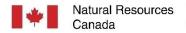


Case Study 2: Agriculture and Agri-Food Canada **Crop Inventory**









Agriculture in Canada and Globally

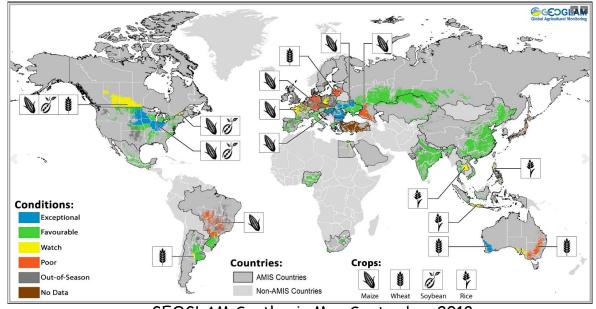
Canada's agricultural landscape is large and complex

- 230,000 farms
- Total farm area: 67.6 million hectares
- Land in crops: 35.9 million hectares

- 8.1% of total GDP
- 5th largest exporter of agricultural products
- Employs 2.2% of Canada's total population

National, regional and global challenges in food supply and demand

- World prices of wheat, rice,
 corn, and soybeans rose 226%
 from 2002 to 2008
- UN Food and Agriculture
 Organization (FAO) estimates
 that food production must
 double in the next 40 years to
 meet global needs
- Competing land uses and other uncertainties introduce risk



GEOGLAM Synthesis Map September 2018 Crop Growing Conditions for Maize, Wheat, Soybean and Rice

Need sound policies and risk management strategies based on accurate, timely and cost effective information

Image source: GEOGLAM.org

Earth Observation at AAFC

- AAFC has been conducting research on EO applications of space science for well over 30 years
 - World leader in agricultural monitoring and mapping, especially related to SAR applications.
 - Focus on research to support existing and future operations, followed by operational implementation
 - Sharing SAR methodology with scientists in other countries and international organizations to support agriculture monitoring
- Recent convergence of technologies (satellite imagery, software tools and hardware) allowed for operational solutions to support government policy development, program implementation and performance measurement
- AAFC can leverage Canadian space assets like RADARSAT-2 and the future RADARSAT Constellation Mission (RCM)
- Within a few years AAFC should be able to frequently image all of Canada's agricultural landscape with high temporal frequency at medium to high resolution and at low or no cost









Crops in Canada

Mandate

Earth Observation at AAFC produces Annual Crop Inventory

- Satellite-based land cover / land use maps
 - Initial focus on Prairie in 2009 and 2010
 - Expanded to entire agricultural extent of Canada in 2011
 - Utilizes multispectral and dual-polarization radar imagery
- Digitally-based and published on GoC Open Data Portal / AAFC Geospatial viewer
 - https://open.canada.ca/en/open-data
- Internally / externally used
 - Yield prediction, rotation patterns, crop acreages
 - Landscape fragmentation, habitat pressures
 - Crop marketing / business planning
 - Many additional uses, beyond agriculture



Landsat 8

Image source: GISGeography.com



RADARSAT-2

Image source: asc-csa.gc.ca

Current Operational Classifier

- STB-EOS Classification System (STB-EOS CS)
- Infrastructure built on series of scripts and GUIs
 - Decision Tree classifier (Breiman, 2001)
 - Permits integration of various data sources
 - In the current operational system, this is limited to set number of images
 - ~ 85% accuracy at national scale
 - Uses optical and SAR imagery
 - SAR includes a few images per season of dual polarization RADARSAT-2 data
 - Only intensity in VV and VH polarization

