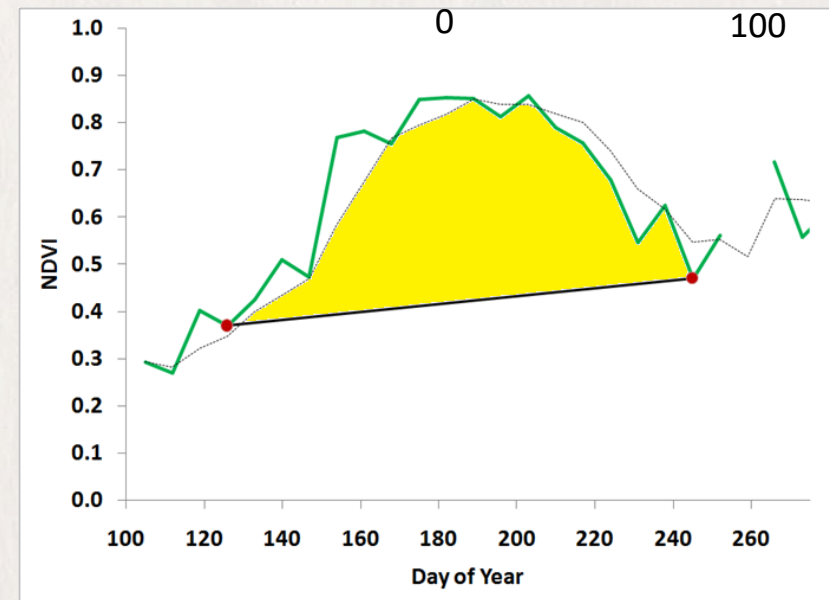
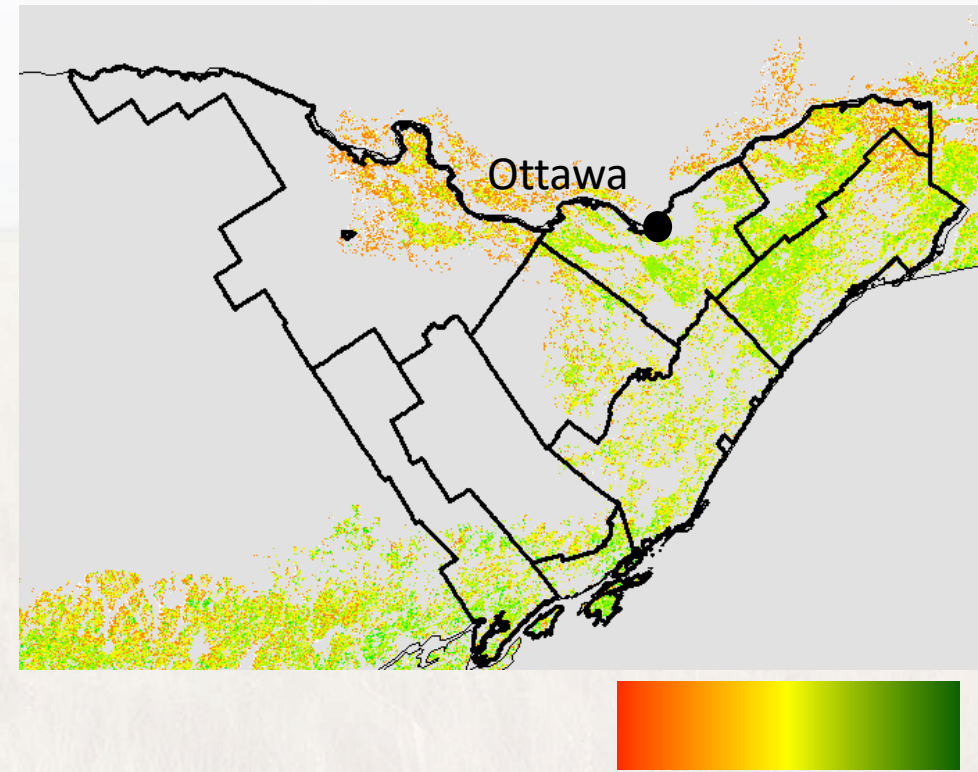
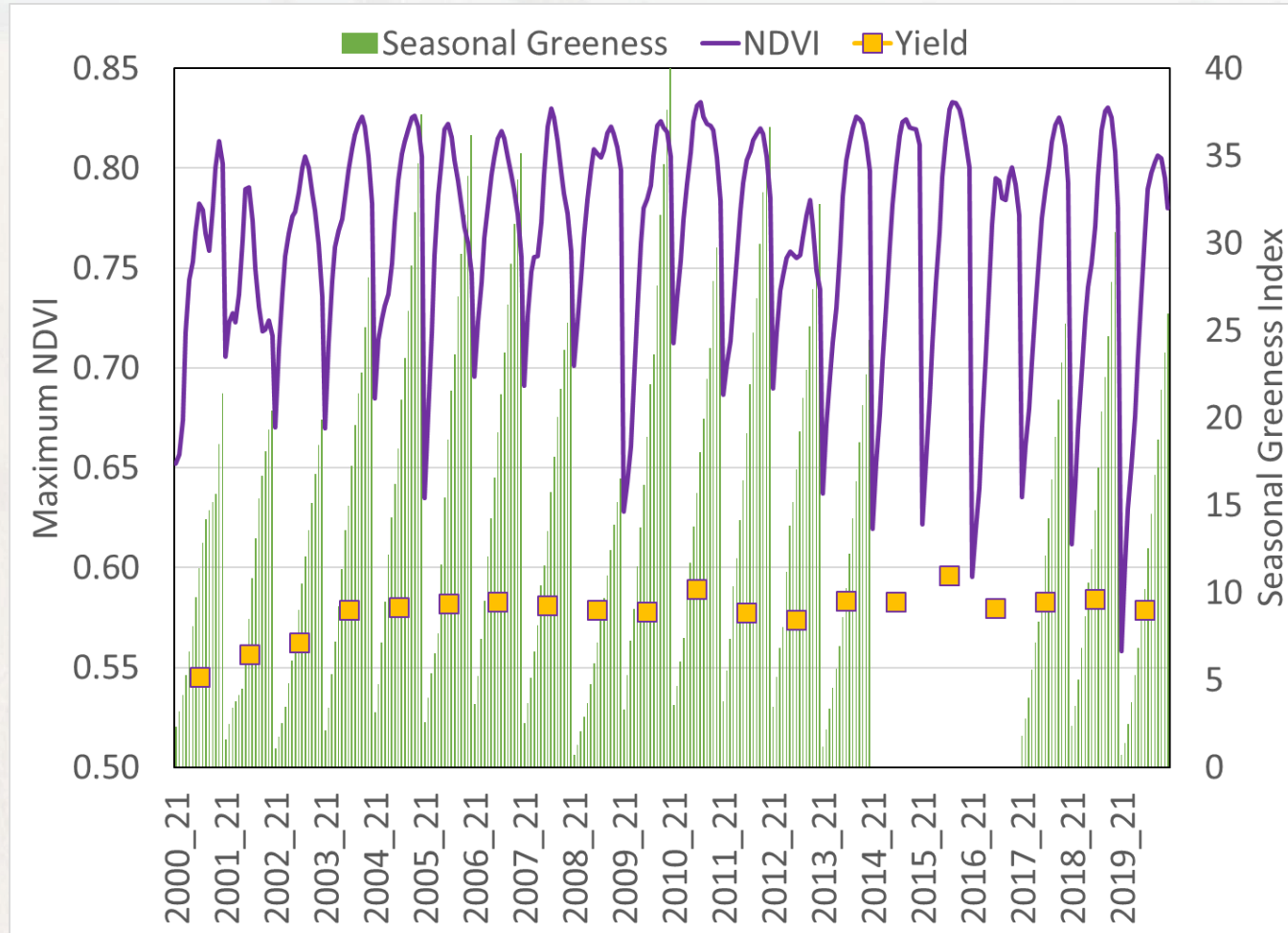
May 11, 2018

July 7, 2018

August 24, 2018

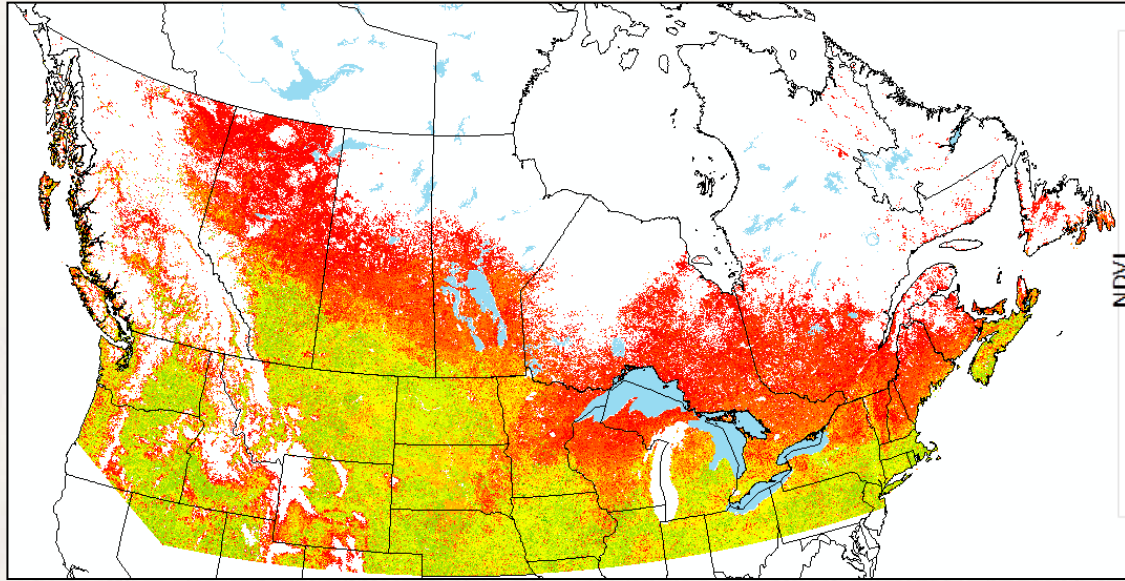
Leaf Area Index from Sentinel-2

Seasonal Greenness for Corn in Eastern Ontario

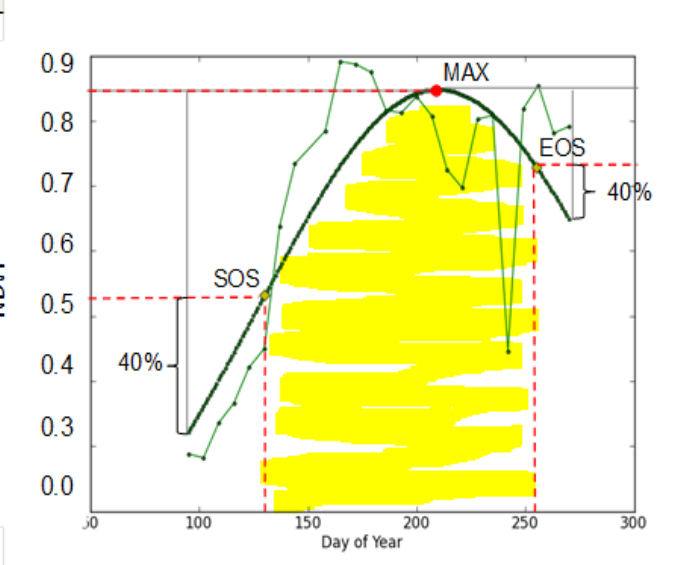


Seasonal Greenness shows potential as it is less cross correlated with other variables in capturing vegetation condition that leads to reductions in crop yield

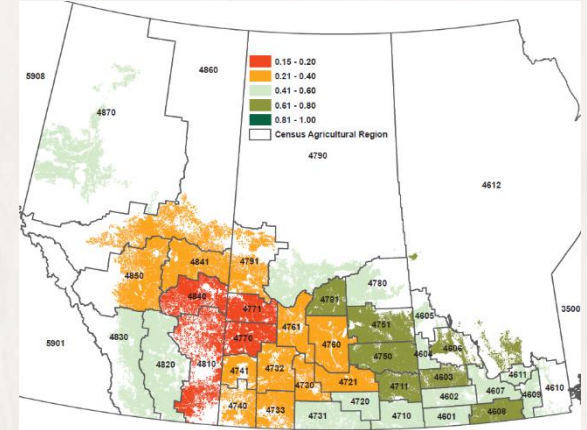
Estimating Growth Stage



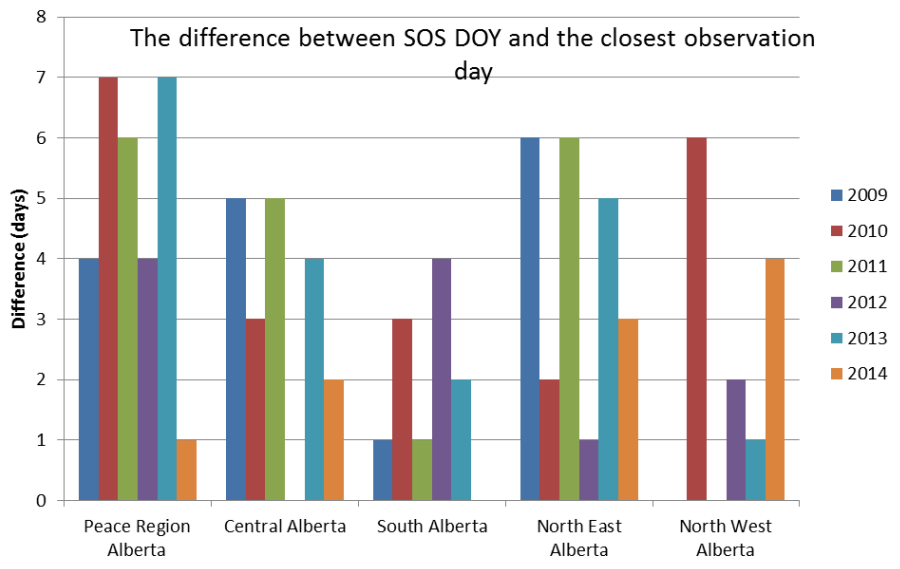
Start of Season estimate from MODIS NDVI Data



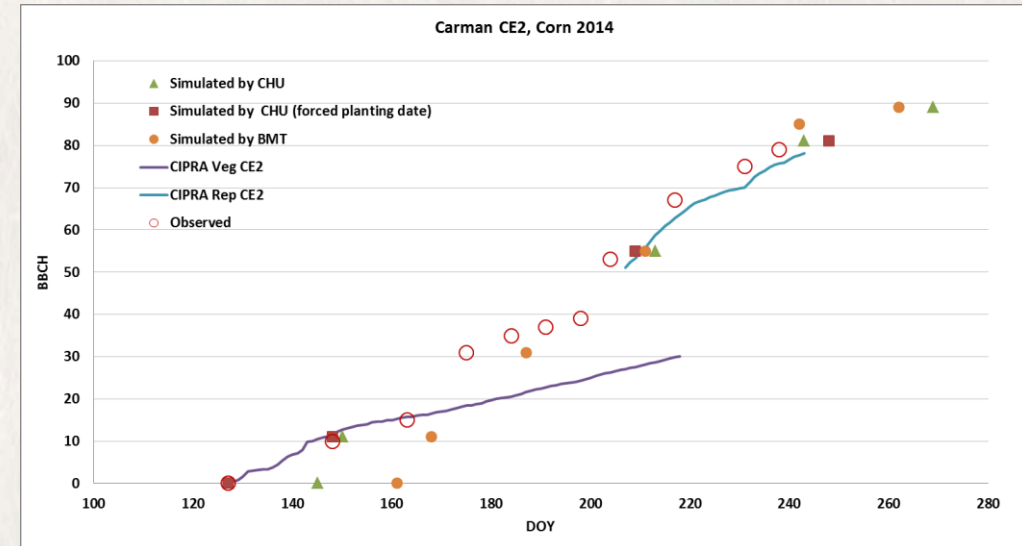
5 Year Correlation Between NDVI and BMT Start of Season



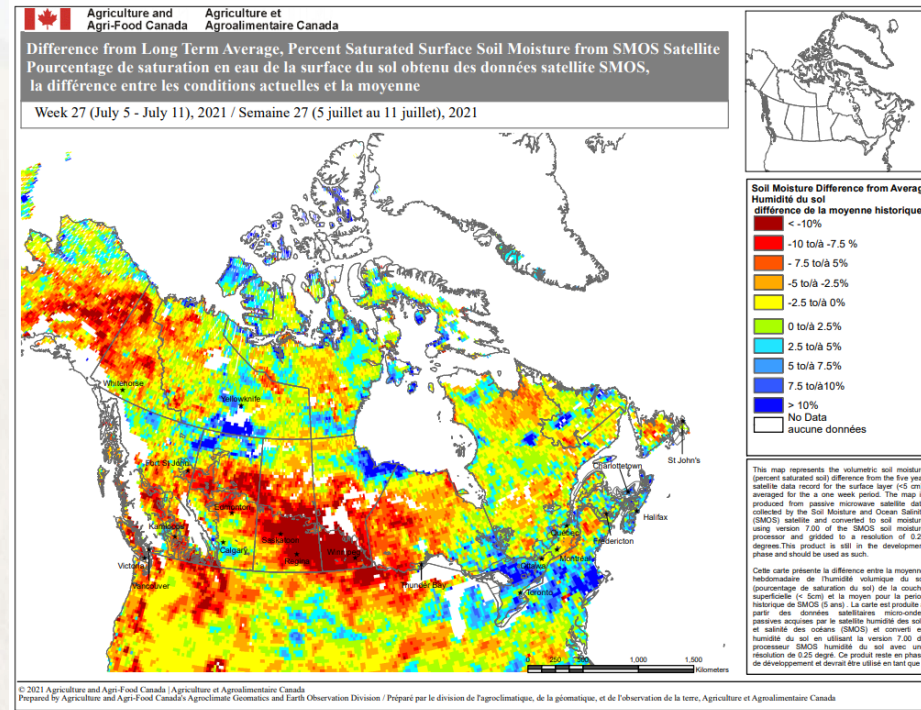
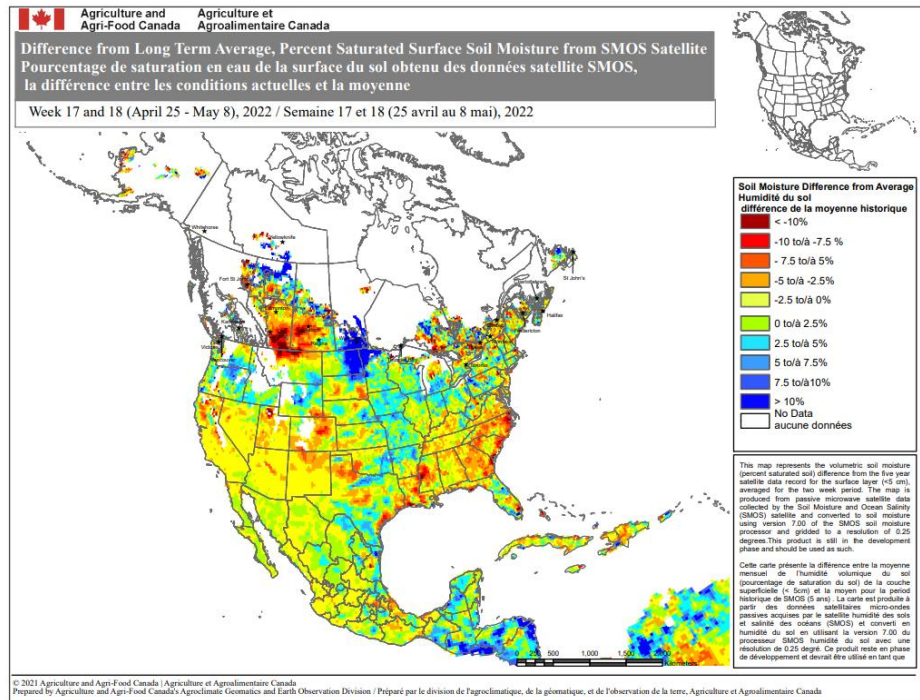
Vegetative Growth Stages from Bio-climatological models



Most regions the seeding was estimated within 8 days or less by NDVI over a multi year period for Prairie Provinces



Satellite Soil Moisture from Passive Microwave



- From Soil Moisture Ocean Salinity Mission (European Space Agency), launched 2009
- Difference from average at weekly, biweekly, monthly accumulation periods
- Detects surface soil moisture but this corresponds with long term moisture deficits when accumulated over longer time periods

Emergence of Satellite Soil Moisture Data Sets

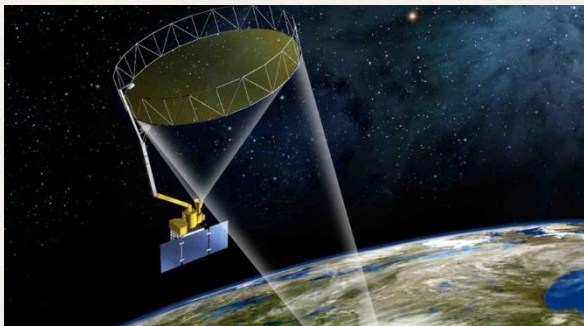


Soil Moisture and Ocean Salinity Mission (SMOS)

European Space Agency

Launched November 2009

L-Band Microwave Radiometer

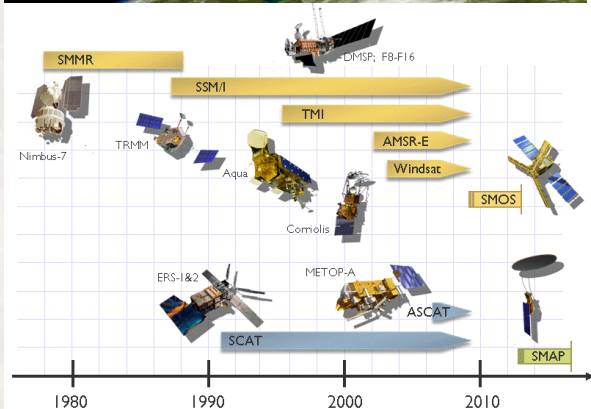


Soil Moisture Active Passive Mission (SMAP)