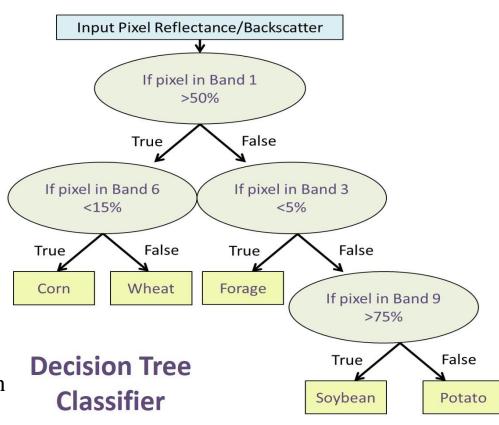


Image source: AAFC, Catherine Champagne

Importance of SAR Imagery for the Inventory

- Optical multi-spectral data (Landsat, DMC, AWiFS, Sentinel2) adequate to classify crops **if** data are available during critical periods of the growing season
- Accuracies greater than 85% have been achieved (overall and for individual crops), but can be significantly degraded by gaps in optical data collection
- Main issue is presence of clouds in multiple (or all) images
- The use of single-frequency, dual polarization (VV and VH) SAR imagery has been shown to increase the overall accuracies
- Current practice is use ~5-6 optical images and ~1-3 SAR images (non-optimal, due to system limitations)
- Overall, integrating SAR data with an adequate national optical coverage improves the accuracy of AAFC's annual crop inventory

Training and Validation Data Pre-processing - Collection



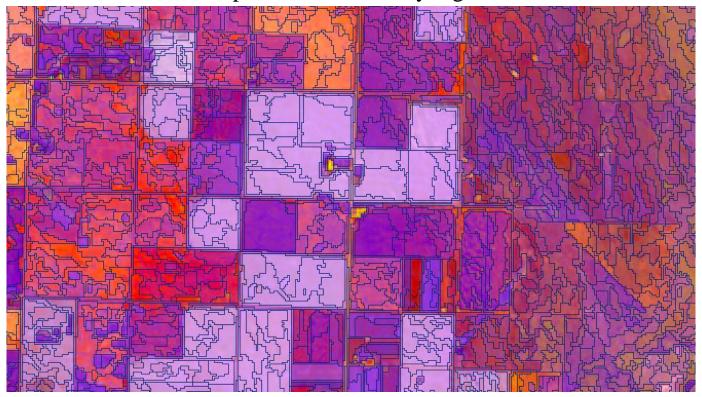
Southern Ontario Data Collection

- Locally provided insurance data (Prairie provinces and Quebec)
- Field data collection campaigns in summer by AAFC Staff in other provinces
- Coverage well dispersed to cover areas of interest
- Large quantities of samples especially for rare and unusual crop types
- Classification success is highly dependent on quality of ground data inputs

Poor ground data = poor classification!

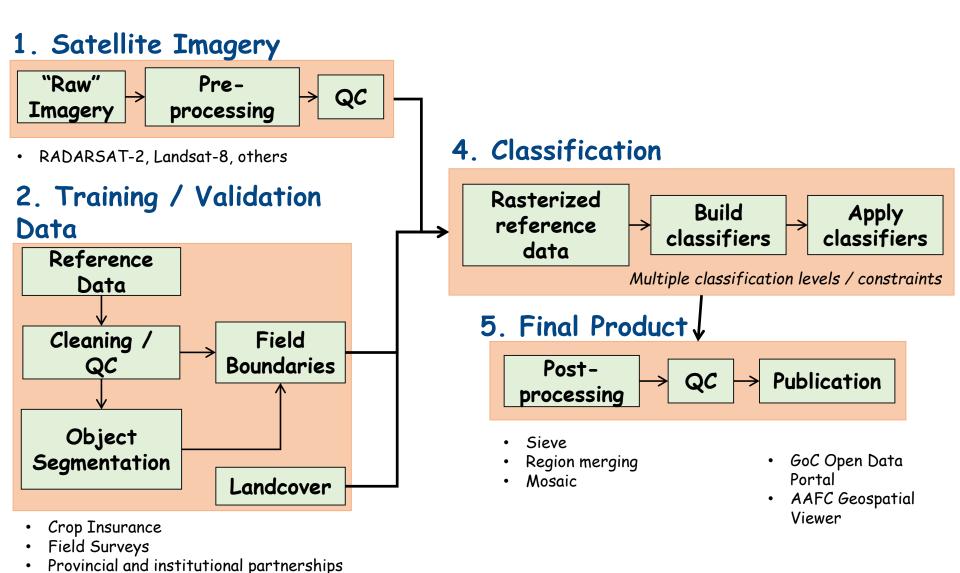
Training and Validation Data Pre-processing – In house Data Processing

- Match to AAFC legend
- Segmentation to simulate field boundaries
 - Not perfect but subsequently smoothed and quality controlled
- Converts field collected points to polygons
 - With a smaller dataset this step could be manually digitized



Object-based segmented current season Landsat image

Current Overall Crop Inventory Operational Methodology



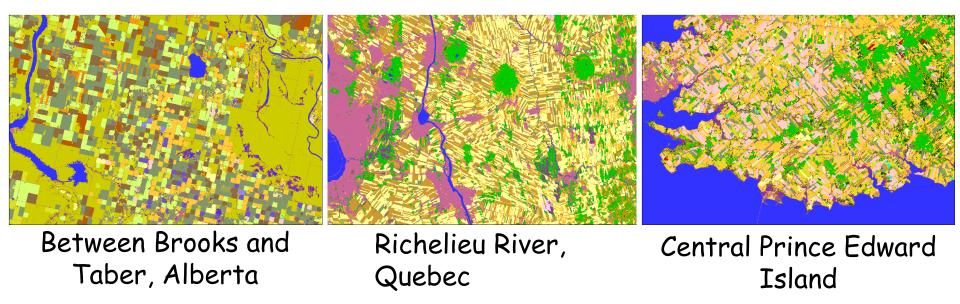
Current SAR Pre-processing Order of Operations

- 1. Download RADARSAT-2 (SLC) data with coverage over the entire agricultural extent of Canada
 - 2 or 3 images per season
 - Fills gaps for limited optical imagery areas
- 2. Convert data to Sigma naught (σ^0)
- 3. Speckle filter data with 7 x 7 Gamma Map filter
- 4. Rational Polynomial function model (ortho/terrain correction)
- 5. Feed corrected VV and VH images into the classifier with optical imagery and field data