NASA

Launched January 2015

L-Band Microwave Radiometer / *L-Band RADAR*

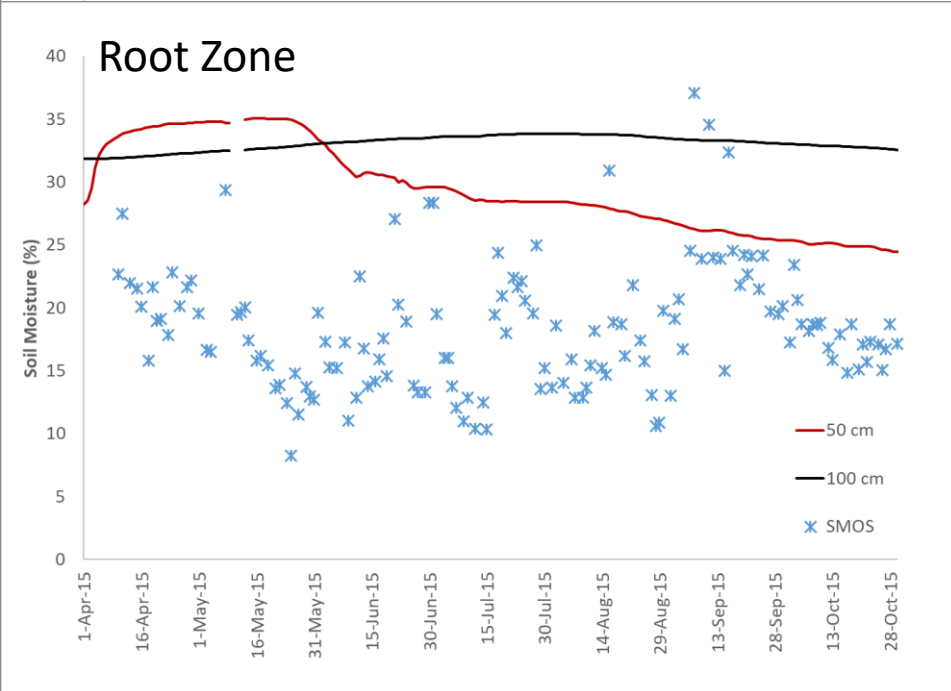
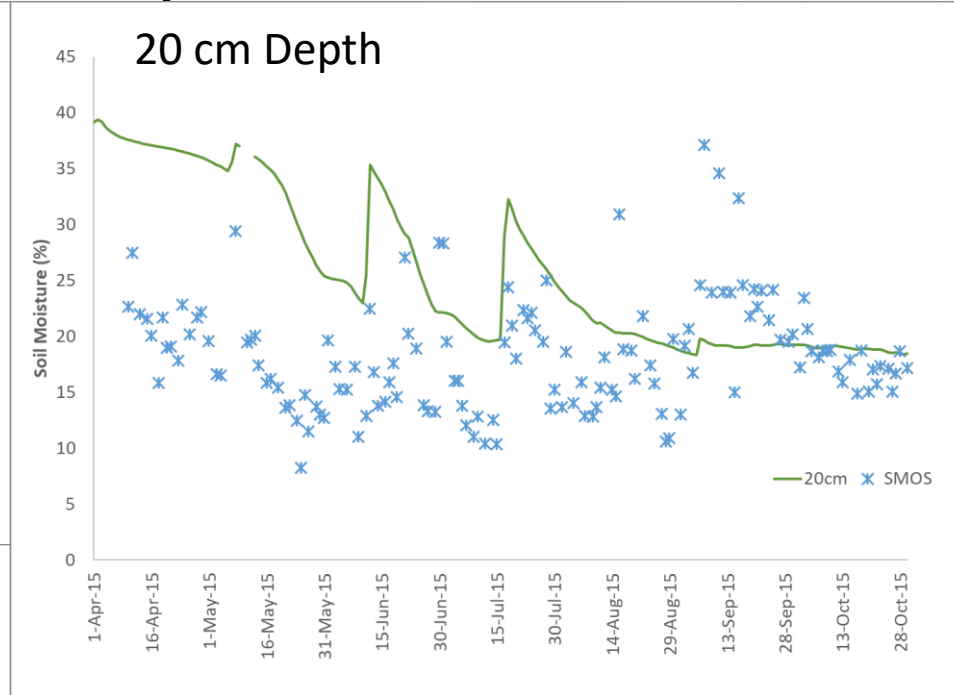
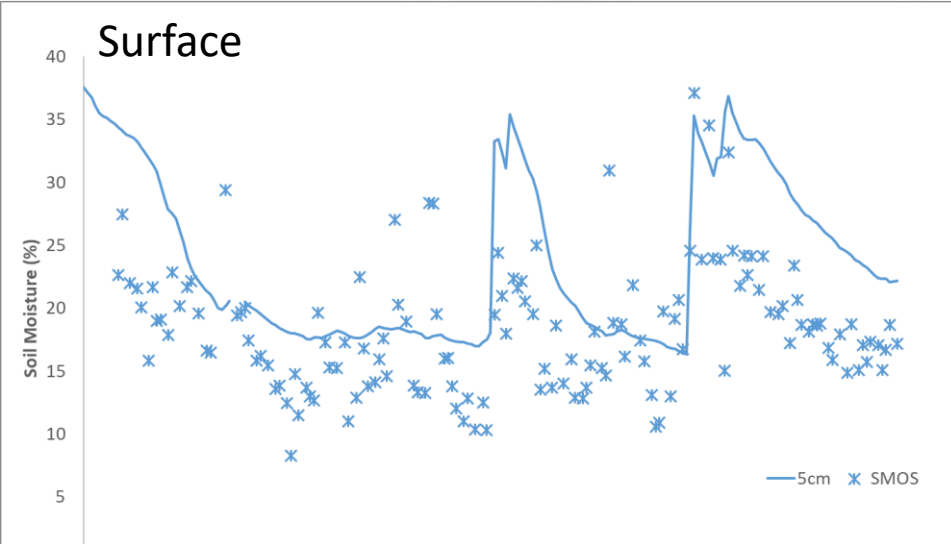


European Space Agency Climate Change Initiative (ESA-CCI)

Blend of Active/Passive Microwave soil moisture

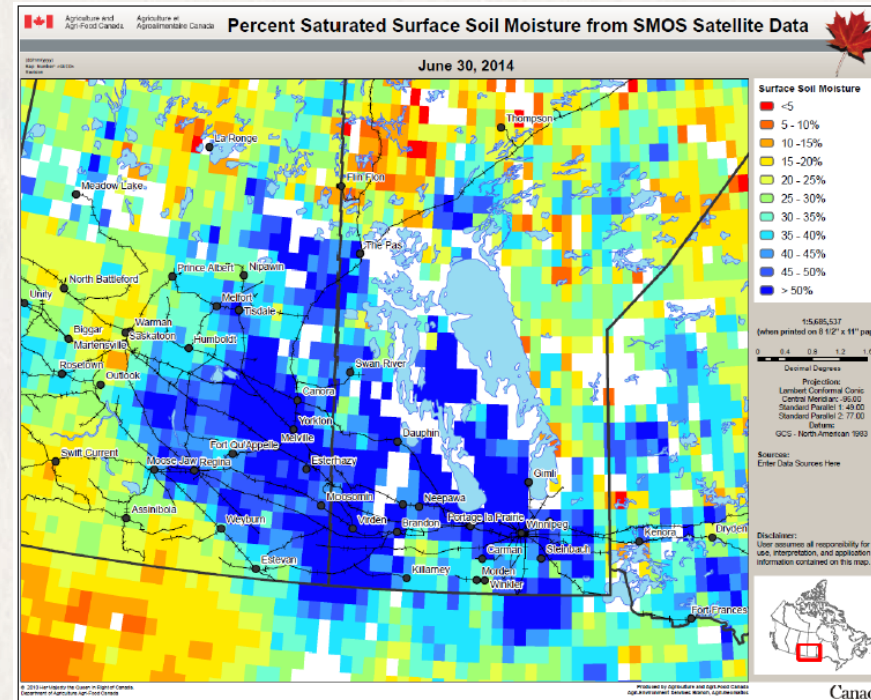
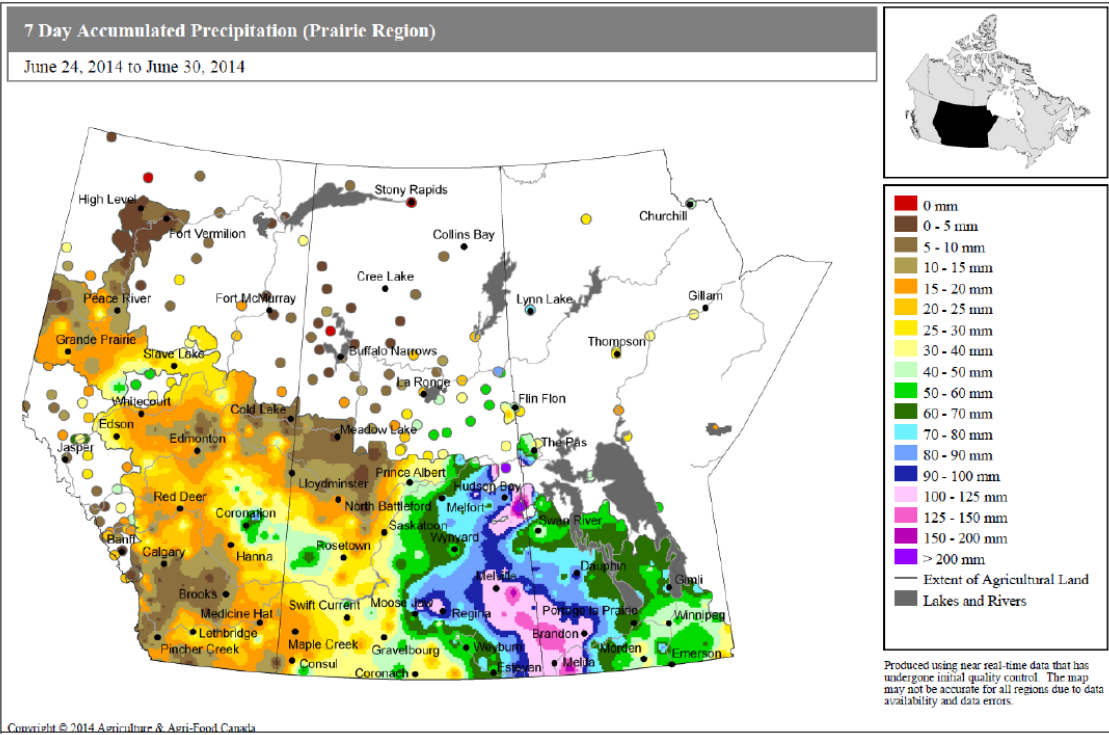
1979-2017 (**1992-2018**)

Surface Soil Moisture Dynamics: Satellite vs In-Situ



- Satellites capture moisture in top 5 cm with reasonably high accuracy; wetting and drying trends captured well
- Can't directly infer water storage from surface measurements
- Use satellite soil moisture as a proxy for profile soil moisture dynamics

Climate Related Production Risk Monitoring

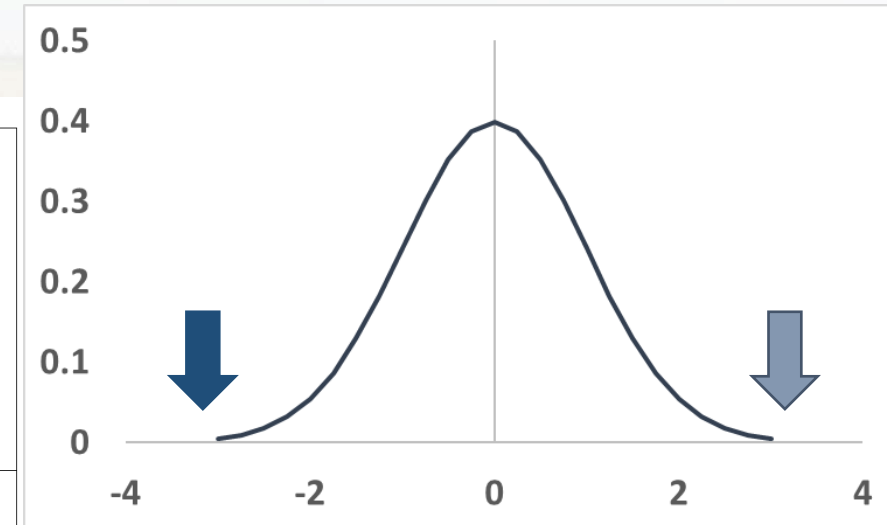
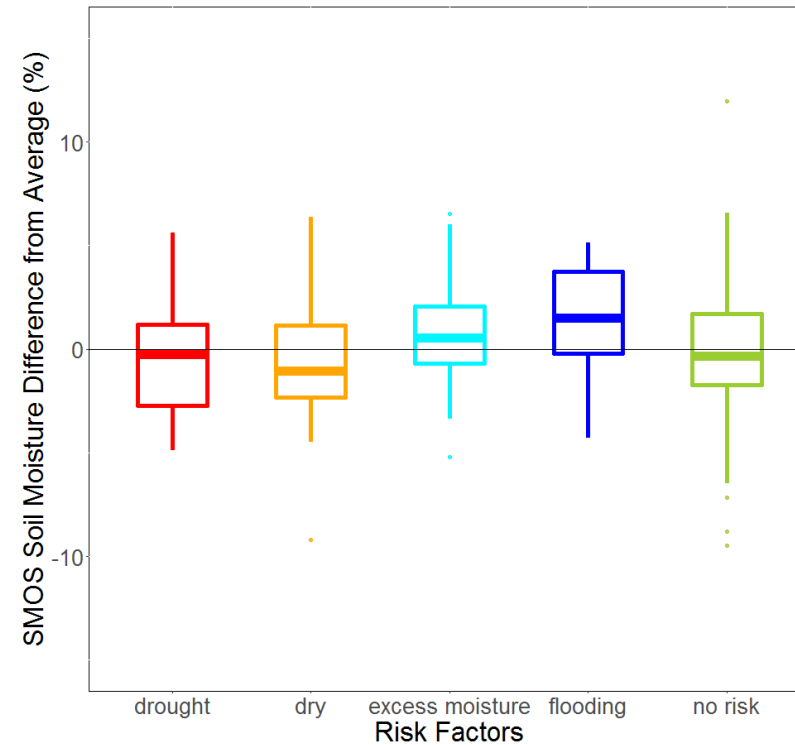
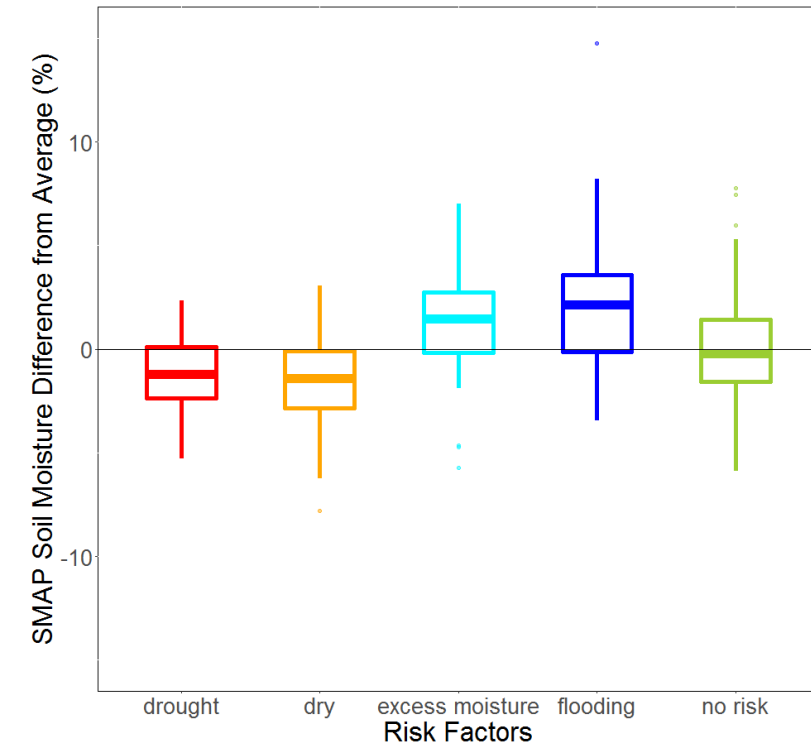


| | BC | AB | SK | MB | ON | QC | ATL |
|----------------------------------|--------------------|-------------------------------|-------------------------------|-------------------------------|-----------------------|--------------------|-------------------------------|
| Two-week forecast | stable | stable | improving | improving | worsening | stable | improving |
| July 8, 2014 | low water supplies | drought | flooding | flooding | heat and lack of rain | | rain, high winds |
| Last Report June 24, 2014 | low water supplies | flooding | wet, low temps, unseeded | wet, unseeded | | | |
| June 10, 2014 | low water supplies | | wet | wet, delayed operations | | | |
| May 27, 2014 | | low soil moisture | | wet, delayed operations | | | low temps, delayed operations |
| May 13, 2014 | | low temps, delayed operations | low temps, delayed operations | low temps, delayed operations | delayed operations | delayed operations | delayed operations |
| Apr 29, 2014 | | low soil moisture | low temps, low soil moisture | low temps, flooding | delayed operations | delayed operations | flooding, delayed operations |

Quantifying Climate Related Risk

SMAP

SMOS



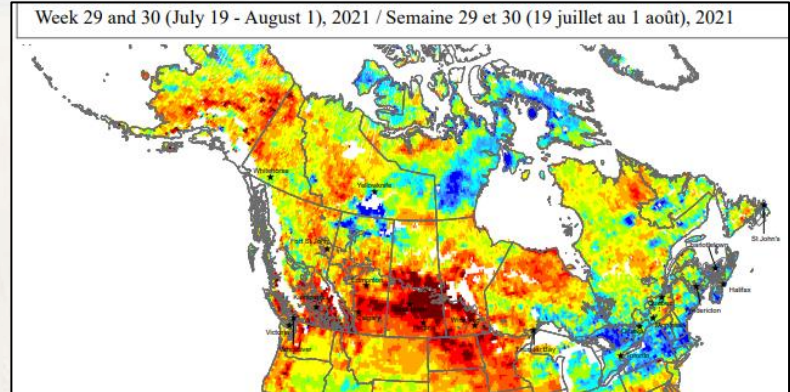
- Assessment of Extreme Events requires context
- Short record satellite data sets often have insufficient data to assess “normal” conditions

Champagne, C.; Zhang, Y.; Cherneski, P.; Hadwen, T. Estimating Regional Scale Hydroclimatic Risk Conditions from the Soil Moisture Active-Passive (SMAP) Satellite. *Geosciences* 2018, 8, 127.

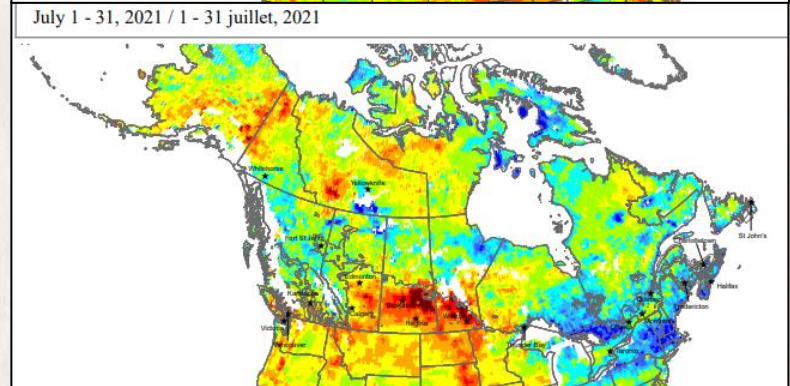
Drought from L-Band Passive Microwave

- Surface soil moisture is very responsive to fluctuating temperature/precipitation conditions and less indicative of long term water storage
- Assimilation of satellite soil moisture into land surface models can capture root zone soil moisture but these values are heavily dependent on other model parameters (primarily soil water holding capacity) which are not well characterized in many areas.
- Analysing soil moisture anomalies over different time scales can better capture agriculturally relevant water deficits (timing, intensity, persistence)