2017 Crop Inventory



AAFC 2017 Annual Crop Inventory Local Examples

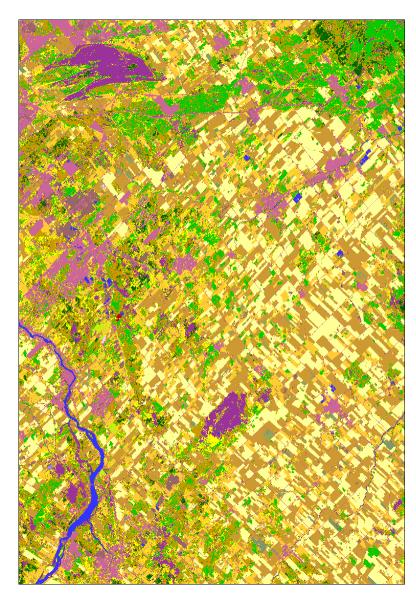


Future SAR Pre-processing Order of Operations

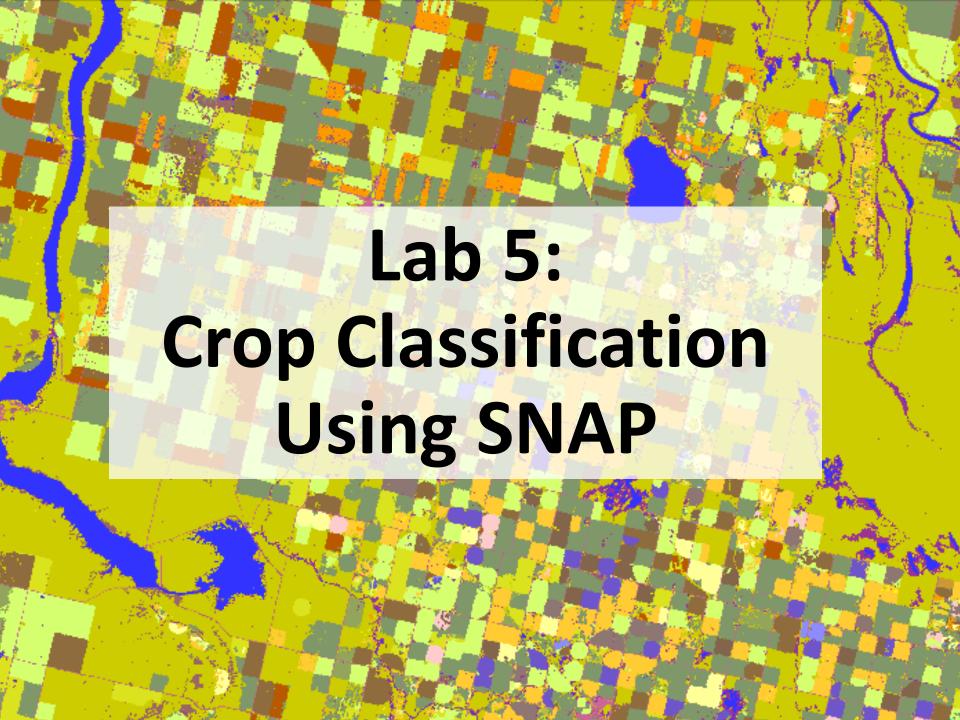
- 1. Download RADARSAT(SLC/GRD) and Sentinel-1 (SLC/GRD) data with coverage over the entire agricultural extent of Canada
 - At least 1 SAR image per month of growing season (either RADARSAT-2, RCM or Sentinel-1)
 - Potential for polarimetric data from RCM (ongoing testing)
- 2. Convert data to Sigma naught (σ^0)
- 3. Speckle filter data with 7 x 7 Gamma Map filter
 - Testing changing to larger filter window size
- 4. Rational Polynomial function model (ortho/terrain correction)
 - Testing radiometric correction to incidence angle
- 5. Feed corrected VV and VH images into the classifier with optical imagery and field data
 - Potential for compact polarimetry variables and/or fully polarimetric decomposition variables
 - New classifier may be Random Forest classifier

Summary

- AAFC is a leader in agriculture monitoring and mapping
- Operational Annual Crop Inventory products freely available since 2011 and used by a variety of internal and external users
- Crop Inventory requires a large amount of EO data, including optical and SAR imagery at key times in the growing season
- Crop Inventory stands to benefit from availability of additional SAR imagery in the future including Sentinels, RCM, SAOCOM, NISAR data, with possible improvements in accuracy, delivery time, etc.



Near Ottawa, Ontario 2017 Crop Inventory

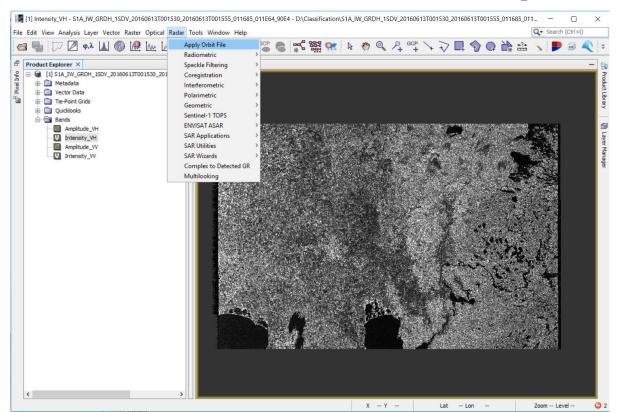


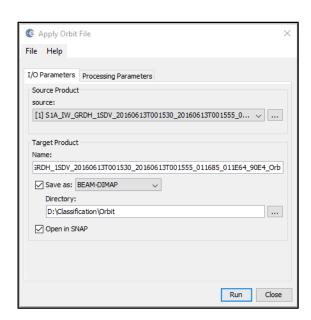
Crop Classification Using SNAP

Sentinel-1 Apply Orbit File 3 Sentinel-1 images over Carman, Manitoba **GRD** June 13, 2016, July 7, 2016 & July 31, Speckle Filter 2016 These images have been pre-processed to co-registration **Geometric Correction** Field data in SNAP Includes normalization to local incidence angle and conversion to 4 shape files: canola, corn, soybeans, Sigma0 wheat Subset Export to .tif Co-registration for use in other classifiers (in Python, R, etc.) **Import Vectors** Random Forest Classifier