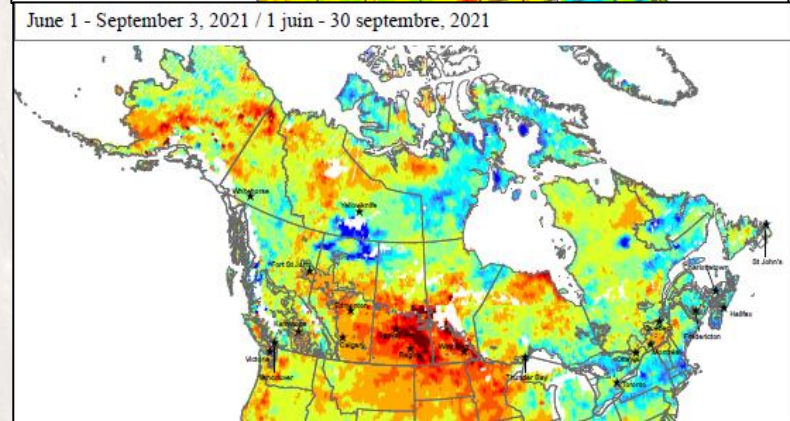
Two Week Anomaly



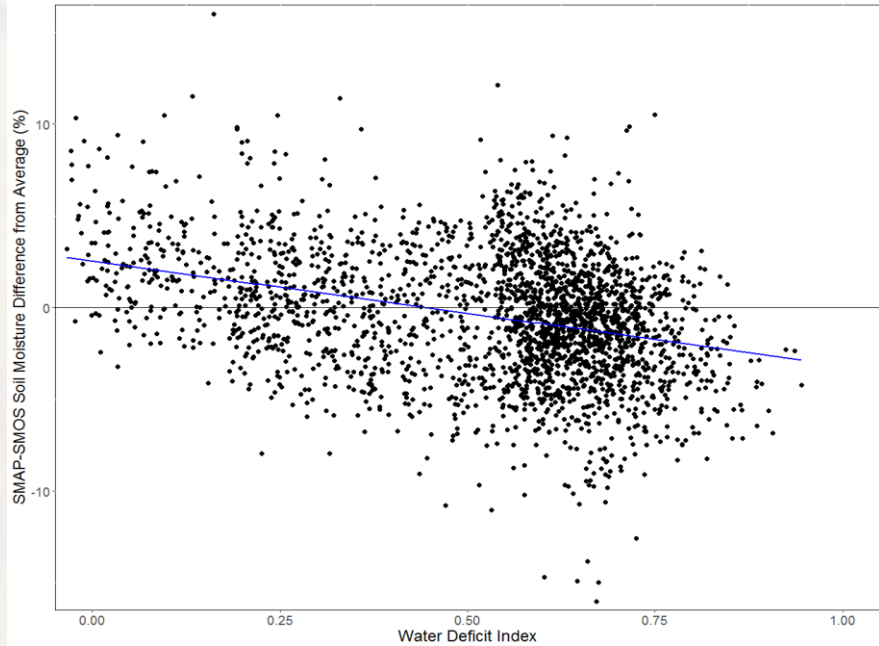
Monthly Anomaly



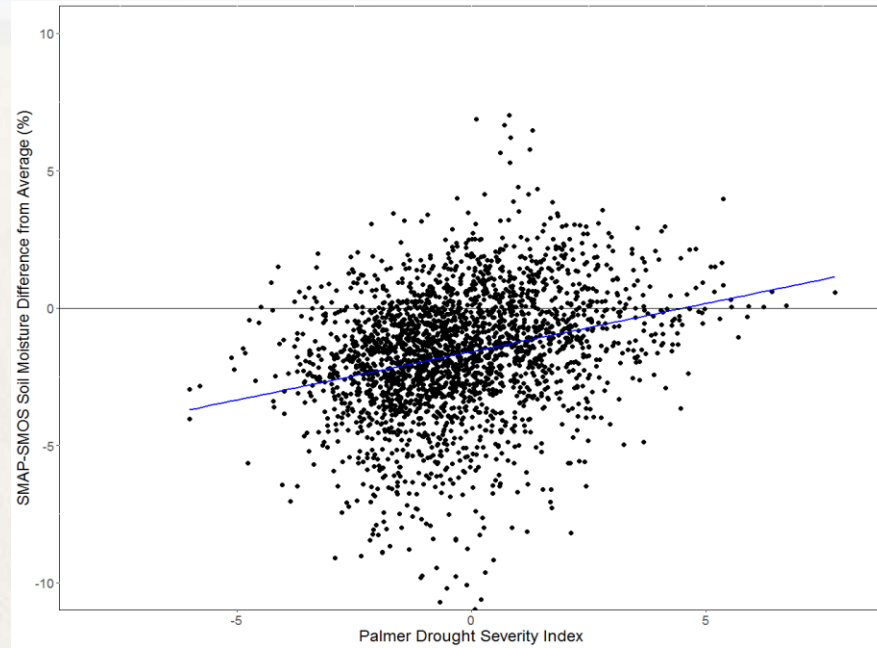
Seasonal Anomaly



Scale and Severity of Drought Impacts Observed by Satellite Data

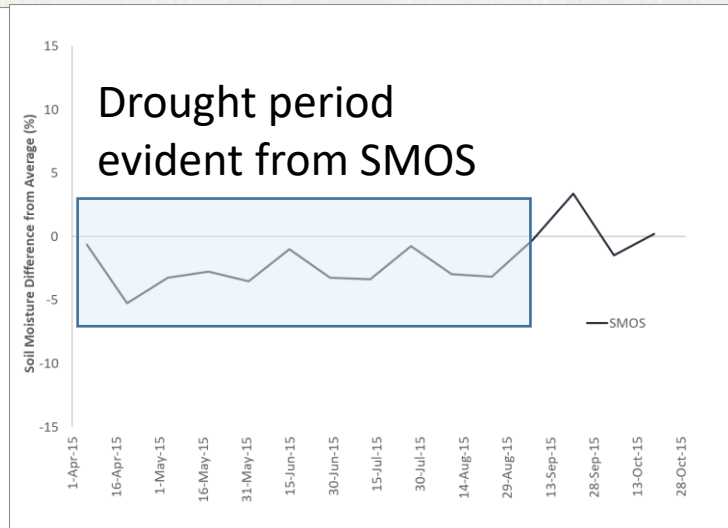
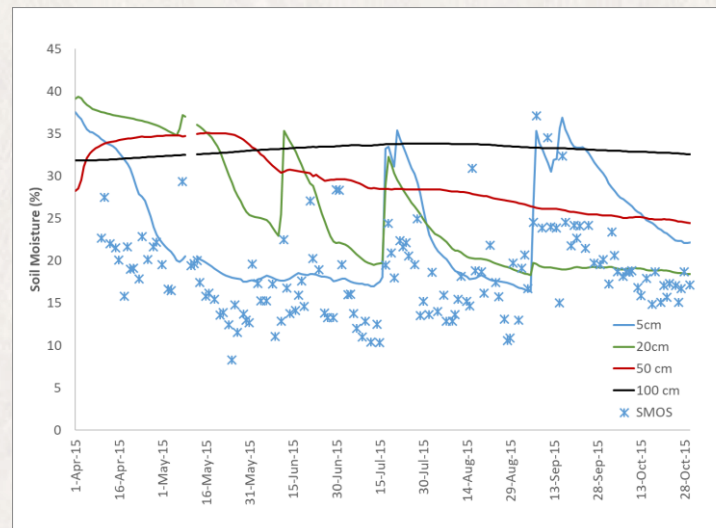


Accumulating Soil Moisture Anomaly vs Palmer Drought Severity Index



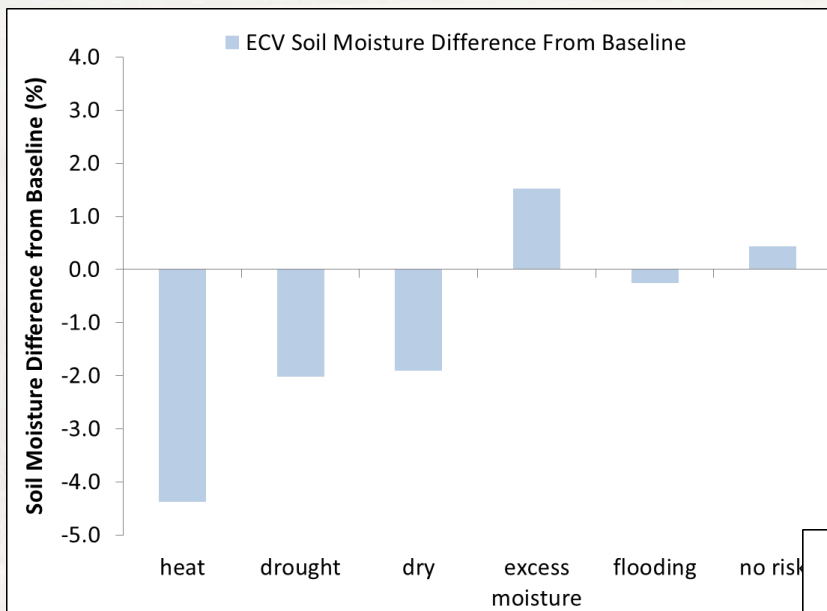
Soil Moisture Anomaly vs Crop Water Demand Index

Satellite surface soil moisture most sensitive to drought processes driven by evaporative demand than long term soil moisture shortages



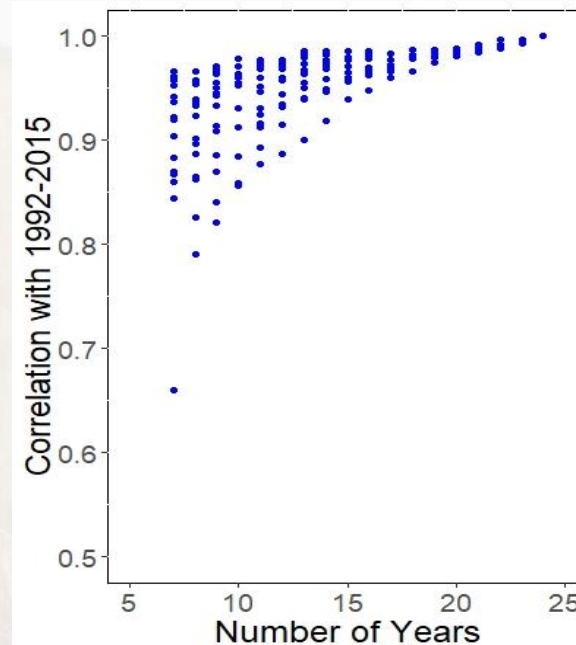
How Sensitive to Drought Conditions are Satellite Soil Moisture Anomalies?

ESA – CCI Long Term
Satellite Soil Moisture
Data Set



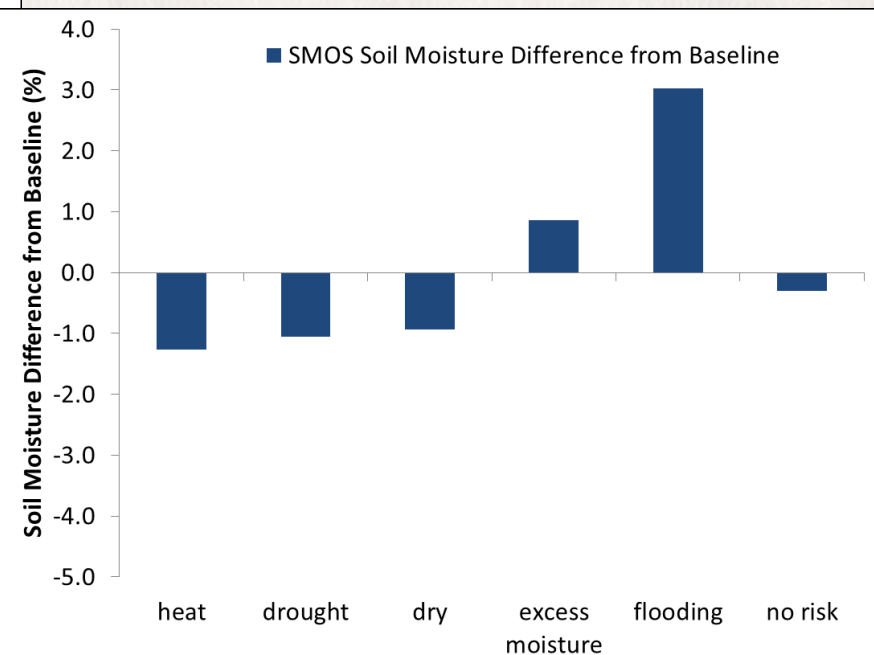
Length of baseline or normalizing period is important: Longer is better!

8 Year Baseline

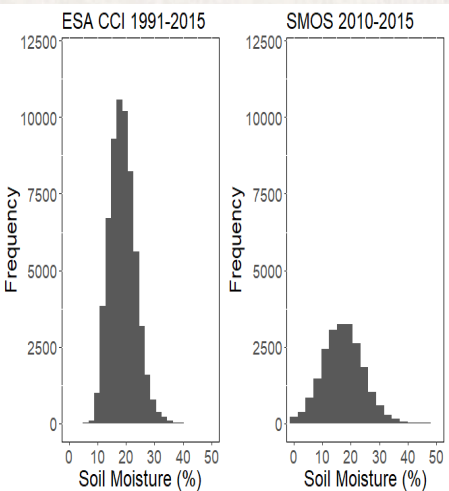


27 Year Baseline

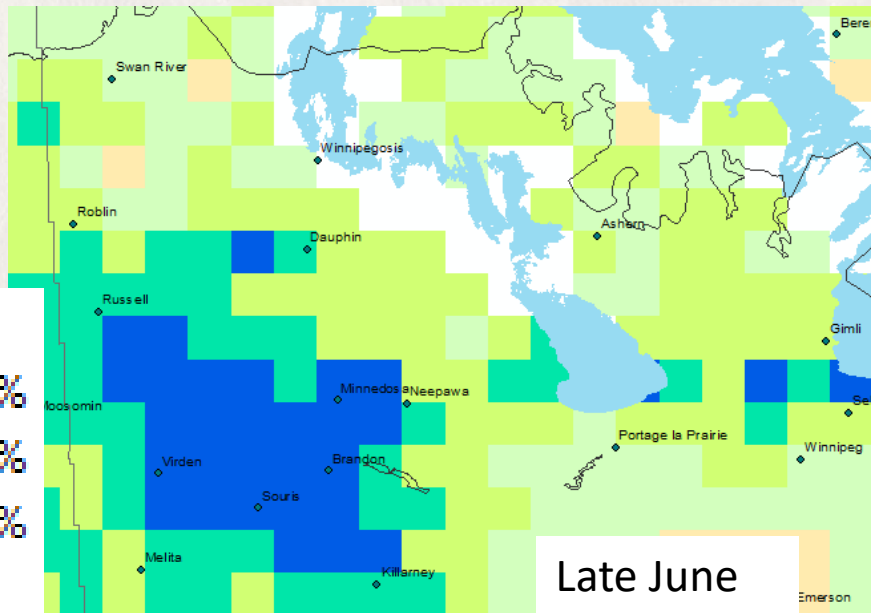
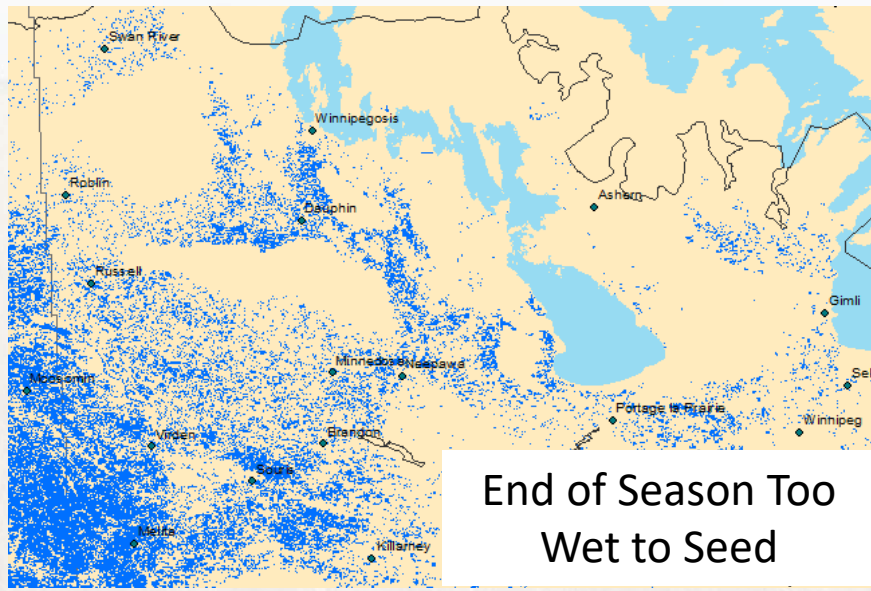
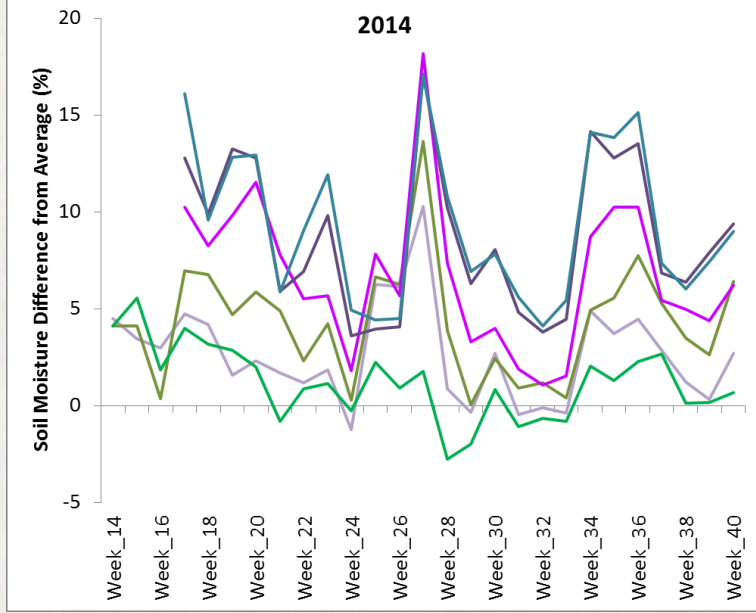
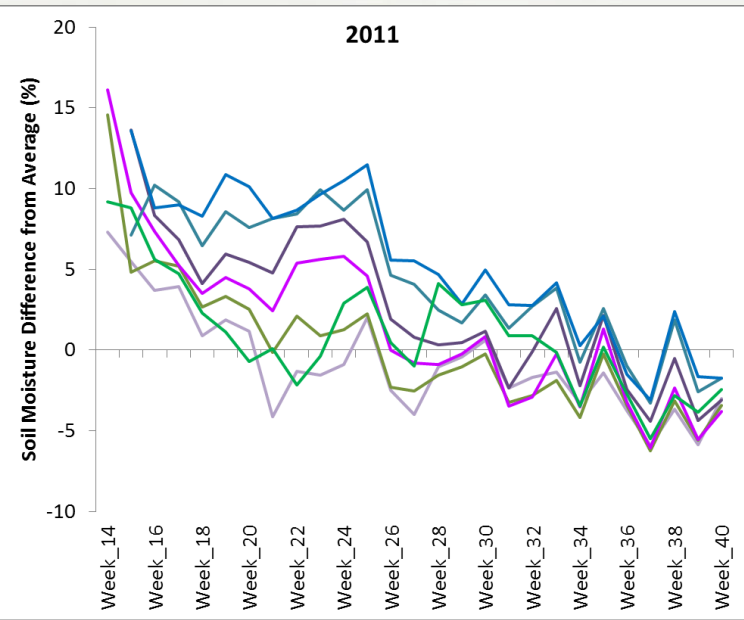
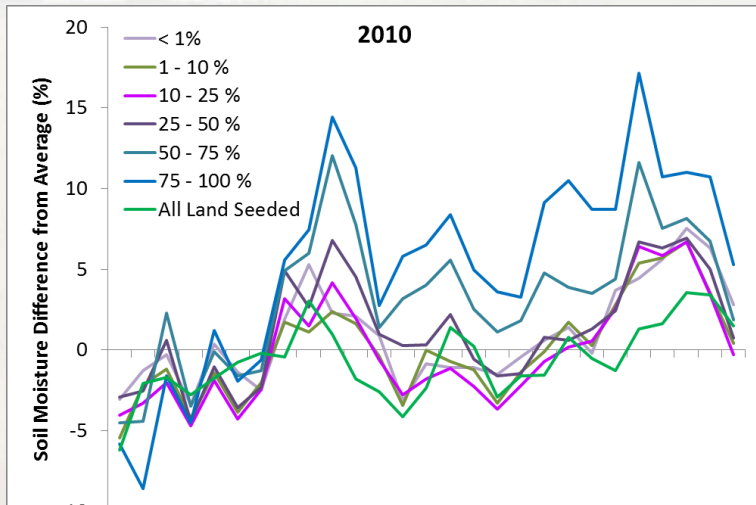
SMOS Soil Moisture
Data Set



Need 12 years or more to robustly determining low yielding years from average, longer for determining high yielding years (18 years)



SMOS Soil Moisture Evolution During Wet Growing Seasons

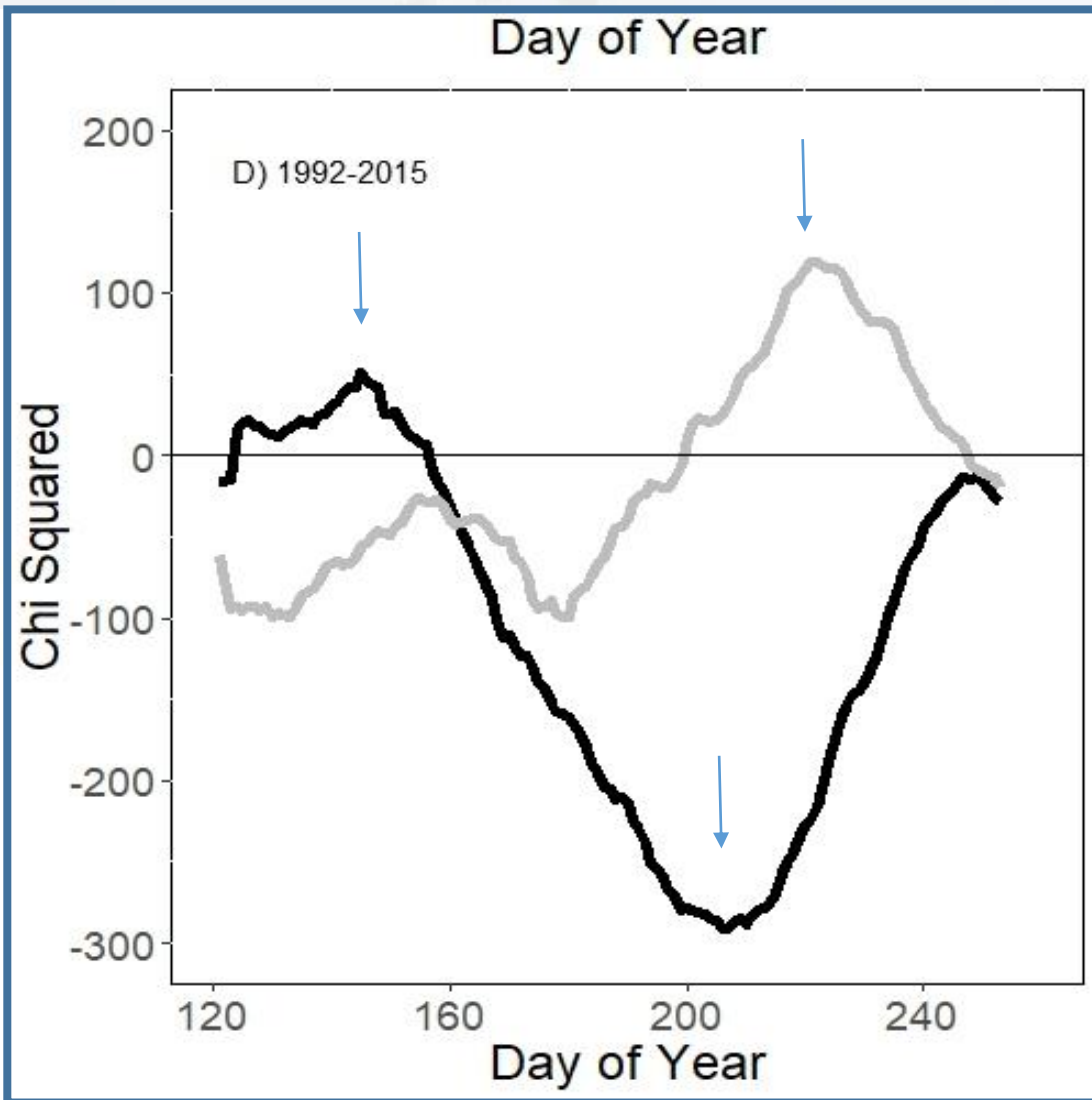


Satellite soil moisture conditions over areas classified as too wet to seed are variable from year to year

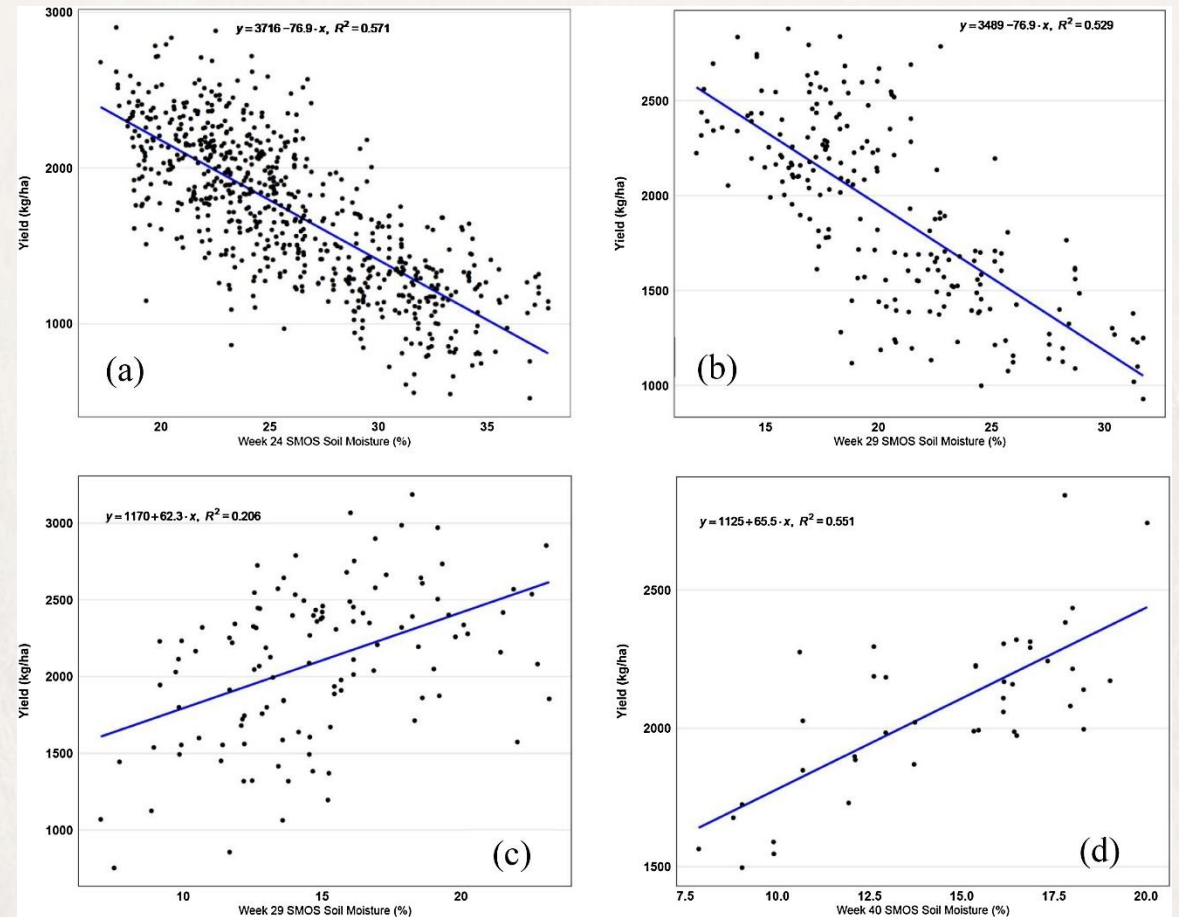
Regression tree model showed most predictive variable is Elevation, with slope, soil water holding capacity, land cover and ecoregion also

- < 25 %
- 25 - 50 %
- 50 - 75 %
- 75 - 90 %
- > 90 %

Sensitivity of Crop Yield to Soil Moisture



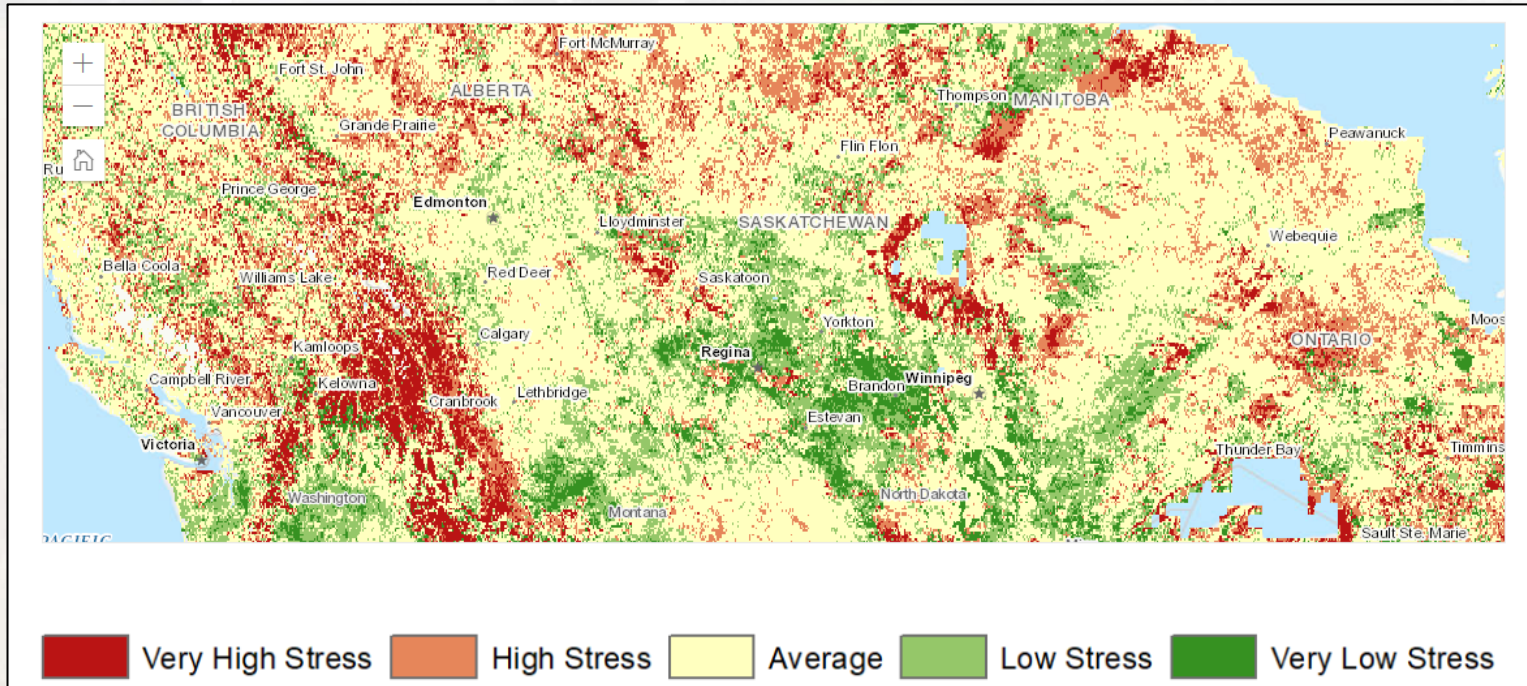
Low yields associated with high spring soil moisture (excess) and low soil moisture in July (drought)



Champagne, C., White, J., Berg, A., Belair, S., Carrera, M. (2019). Impact of soil moisture data characteristics on the sensitivity to crop yields under drought and excess moisture conditions, 11(4), <http://dx.doi.org/10.3390/rs11040372>

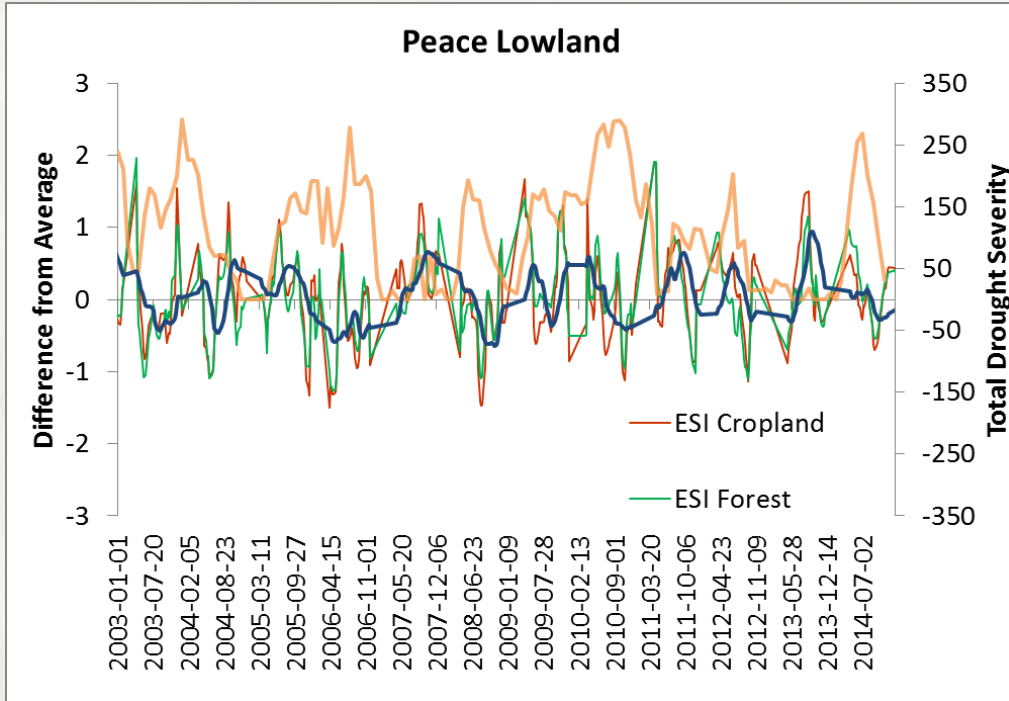
White, J., Berg, A.A., Champagne, C., Warland, J., Zhang, Y. (2019). Canola yield sensitivity to climate indicators and passive microwave-derived soil moisture estimates in Saskatchewan, Canada, 268 354-362. <http://dx.doi.org/10.1016/j.agrformet.2019.01.004>

Evaporative Stress Index (ESI)

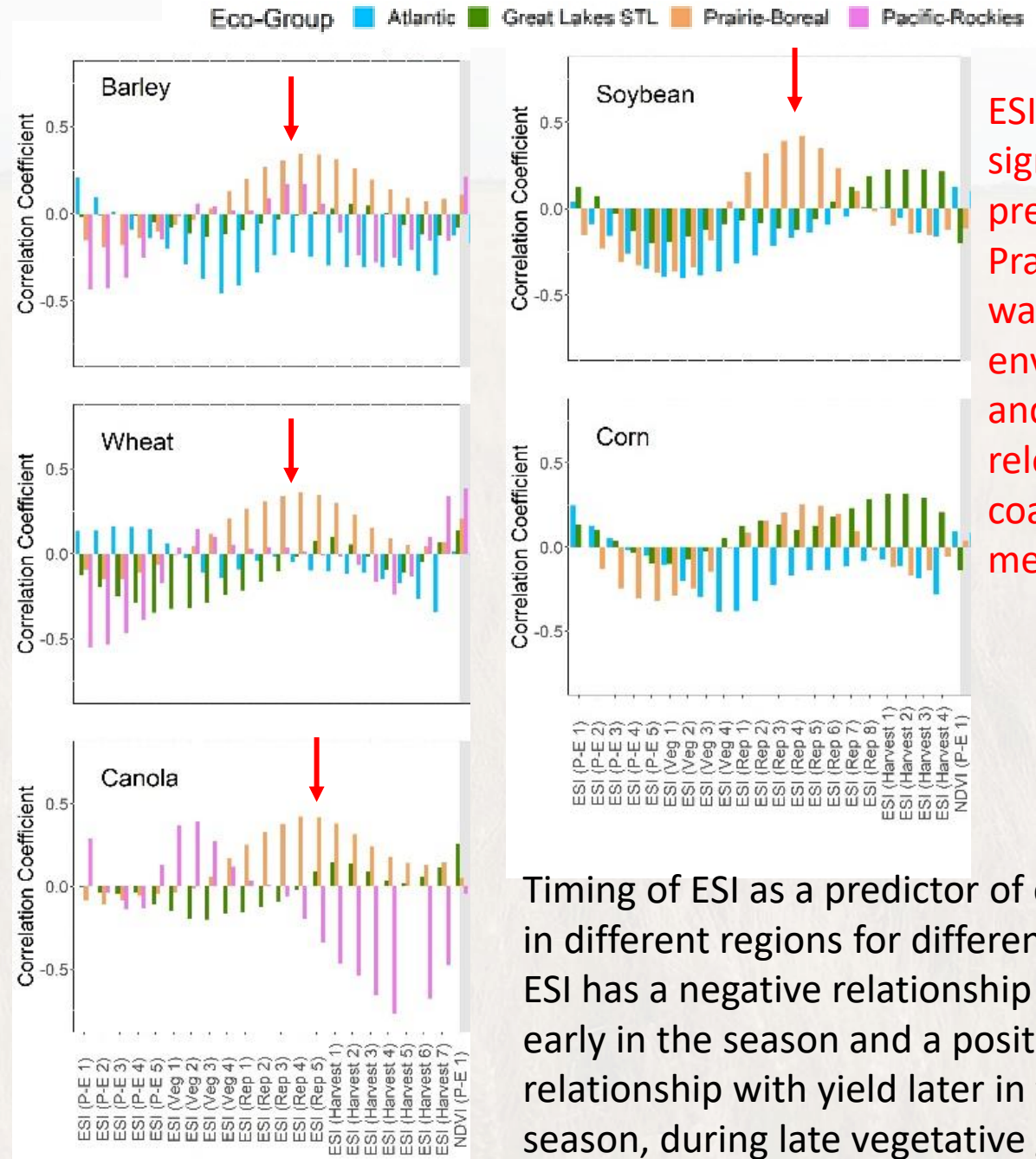


- Based on thermal/optical data from MODIS sensor quantifying rates of evapotranspiration at surface
- Data produced by NASA and distributed through SERVIR global
- *Mapping daily evapotranspiration at field to continental scales using geostationary and polar orbiting satellite imagery. Hydrology and Earth System Sciences, 15, 223-239 (2011). Anderson, M.C., Kustas, W.P., Norman, J.M., Hain, C.R., Mecikalski, J.R., Schultz, L., González-Dugo, M.P., Cammalleri, C., D'Urso, G., Pimstein, A., & Gao, F.*

Application of ESI



- Evaporative stress fluctuates in time with lows when high temperatures and low rainfall lead to dry periods; clearer trends are illustrated with temporal moving average
- Temporal trend shows higher evaporative stress during drought periods; in some cases high peaks in evaporative stress leading into more severe drought periods

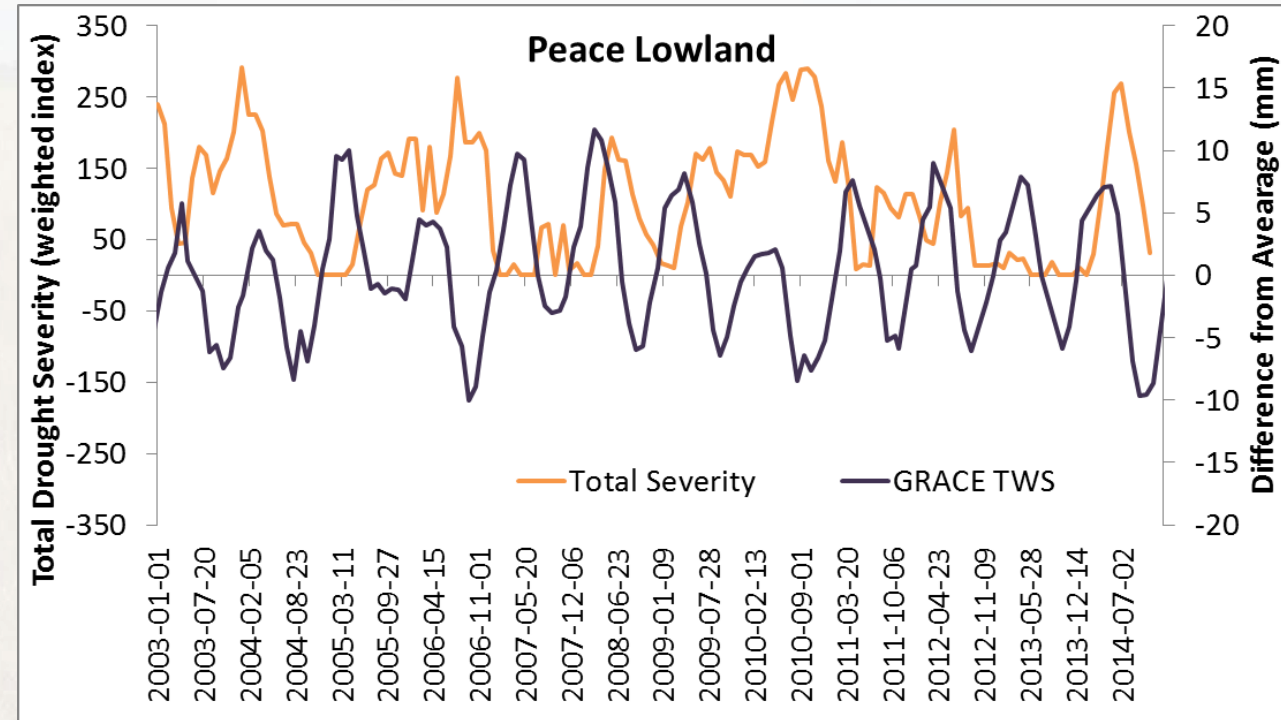
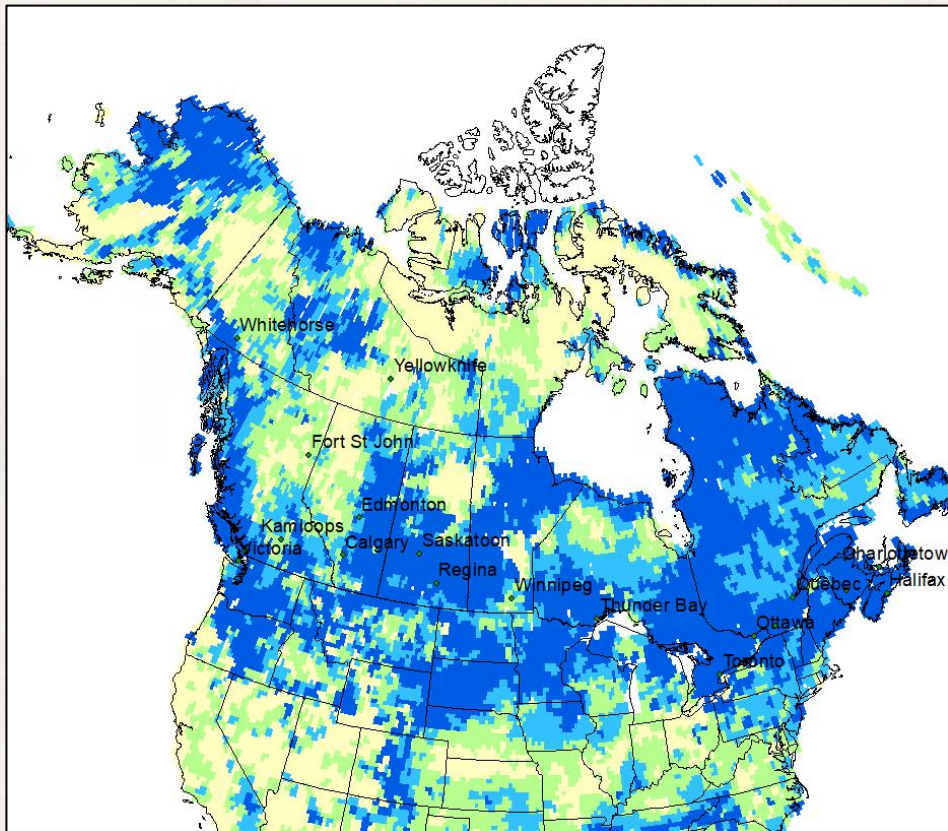


ESI is significant predictor in Prairies (more water limited environment and less relevant in coastal/maritime regions)

Timing of ESI as a predictor of crop yield in different regions for different crops. ESI has a negative relationship with yield early in the season and a positive relationship with yield later in the season, during late vegetative growth and seed reproduction.

GRACE: Gravity Recovery and Climate Experiment

- Total Water Storage Data set from NASA



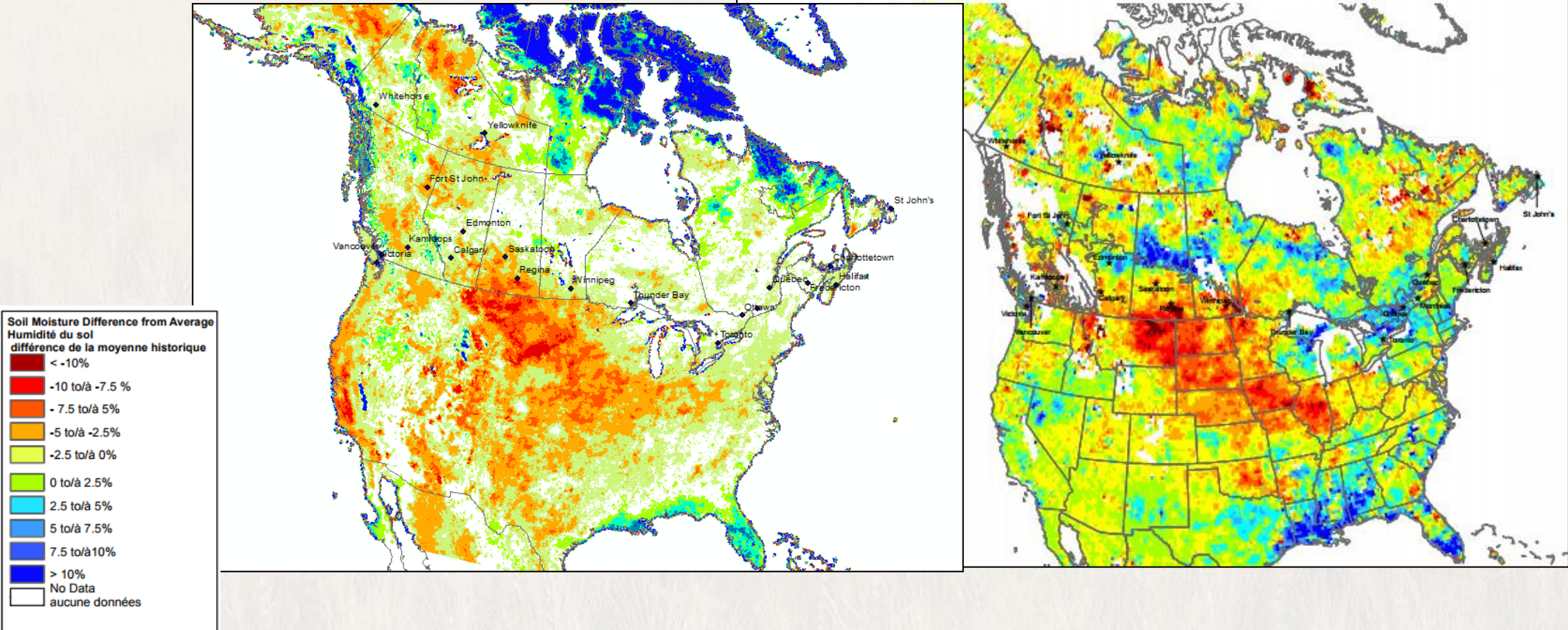
GRACE + GRACE FO has continuous data series from 2003 – present

Trends well with long term drought anomalies



Soil Moisture for Drought Monitoring

Calculation of Drought indicators using CMC soil moisture data



RDPS 10 cm

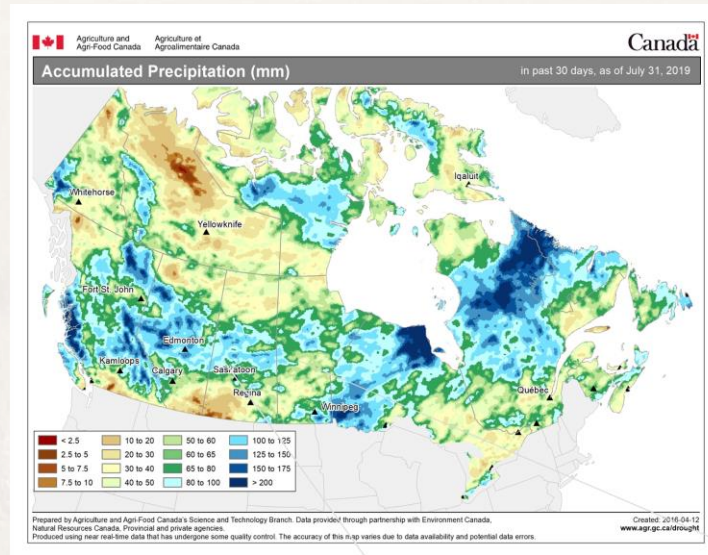
SMOS

Soil Moisture Difference from Average June 2017

System to calculate drought indicators – build portal to view data sets

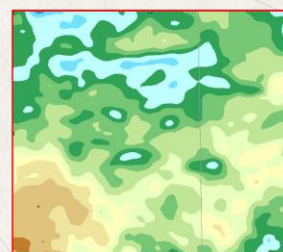
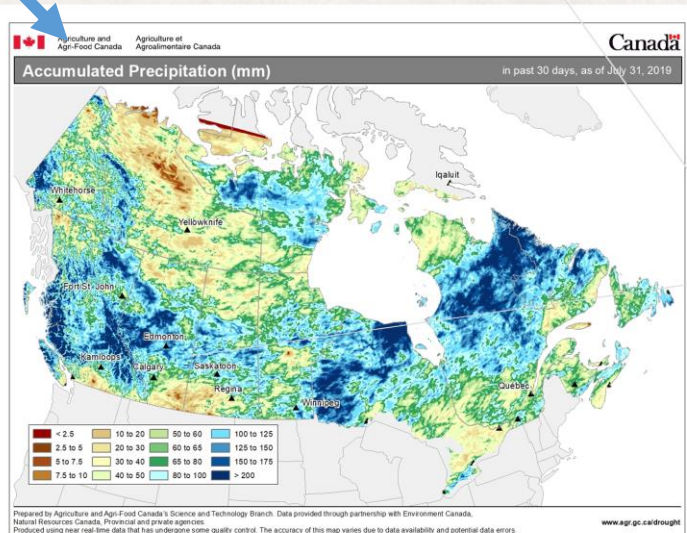
Numerical Weather Data: Near-term forecast, analysis and Re-Analysis Data

2.5 km High Resolution Deterministic Prediction System (HRDPS & HRDPA)

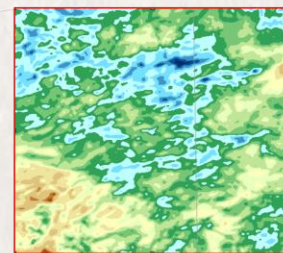


10 km Regional Deterministic Prediction System (RDPS & RDPA)

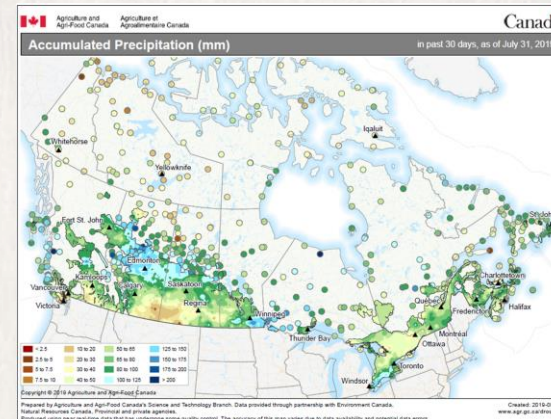
- Initial conditions and near term forecasts from ECCC Numerical Weather Prediction system combine Global Environmental Multiscale Model (GEM) with land surface prediction systems
- Canadian Precipitation Analysis (CaPA) produces an optimal estimate of precipitation using deterministic models incorporating in situ gauges and Doppler radar estimates



10km



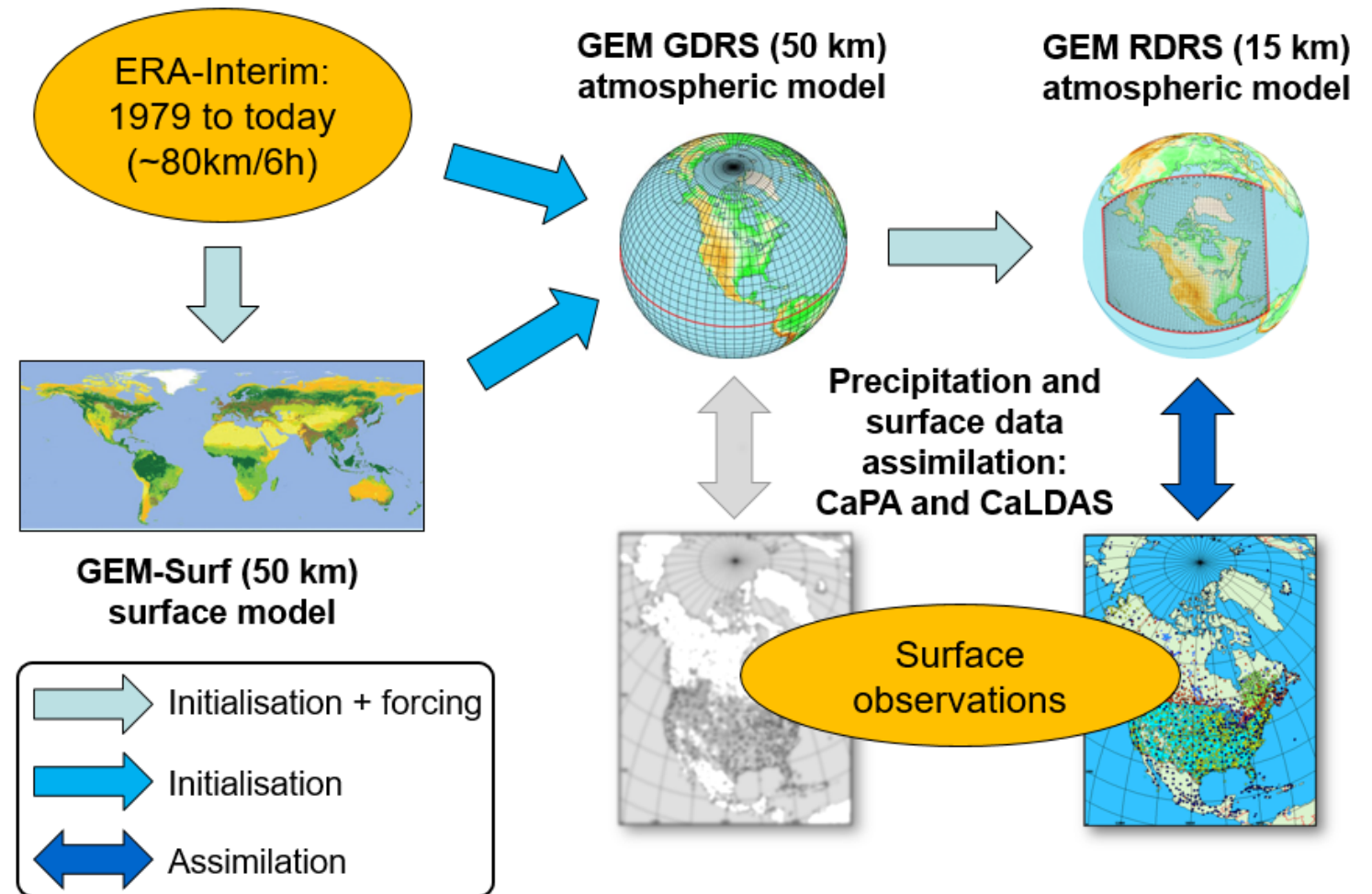
2.5km



Regional Reanalysis (RDRS)

- 39-Year Precipitation and Ground Surface Reanalysis
- Regional: RDRS
 - Modified REPS control member config. (cover Arctic Ocean)
 - GEM 4.8, LAM, 15-km
 - 12-h cycle/24-h reforecast
 - **Coupled with CaLDAS/CaPA**
- 1980-2018 Selected Variables Made Available and have been processed to 24 hour values
- Using as historical baseline for calculating drought indicators, agroclimate anomalies

Methodology

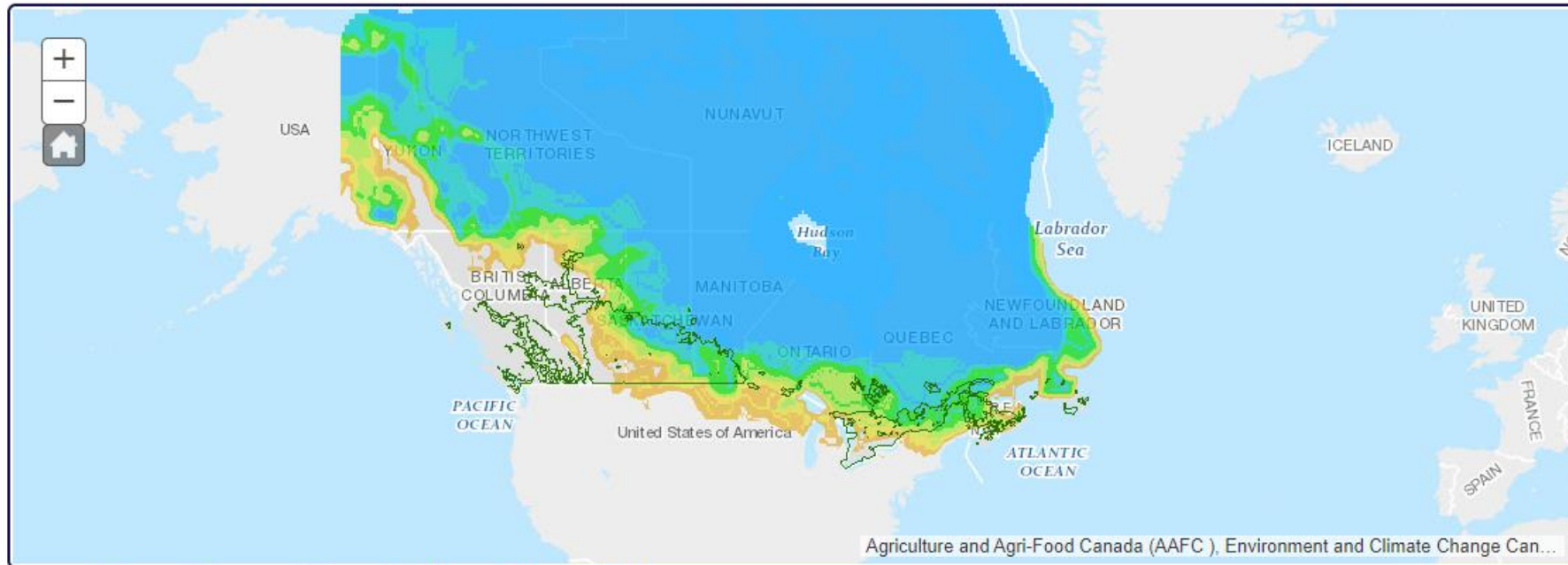


Agroclimate Forecasts

- Extreme Weather Indicators
 - Indicators of temperature, precipitation, heat & wind at forecasts from 1-4 weeks ahead (from Environment Canada 32 day forecasts), updated daily

Ice freeze days (herbaceous crops) in dormant period (<-15°C)

[Index maps for the National region \(PDF, ~2MB \)](#)



< 1 1 to 2 2 to 3 3 to 4 4 to 5 5 to 6 6 to 7 days

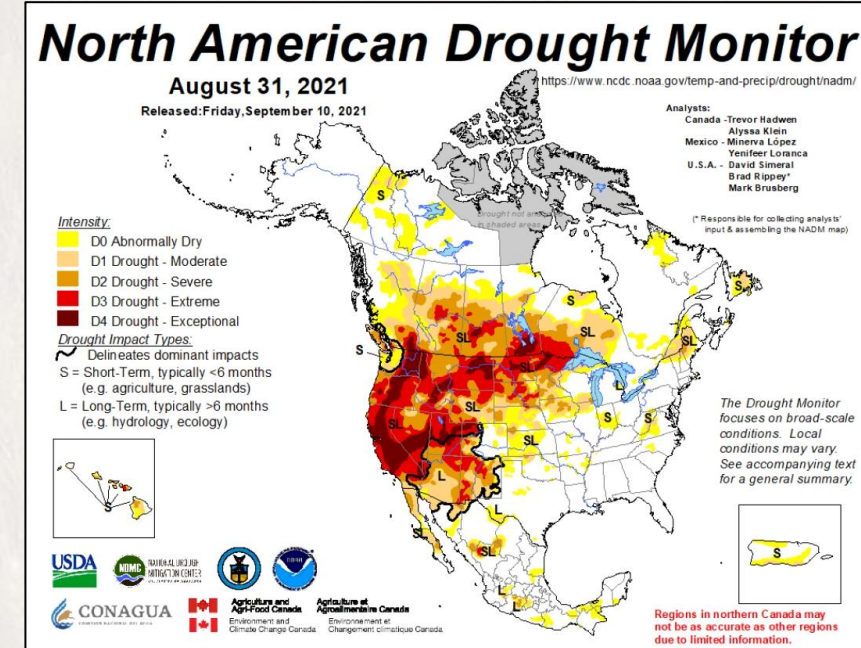
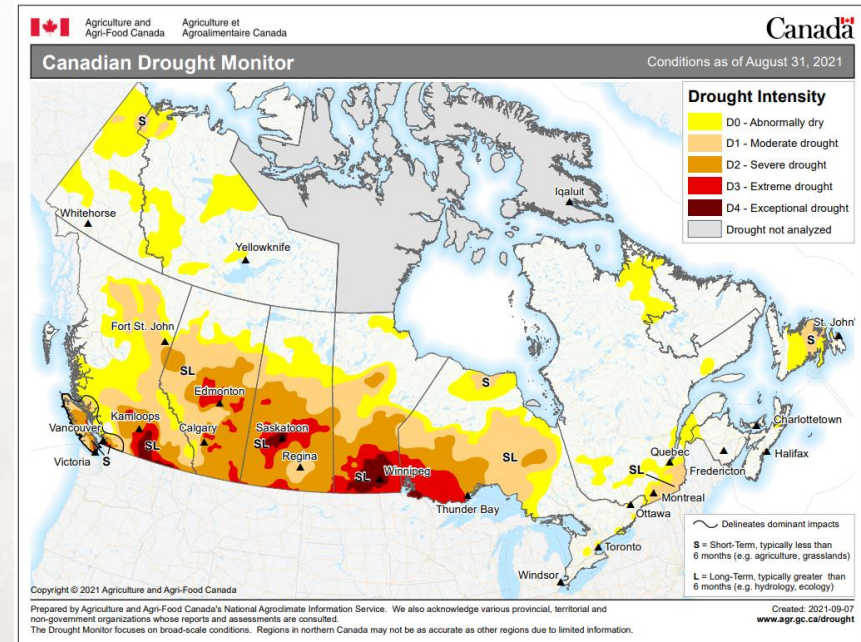
► Description - Legend

Drought Monitoring in Canada

- AAFC is the lead for drought monitoring in Canada. Drought indicators are key to this activity and research and development are ongoing.
- Process is collective through the North America Drought Monitor, ongoing since 2003.
- Uses a convergence of evidence approach to analyse drought severity based on percentiles; inputs include agroclimate indices calculated using Versatile Soil Moisture Budget Model (Palmer, SPEI, SPI), satellite based indicators, modelled precipitation analysis data, regional expert reports

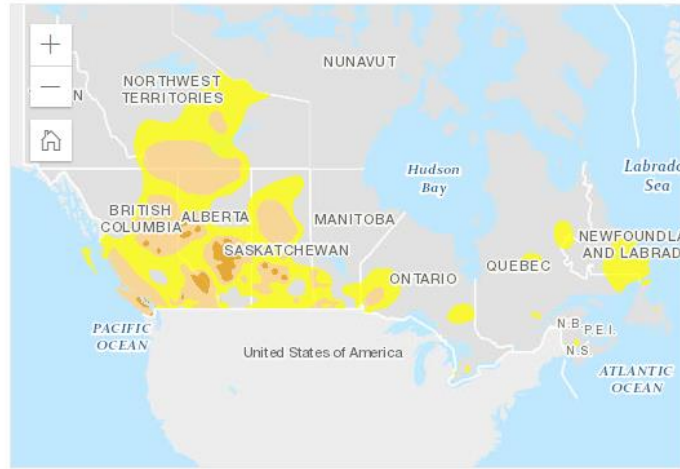
Drought categories are based on percentiles, which relate to the statistical return period.

| | |
|--------------------------|--------------------|
| D1- Moderate Drought | 1 in 5 year event |
| D2 – Severe Drought | 1 in 10 year event |
| D3 – Extreme Drought | 1 in 20 year event |
| D4 – Exceptional Drought | 1 in 50 year event |

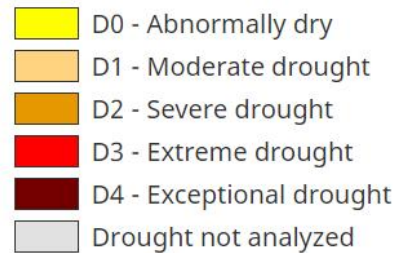


Drought Outlook Forecasts

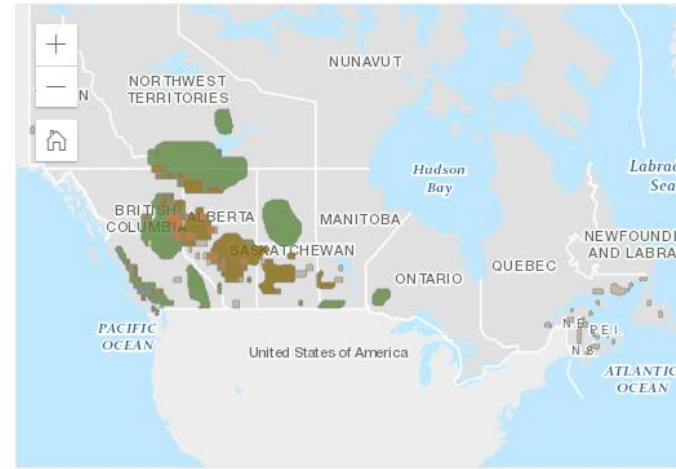
Drought conditions as of March 31, 2023



Legend for current drought conditions:



Drought Outlook for end of the following month

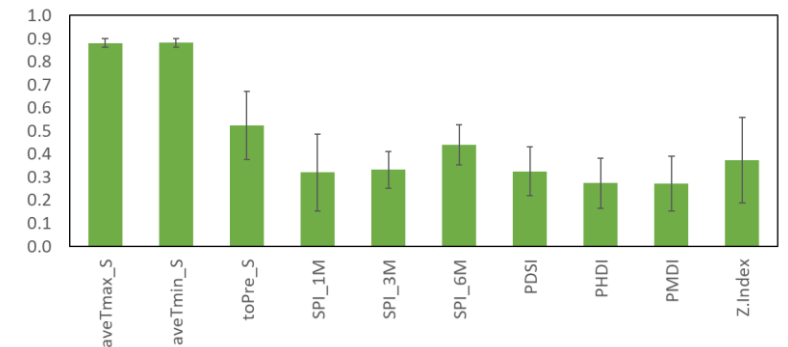


Legend for drought outlook:



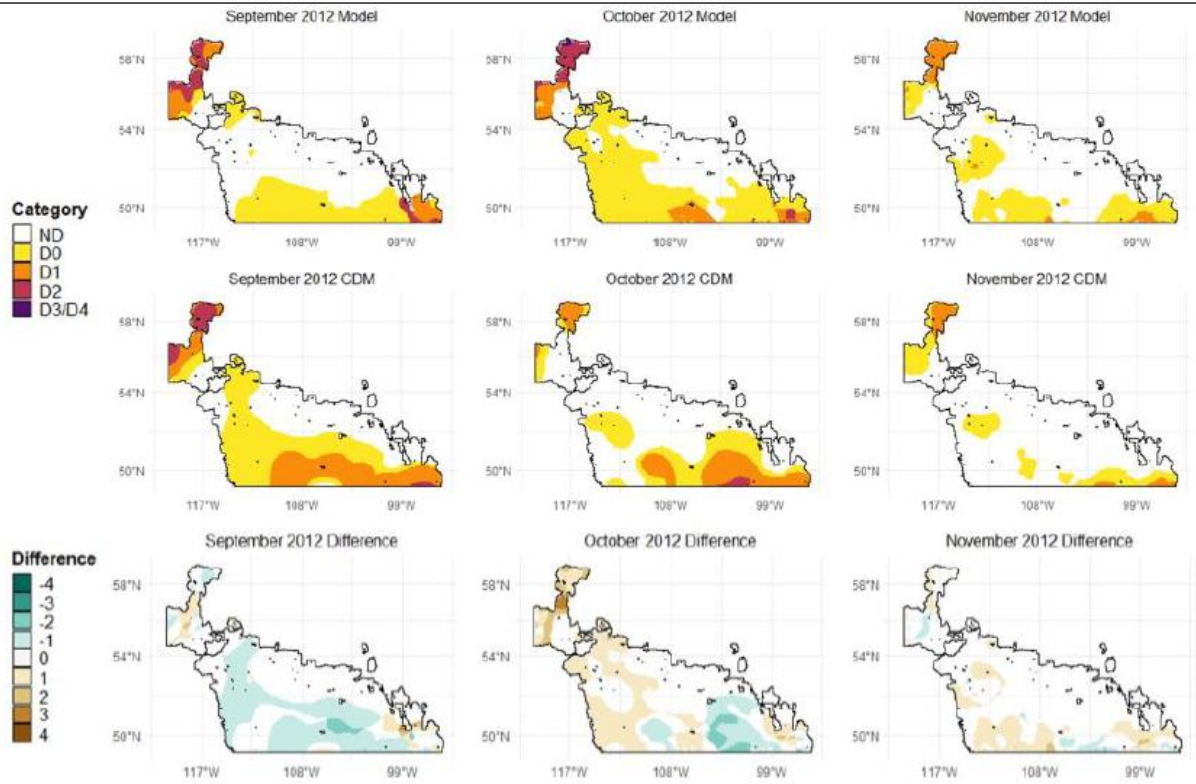
- Indicators used in drought forecast model as dominant predictors tend to be mid-term for separating drought/no drought (6 month range)
- Indicators for determining severity of each drought class are in the 9-12 month range

Correlation Between Forecasted and Station Based Indices 2021



- Available on website <https://agriculture.canada.ca/en/agricultural-production/weather/canadian-drought-outlook> and at <https://open.canada.ca/data/en/dataset/2c82daab-f6d9-4b19-96b5-238249e09fb9>

Development of High Resolution Monitoring and Forecasting System – *Jacob Mardian (UofG)*



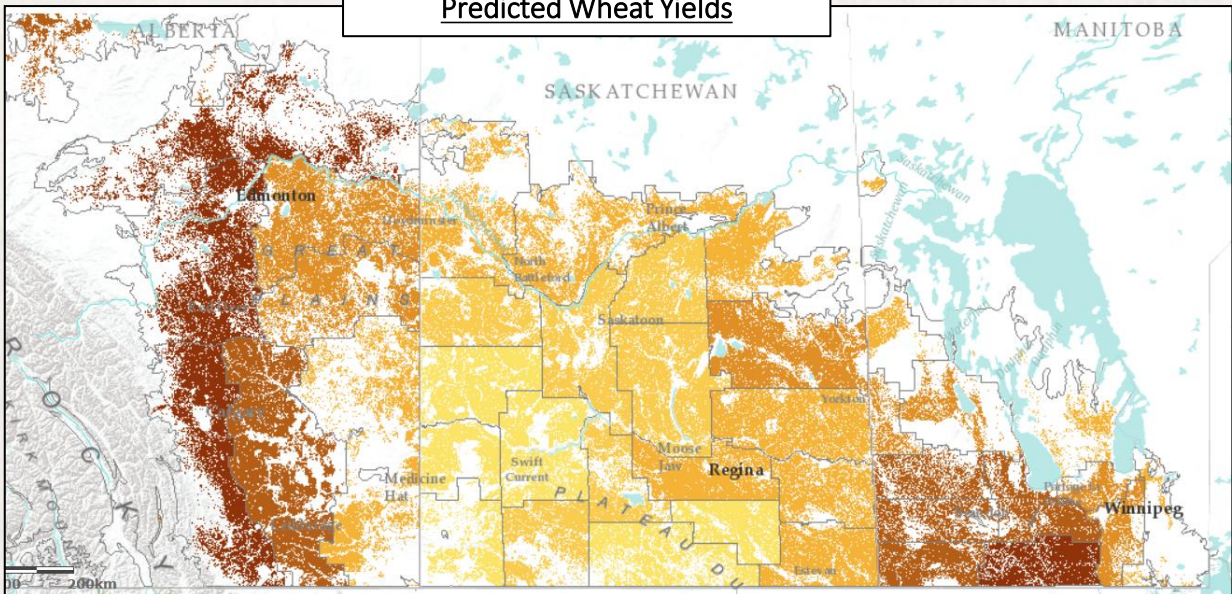
- Using an eXtreme Gradient Boosted Decision Tree model at 5km grid cells to predict Canadian Drought Monitor categories
- Model was trained over agricultural regions of 3 Prairie provinces for 2005 – 2009 and evaluated using data from 2010 to 2019.
- Used Shaley Additive Explanation variable importance metric to evaluate explanatory value of each input variable
- Includes spatial and temporal propagation metrics
- Can be expanded to include both real time and forecasted variables

From [Mardian, J., C. Champagne, A. Berg and B. Bonsal \(in review\)](#). ***Machine Learning Framework for Predicting and Understanding the Canadian Drought Monitor in Water Resources Research.***

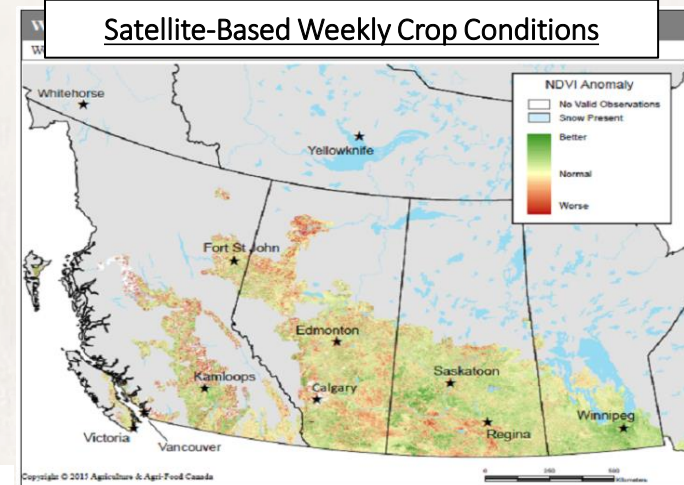
Canadian Crop Yield Forecasting System

- National model used within AAFC to forecast end of season crop yields based on earth observation based Normalized Difference Vegetation Index (NDVI) and climate based indicators
- Climate based indicators come from ~450 weather stations in Canada. Heat and water accumulation and evapotranspiration are modelled using Versatile Soil Moisture Budget Model. These inputs are used as predictors in the yield forecast model
- The model uses a robust linear regression using top ranked predictors from a leave one out cross validation process
- Predictors identified from historical training data are used to predict current year's yield
- Model has been adapted by Statistics Canada to replace the July and September Farm Survey

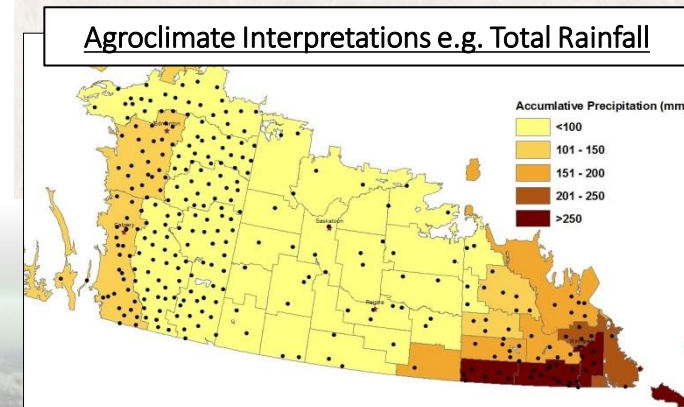
Predicted Wheat Yields



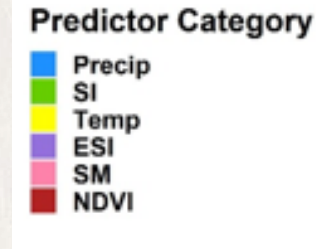
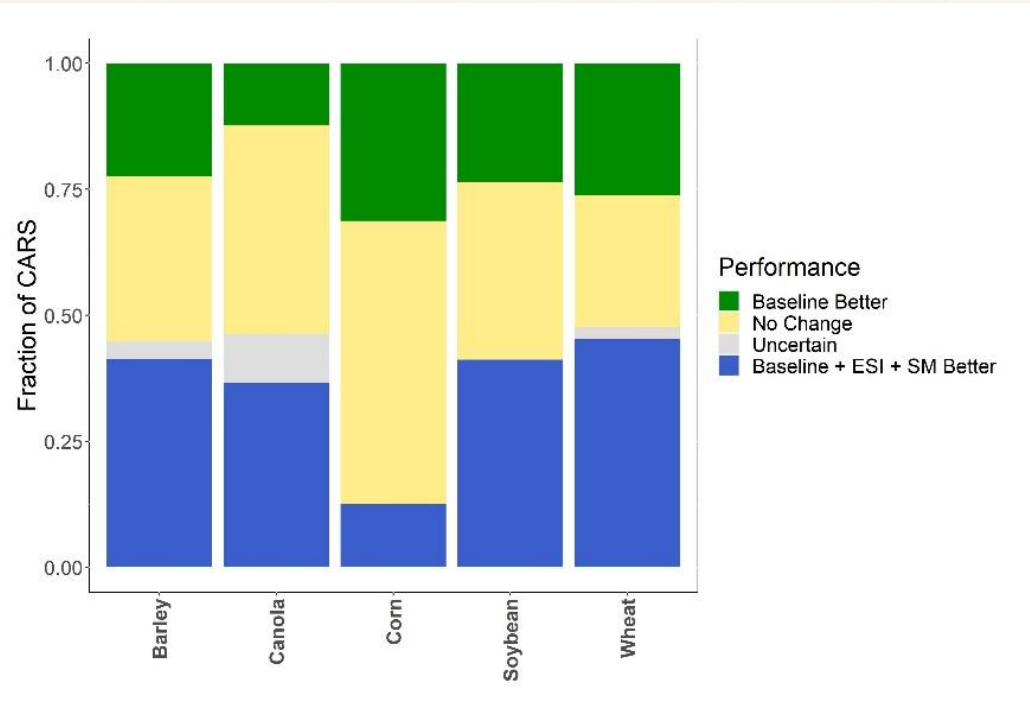
Satellite-Based Weekly Crop Conditions



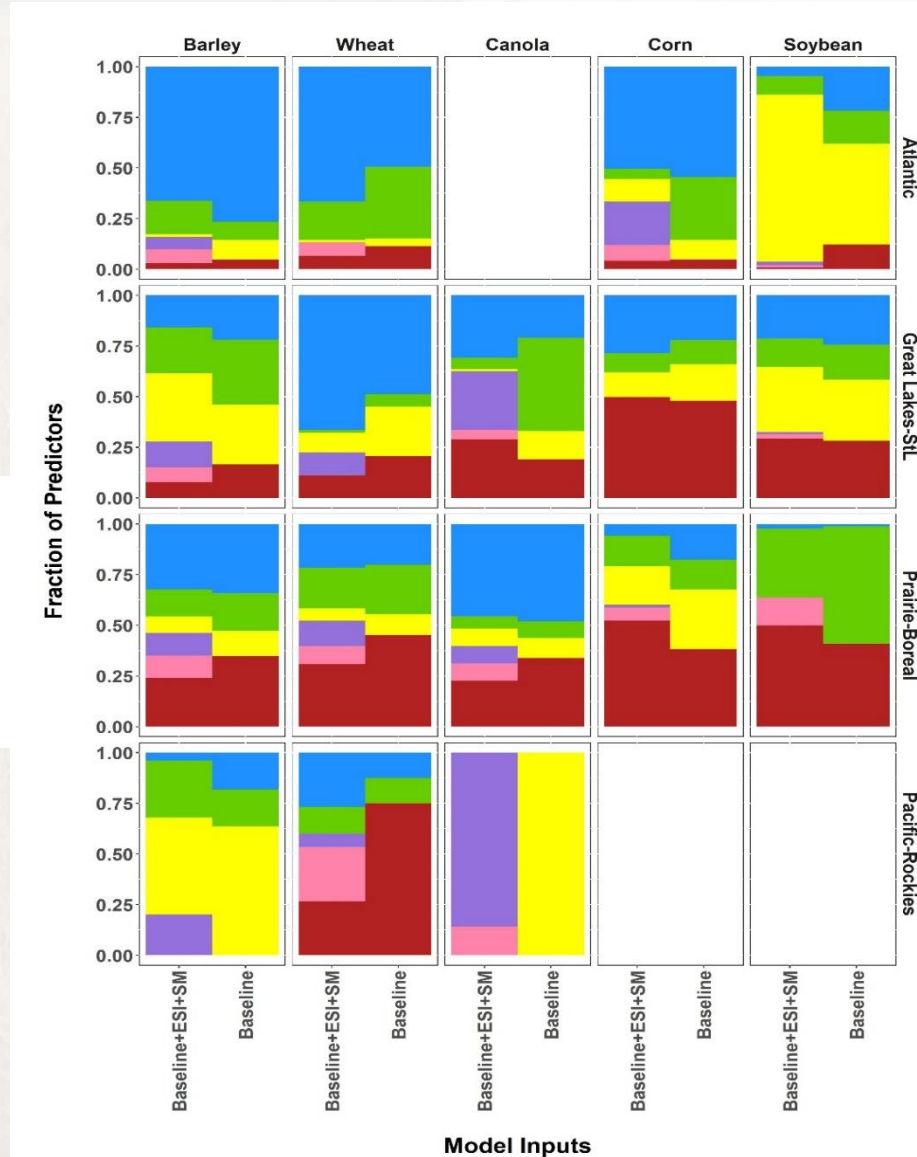
Agroclimate Interpretations e.g. Total Rainfall



Improving Yield Forecasts with Multifrequency EO

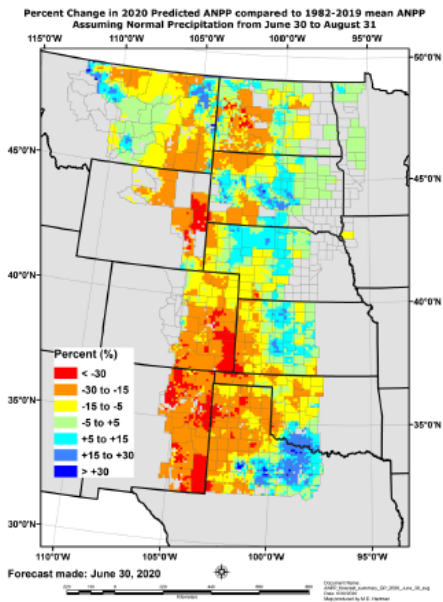


- Inclusion of satellite soil moisture and ESI improved crop yield prediction in many regions particularly for canola, barley and wheat
- SM was selected as a predictor in different regions for different crops, often displacing weather related variables or NDVI



Grass-Cast Canada

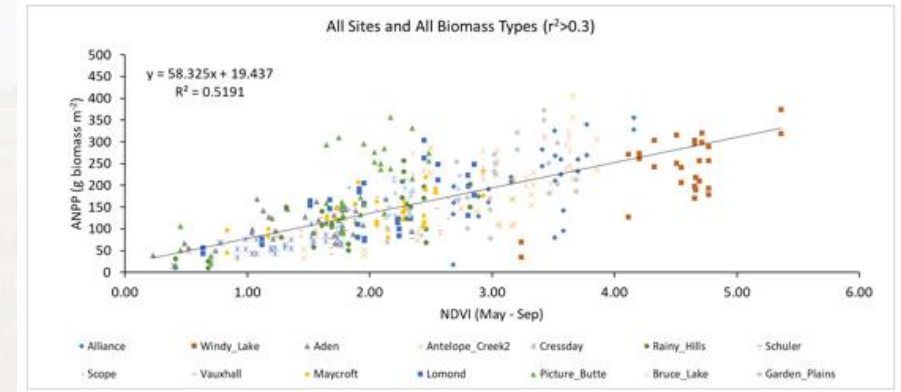
“Grass-Cast” – the Grassland Productivity Forecast



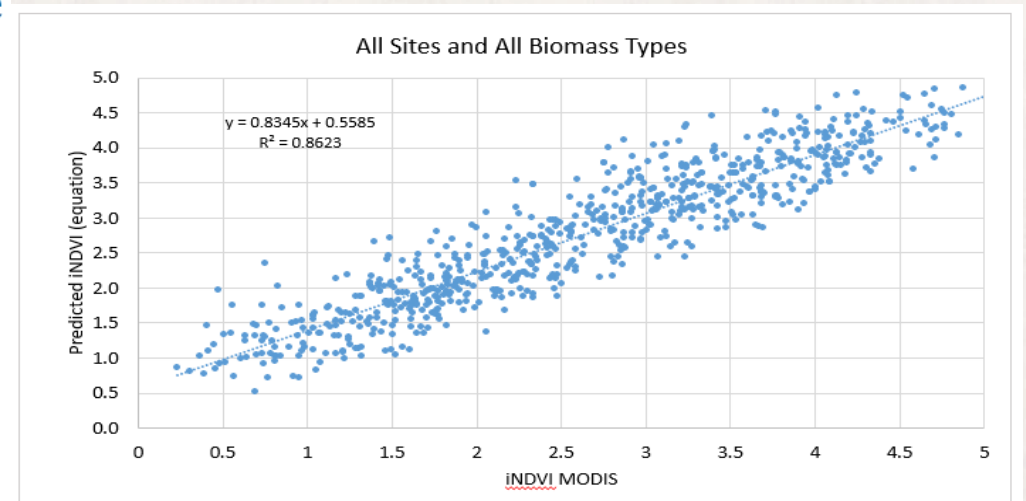
Based on **observed weather** to date
+ **future weather scenarios...**
we expect **grassland productivity**
in *your* area...
to be **X% higher** or **lower** than
your area's long-term average.

Dannele Peck, Bill Parton & Melannie Hartman
USDA Climate Hubs & Colorado State University

Observed Biomass with MODIS iNDVI



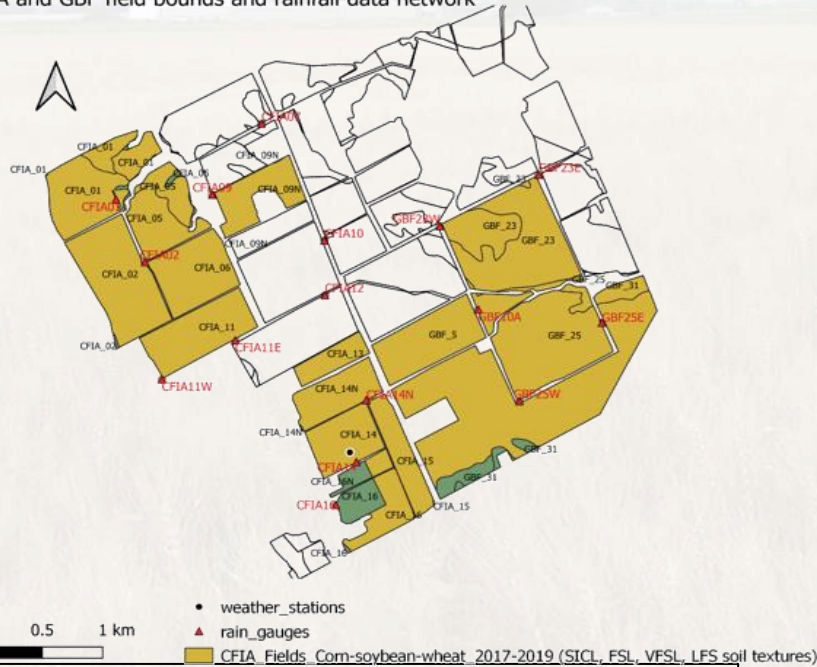
iNDVI vs. AET and MNDVI equation



- Developed relationship between grassland biomass, actual ET and NDVI to build an ensemble predictive model for grassland productivity in Canada

Optimization of Crop Models

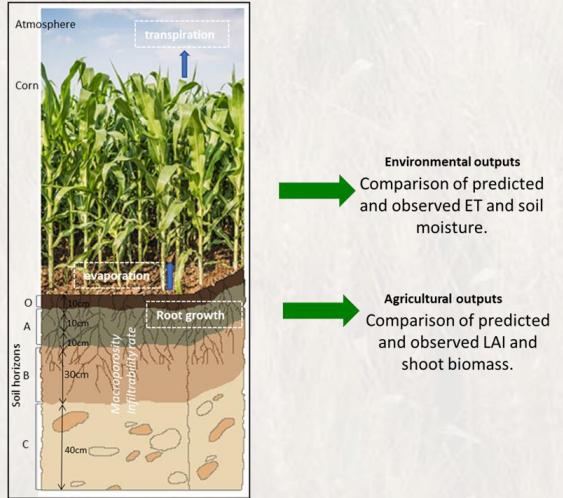
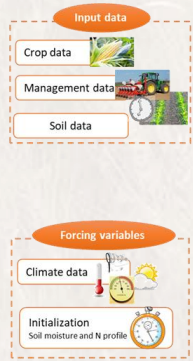
CFIA and GBF field bounds and rainfall data network



- Use gridded climate data, remotely sensed Leaf Area Index and Surface Soil Moisture to optimize crop models
- Optimize seeding date, seeding density, soil moisture initial conditions and soil water content at field capacity
- Initial work focused on adjusting crop models to estimate ET and soil moisture to accept remotely sensed estimates

Saadi, Sameh & Pattey, Elizabeth & Jégo, Guillaume & Champagne, Catherine. (2022). Prediction of rainfed corn evapotranspiration and soil moisture using the STICS crop model in eastern Canada. *Field Crops Research*. 287. 108664. 10.1016/j.fcr.2022.108664.

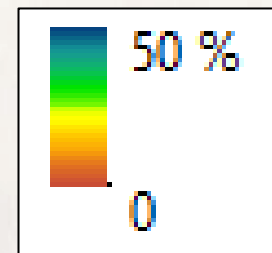
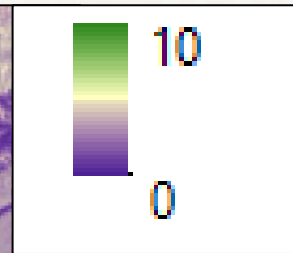
STICS MODEL



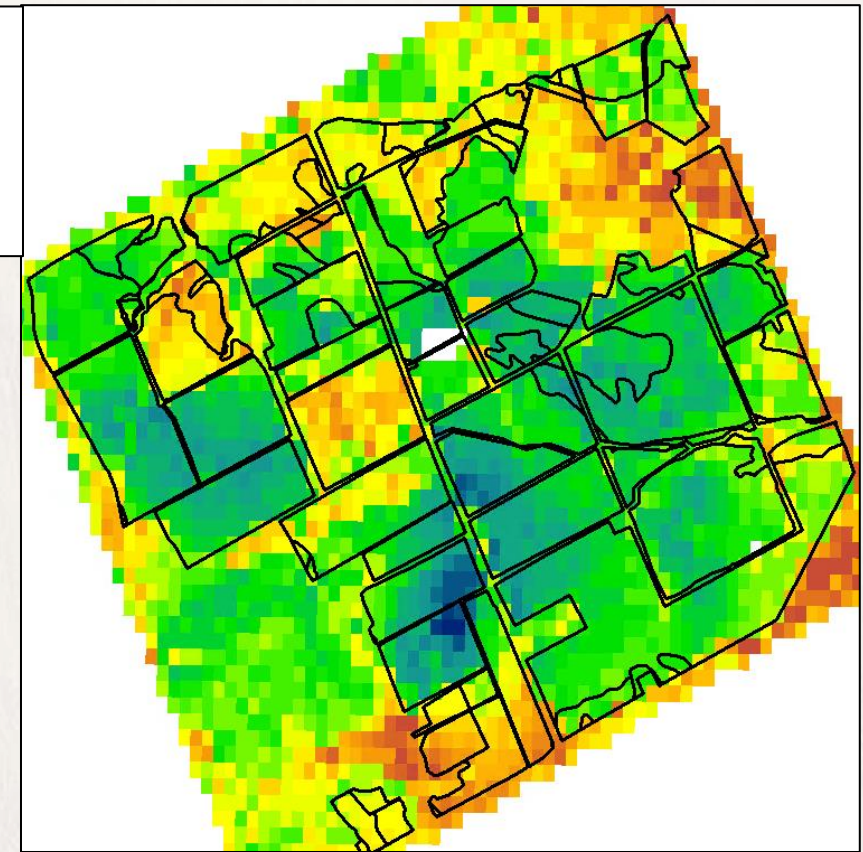
Earth Observation Data for Model Optimization



- 10m resolution Leaf Area Index derived from time series Sentinel-2 data
- Evaluated several methods: Index (MTVI-2), Sentinel-2 Agri/SNAP Toolbox and LEAF Toolbox (Google Earth Engine)

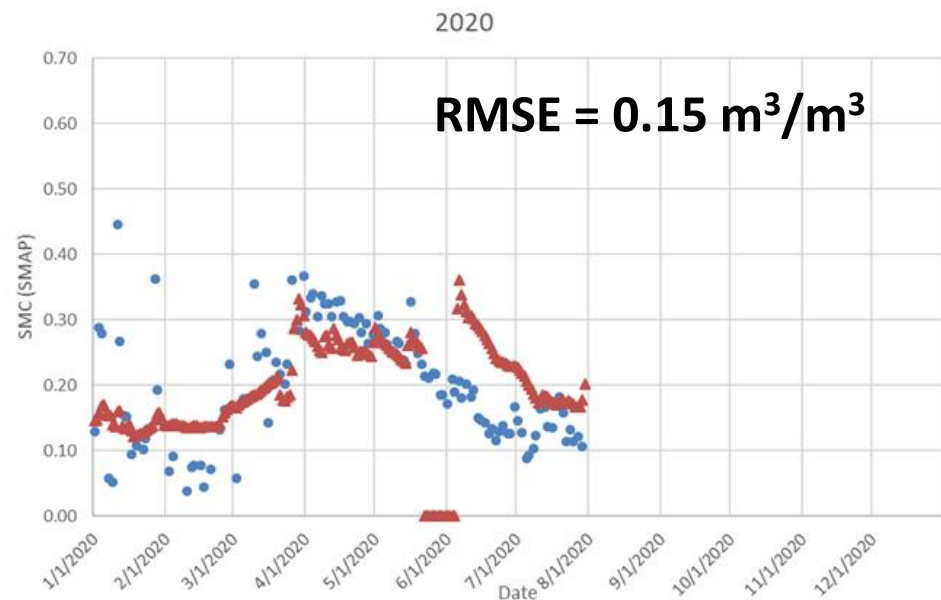
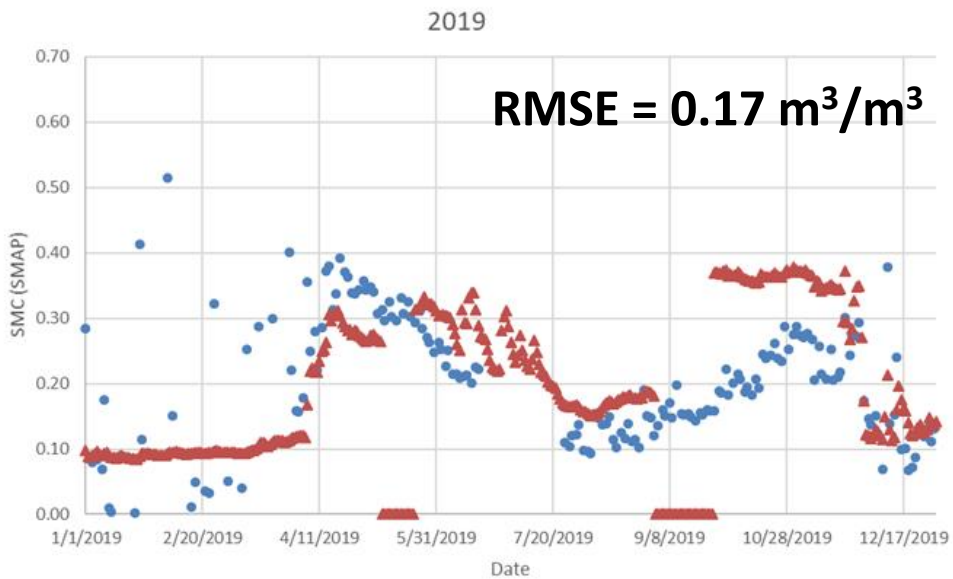
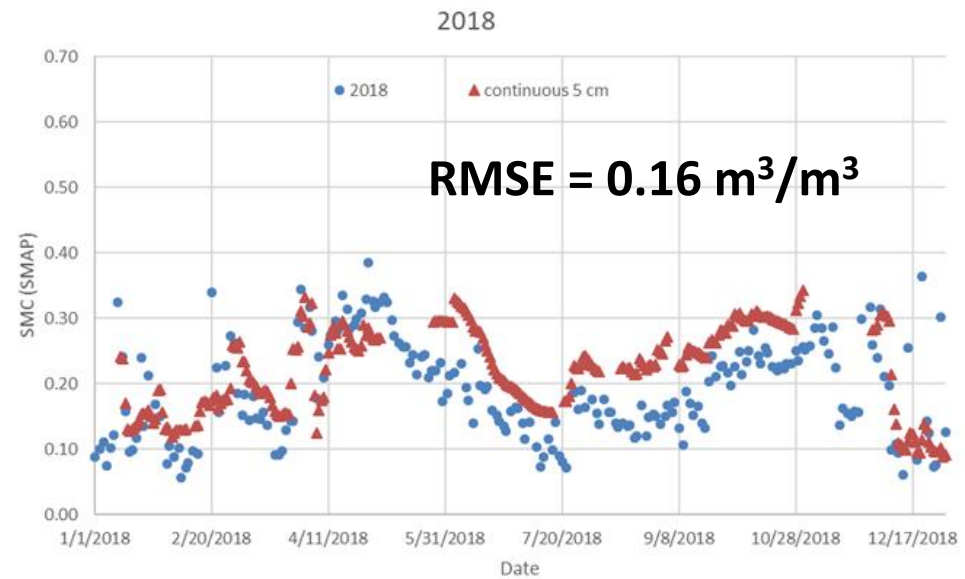
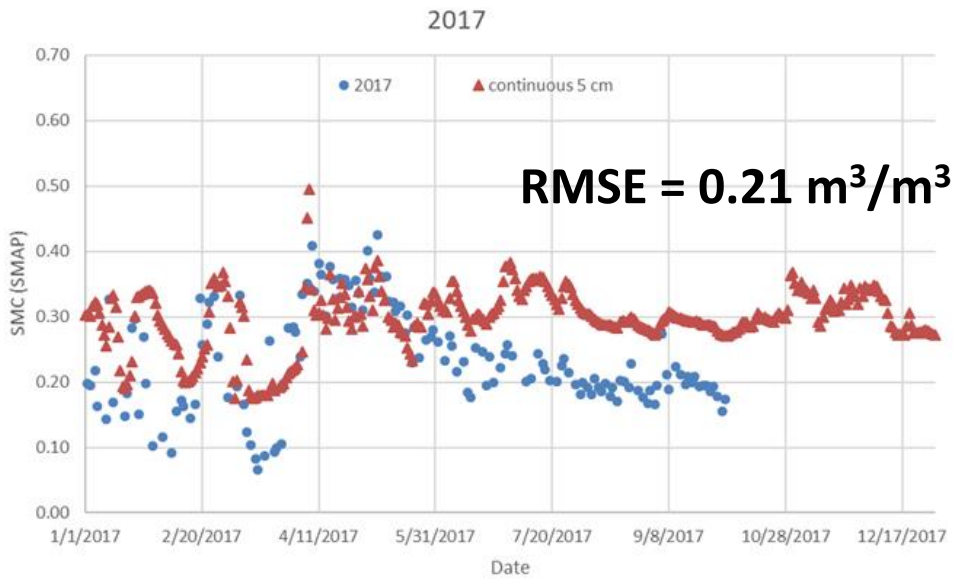


EO data sets can be used to optimize unknown variables in model (seeding date/density or soil texture / soil moisture) initialization (*Jégo et al 2012, Field Crops Research*)



- 100m resolution surface soil moisture from Vandersat (SMAP + Sentinel-1)
- Compared against coarse resolution from SMOS/native SMAP & ground measurements

Satellite Soil Moisture Data

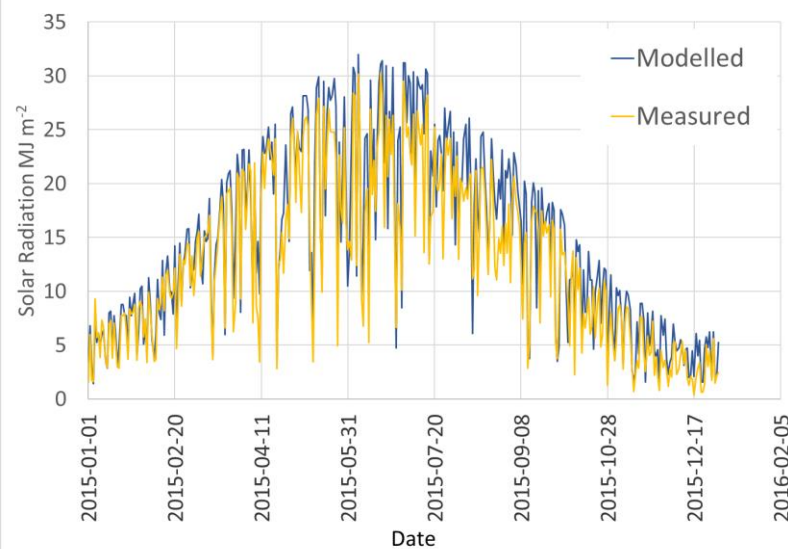
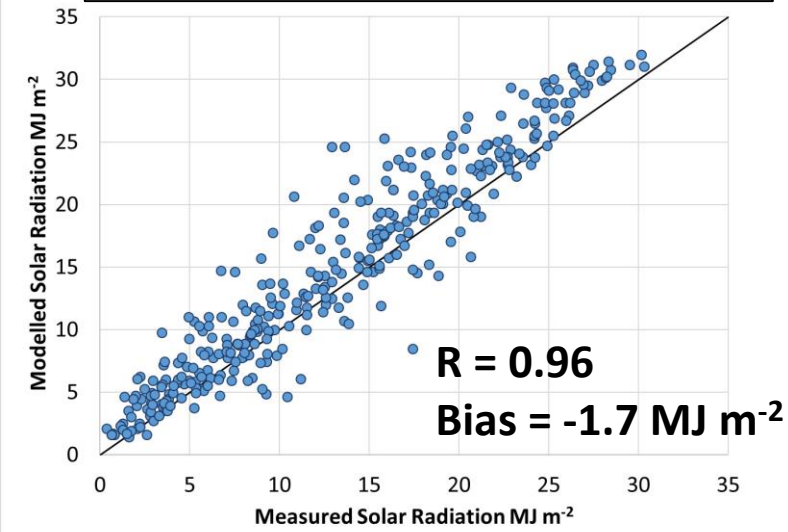


High resolution soil moisture for Vandersat shows good capturing of trends at local level

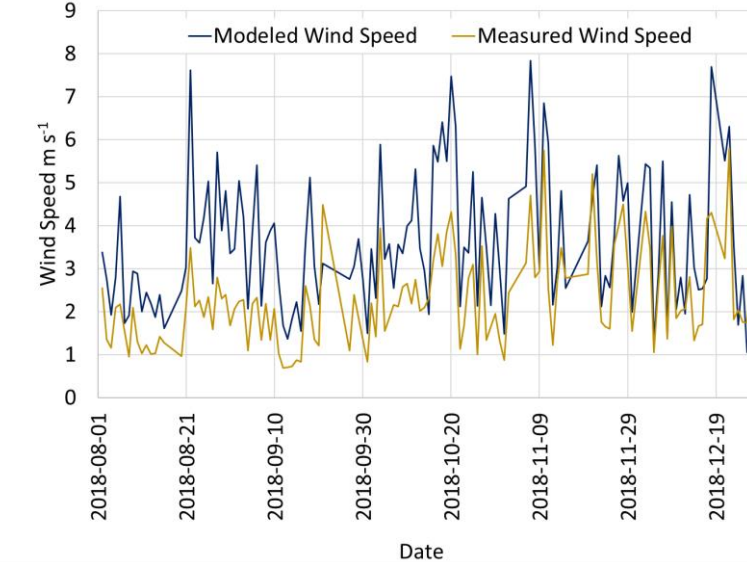
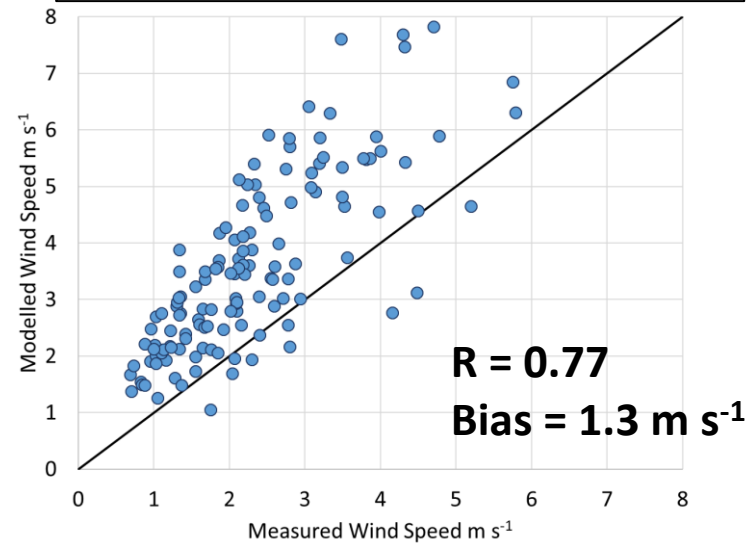
Biases are still too high to distinguish between soil properties in high pedodiversity regions

Other Key Gridded Meteorological Data Sets

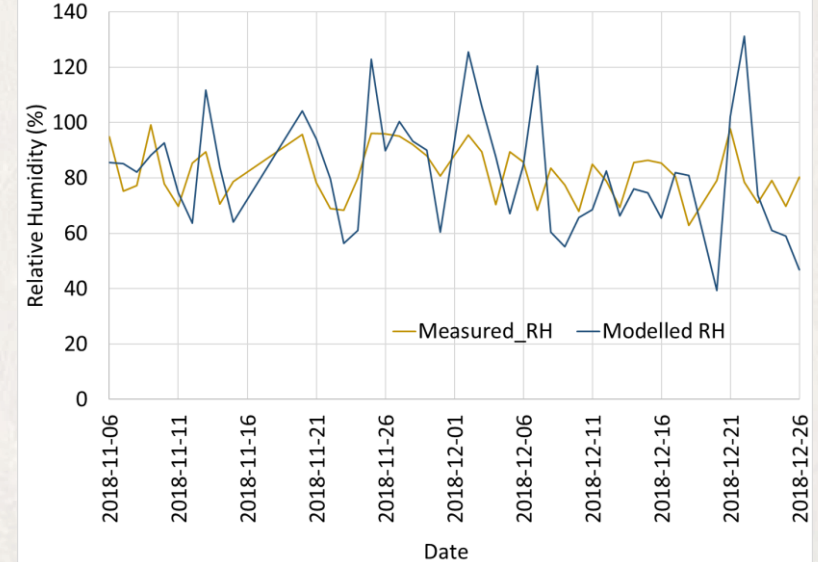
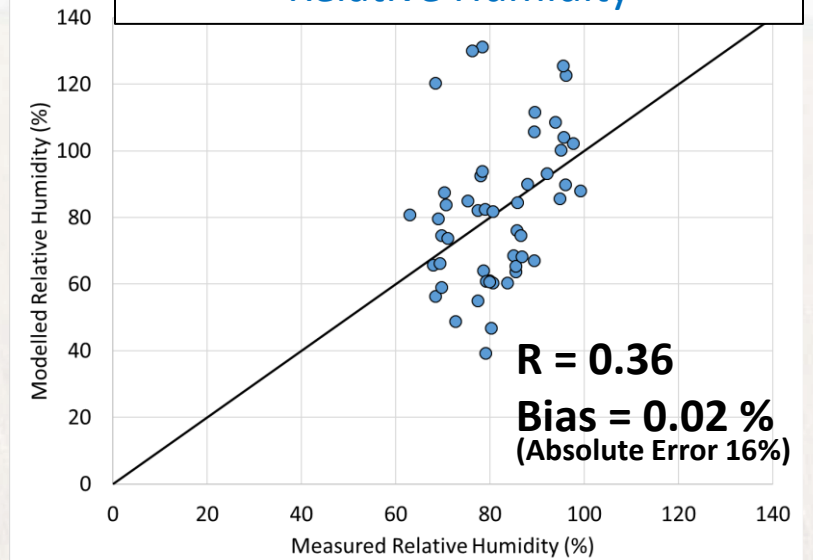
Solar Radiation

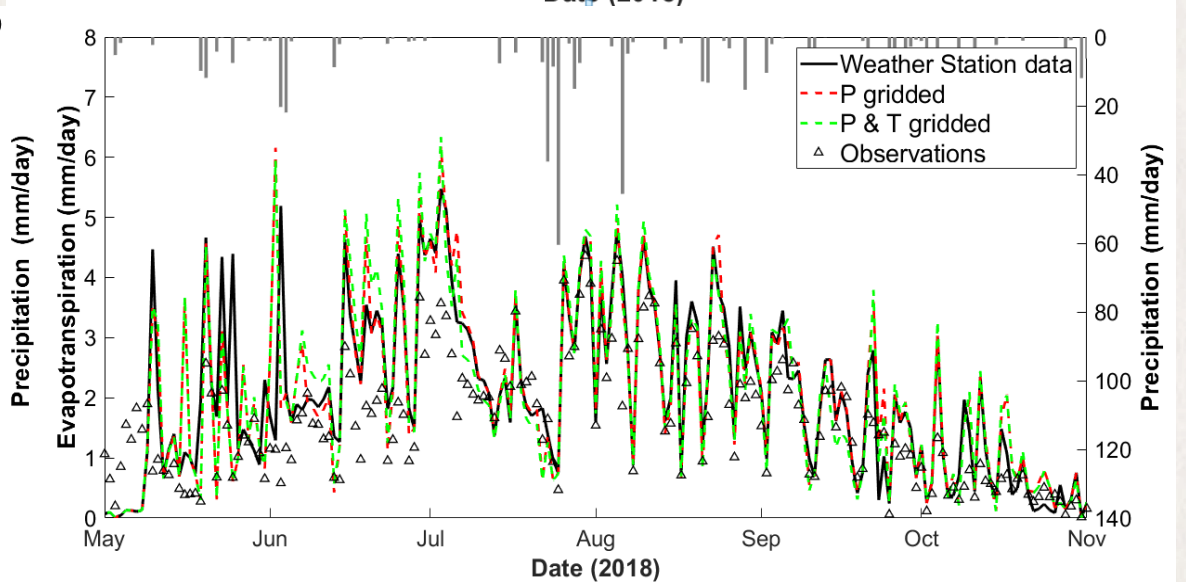
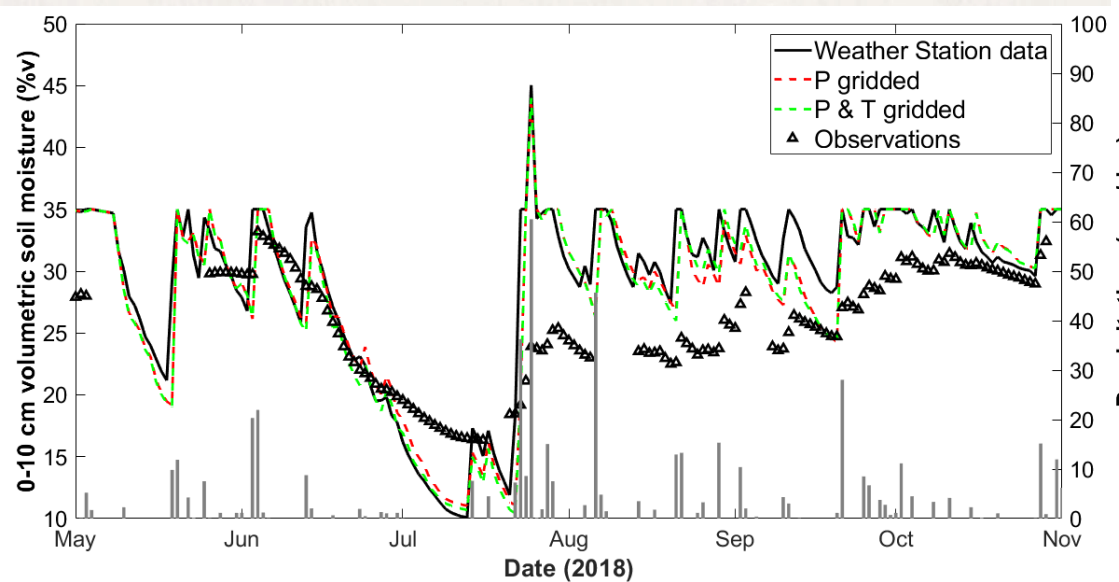
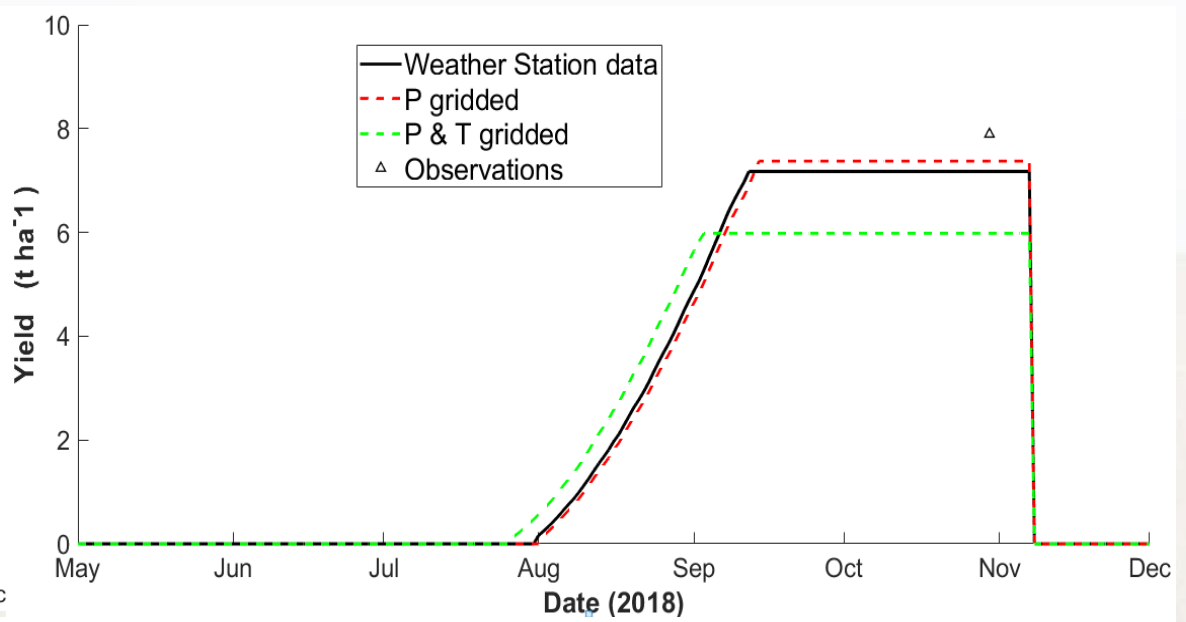
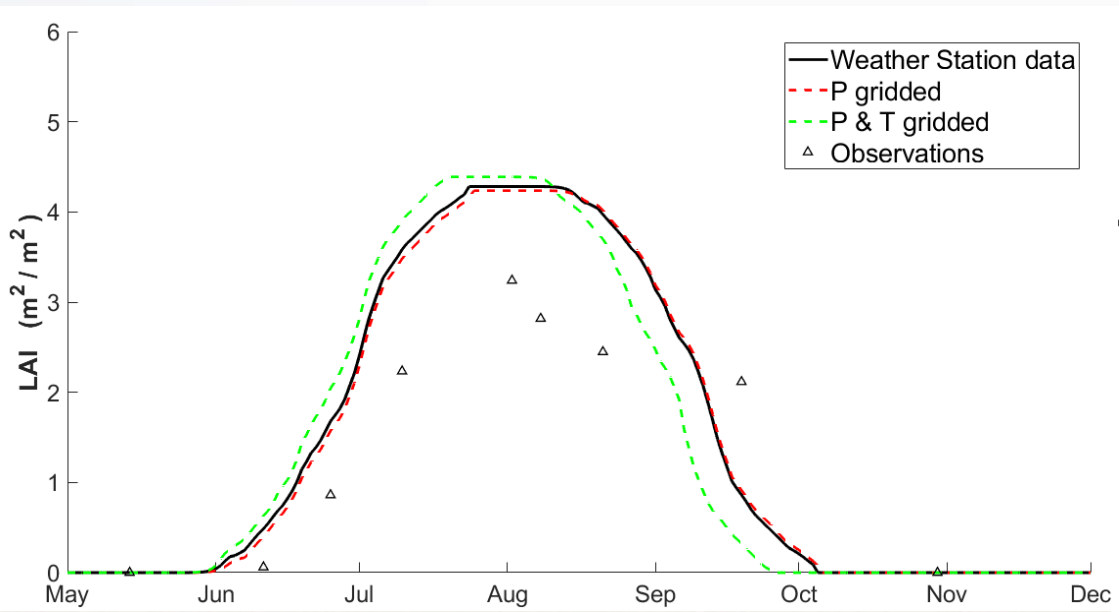


Wind Speed



Relative Humidity

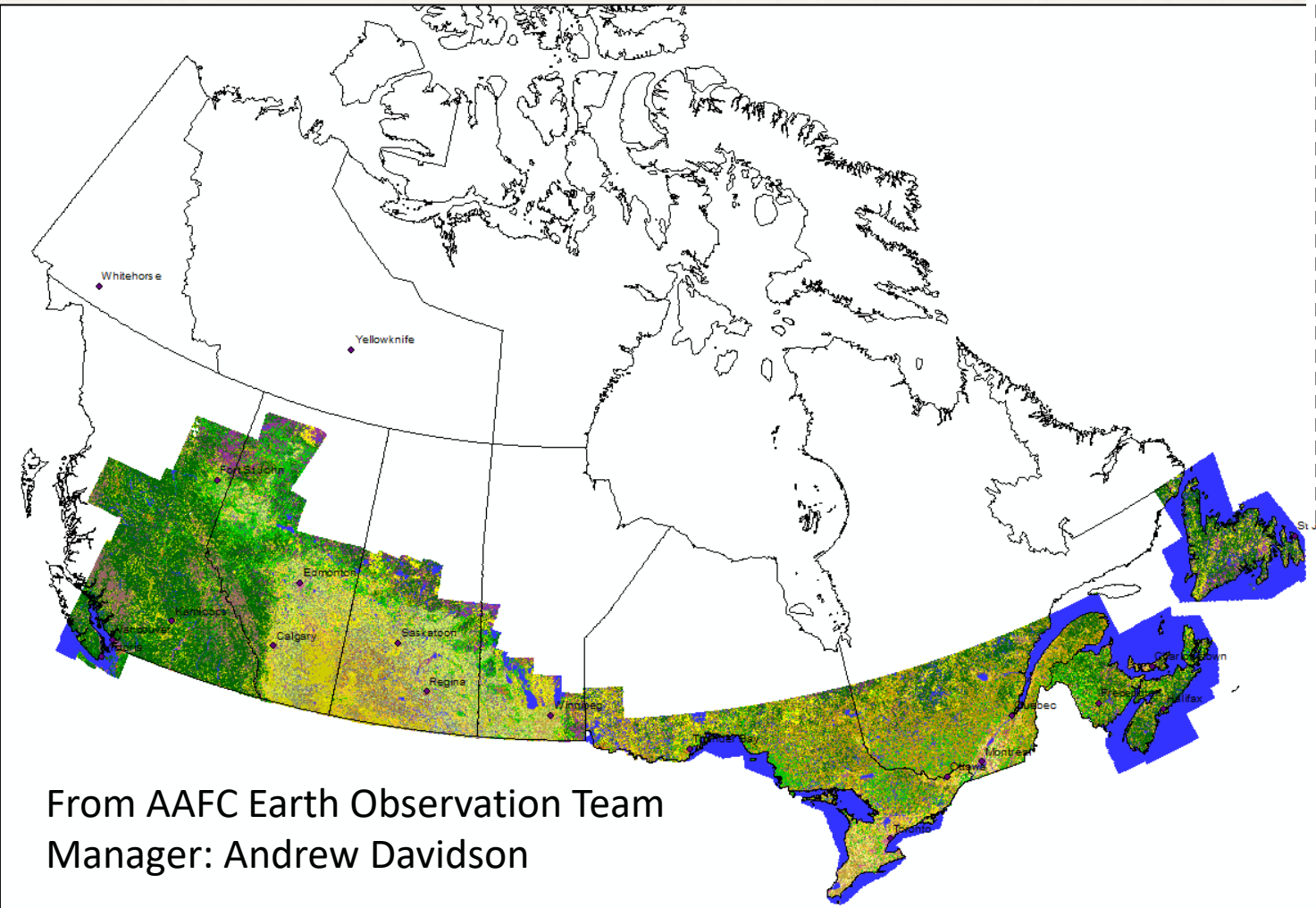




- Evaluated impact of using gridded precipitation and temperature data from HRDPS/HRDPA in STICs crop model for field site in Ottawa ON (results shown for 2018)
- Impact on Crop Models

Other Key EO Data Sets: Crop Inventory

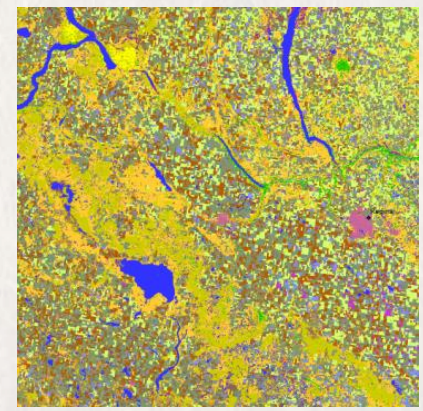
- Created annually using field observations, and multi-temporal optical remote sensing (Landsat) and radar remote sensing (Radarsat-2)



Legend

| | |
|---|--|
| Water / Eau | Safflower / Carthame |
| Exposed Land / Sol nus | Sunflowers / Tournesols |
| Developed / Zones développées | Soybeans / Soja |
| Shrubland / Arbustes | Pulses / Légume à Gousse |
| Wetland / Terres humides | Peas / Pois |
| Grassland / Prairies | Beans / Fèves |
| Agriculture (undifferentiated) / Agriculture (indifférenciée) | Lentils / Lentilles |
| Hay / Pasture / Cultures pérennes et pâturages | Vegetables / Légumes |
| Too Wet to be Seeded / Trop humide pour le semis | Potatoes / Pomme de terres |
| Fallow / Jachère | Sugarbeets / Betteraves à sucre |
| Cereals / Céréales | Other Vegetables / Autres légumes |
| Barley / Orge | Fruits / Fruits |
| Other Grains / Autres céréales | Berries / Baies |
| Millet / Millet | Orchard / Vineyards / Vergers / Vignobles |
| Oats / Avoine | Other Fruits / Autre fruits |
| Rye / Seigle | Grape / Raisins |
| Spelt / Épeautre | Hops / Houblon |
| Triticale / Triticale | Sod / Gazonnière |
| Wheat / Blé | Herbs / Fines Herbes |
| Winter Wheat / Blé d'hiver | Nursery / Pépinière |
| Other Wheat / Autres blés | Buckwheat / Sarrasin |
| Corn / Maïs | Canary Seeds / Alpipiste des Canaries |
| Tobacco / Tabac | Hemp / Chanvre |
| Ginseng / Ginseng | Other Crops / Autres cultures |
| Borage / Bourrache | Forest (undifferentiated) / Forêt (indifférenciée) |
| Camelina / Caméline | Coniferous Trees / Forêt de conifères |
| Canola / Rapeseed / Canola / Colza | Deciduous Trees / Forêt de feuillus |
| Flaxseed / Graines de lin | Mixed Trees / Forêt mixte |
| Mustard / Moutarde | |

From AAFC Earth Observation Team
 Manager: Andrew Davidson



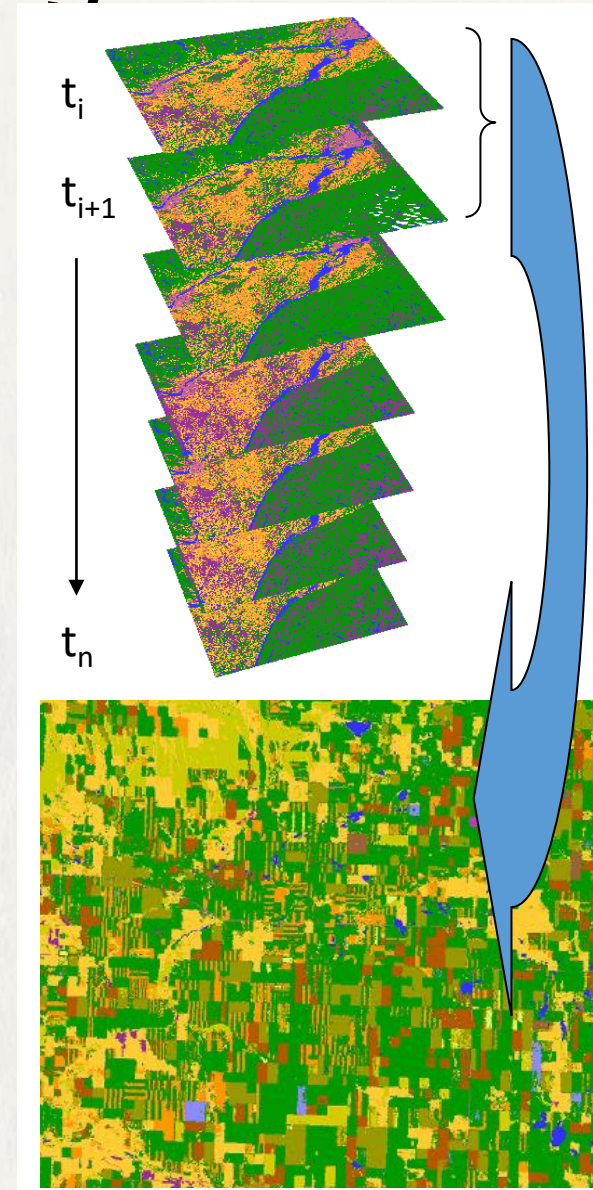
Working on cover crops, tillage, growth stage, grasslands, biomass

Agricultural Land Use Change

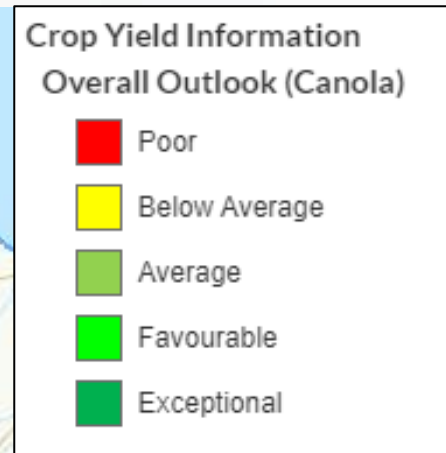
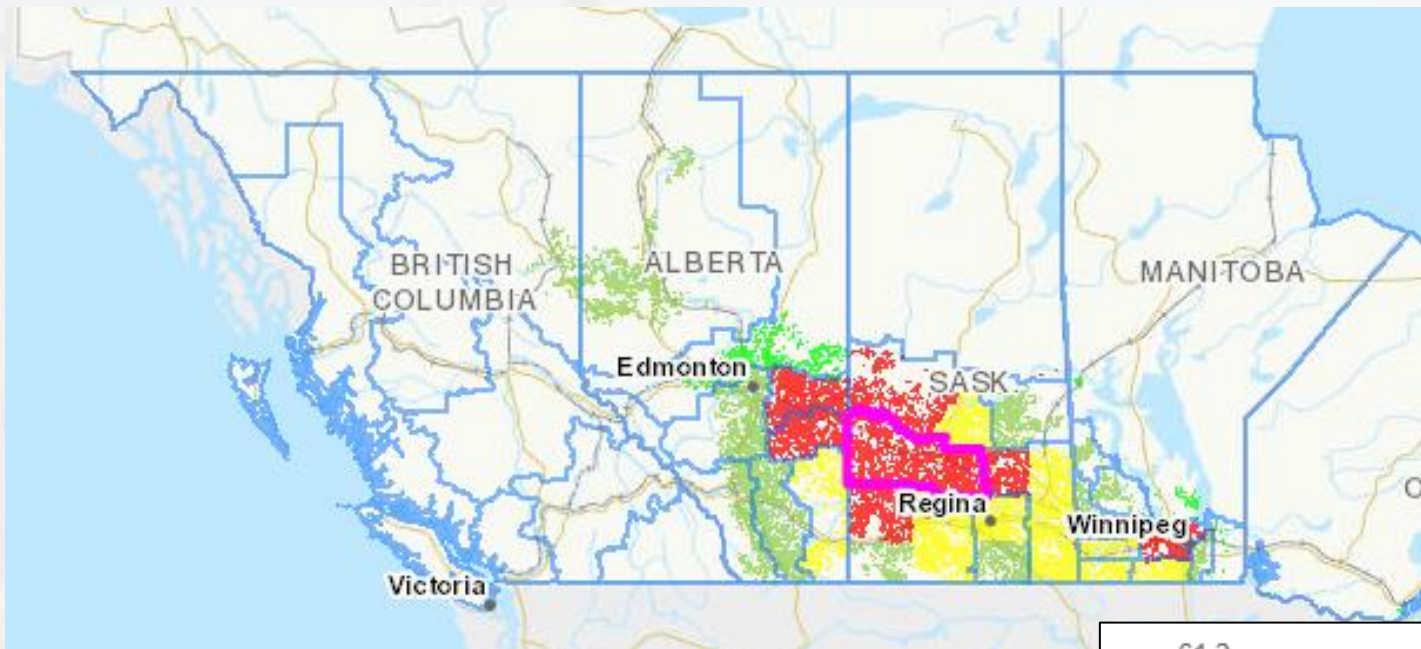
- Land use change indicators will indicate “where”, “how much” and “how” agricultural land use has changed.

Importance

- Allows annual land use changes to be tracked between important cover types.
 - Forest to agriculture,
 - Grassland to cropland,
 - Urbanization of agricultural land,
 - Perennial to annual crops.
- Such changes are key to estimating carbon sinks and sources in agricultural landscapes
- Allows spatially specific integration with other bio-physical data sets (i.e. detailed soils).

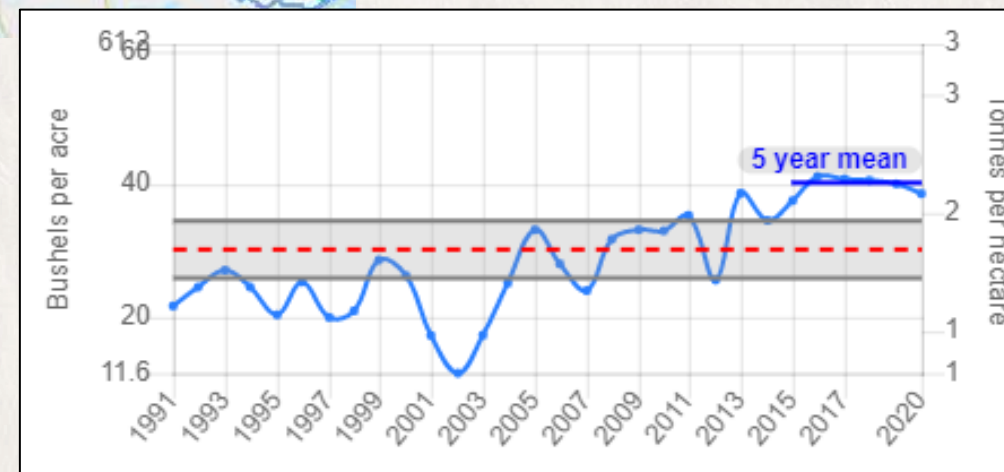


Tools: Canadian Crop Metrics Application



From AAFC Geomatics Team
Manager James Ashton

- Overall yield outlook, production forecasts, crop stress indicators, extreme weather indicators, drought, satellite data, historical analysis & graphing tools



Estimated canola yield for Saskatoon Region

Canadian Crop Metrics Application:
www.agriculture.Canada.ca/atlas/cropmetrics



Global Crop Monitoring

JECAM
Joint Experiment for Crop Assessment and Monitoring

Geo GROUP ON EARTH OBSERVATIONS

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JECAM GOALS

The overarching goal of JECAM is to reach a convergence of approaches, develop monitoring and reporting protocols and best practices for a variety of global agricultural systems.

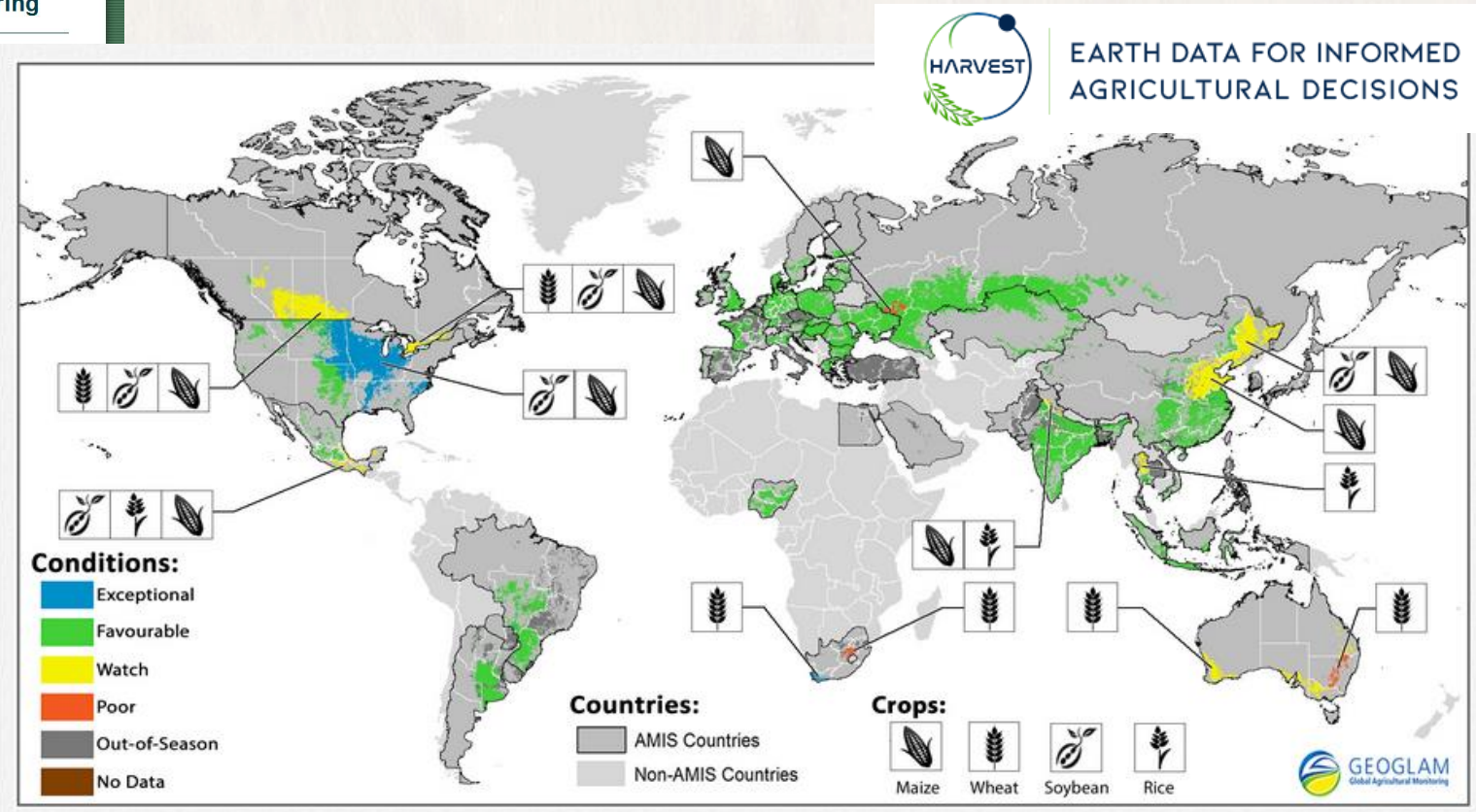
Agriculture Community of Practice

Joint Experiment of Crop Assessment and Monitoring

The overarching goal of JECAM is to reach a convergence of approaches, develop monitoring and reporting protocols and best practices for a variety of global agricultural systems. JECAM will enable the global agricultural monitoring community to compare results based on disparate sources of data, using various methods, over a variety of global cropping systems. It is intended that the JECAM experiments will facilitate international standards for data products and reporting, even supporting the development of a global system of systems for agricultural crop assessment and monitoring. The initiative is developed in the framework of GEO Global Agricultural Monitoring (GEOSS Task AG0703 a) and Agricultural Monitoring (GEOSS Task AG0703 b).

Take successful national monitoring and improve methods to expand this globally

Produce crop forecasts based on scientifically sound, geospatial data to reduce spurious market speculation that follows environmental and political events



Take home messages

- The National Agroclimate Information Service leverages raw weather and climate data to translate this to information useful for assessing risks, monitoring production and analysing the sustainability of the agricultural sector in Canada – key tools for mitigating and adapting to climate change
- Station networks, satellites, modelled data, crowd sourcing all have a role to play in getting full picture of climate related impacts to agriculture and beyond
- Improving the spatial resolution and richness of indicators will strengthen our ability to respond to extreme weather as it happens; earth observation and integration of this data into models is critical to providing this information
- User-friendly tools, simplified data sets make these accessible to diverse user groups to interpretation and decision making
- Find everything at: <https://agriculture.canada.ca/en/agricultural-production/weather>

Trevor Hadwen



Yinsuo Zhang



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Aston Chipanshi



Laura Richard



Tyler Black

Richard Warren



Marilee Pregitzer

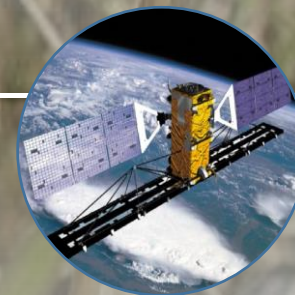
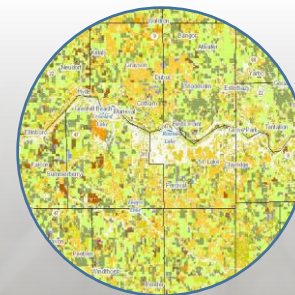




Agriculture and
Agri-Food Canada

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