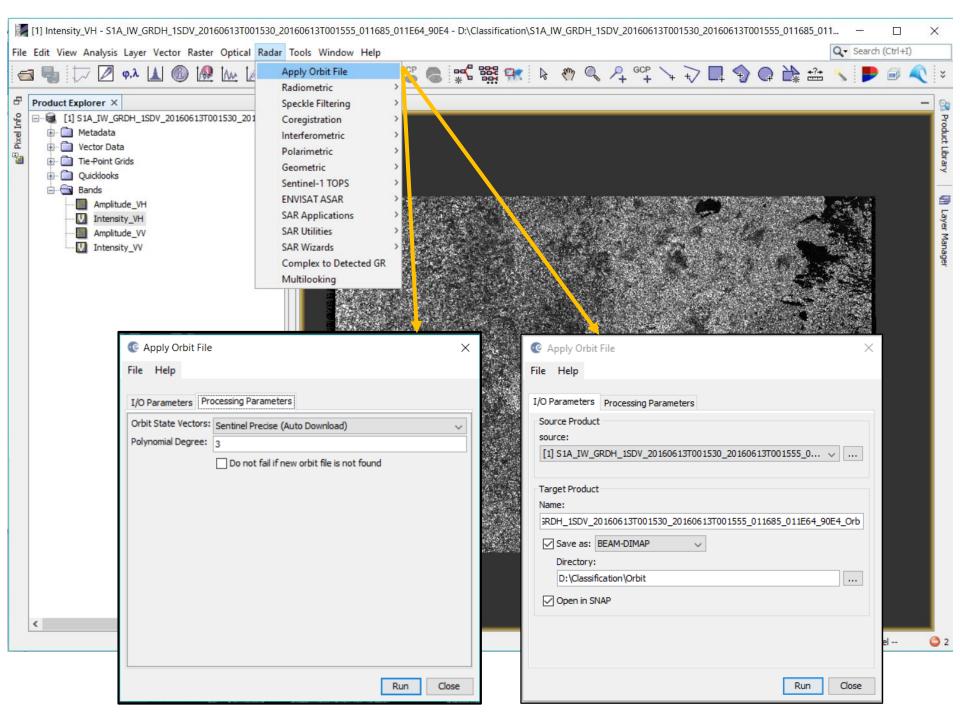
Pre-processing Steps – Apply Orbit File

- 1. Go to Radar Menu >> Apply Orbit File:
 - a) I/O Parameters tab: source → raw image + Target product
 - b) Processing Parameters tab: Orbit State Vectors→
 Sentinel Precise Auto Download; Polynomial
 Degrees → 3
 - c) Click Run and Close window when completed





Apply Orbit File		×
File Help		
I/O Parameters Pro Orbit State Vectors: Polynomial Degree:	Sentinel Precise (Auto Download)	~
	_ DO FIOL FOR IT HEN GLOVE RICE IS THAT TOURING	
	Run	Close

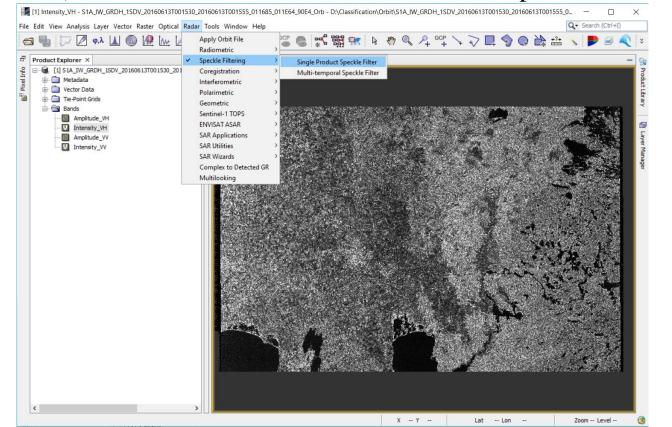


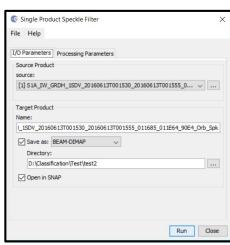
Pre-processing steps – Apply speckle filter (Gamma MAP 7 x 7)

- 1. Go to Radar Menu >> Speckle Filtering >> Single Product Speckle Filter:
 - a) I/O Parameters tab: source → Apply Orbit Image + Target product

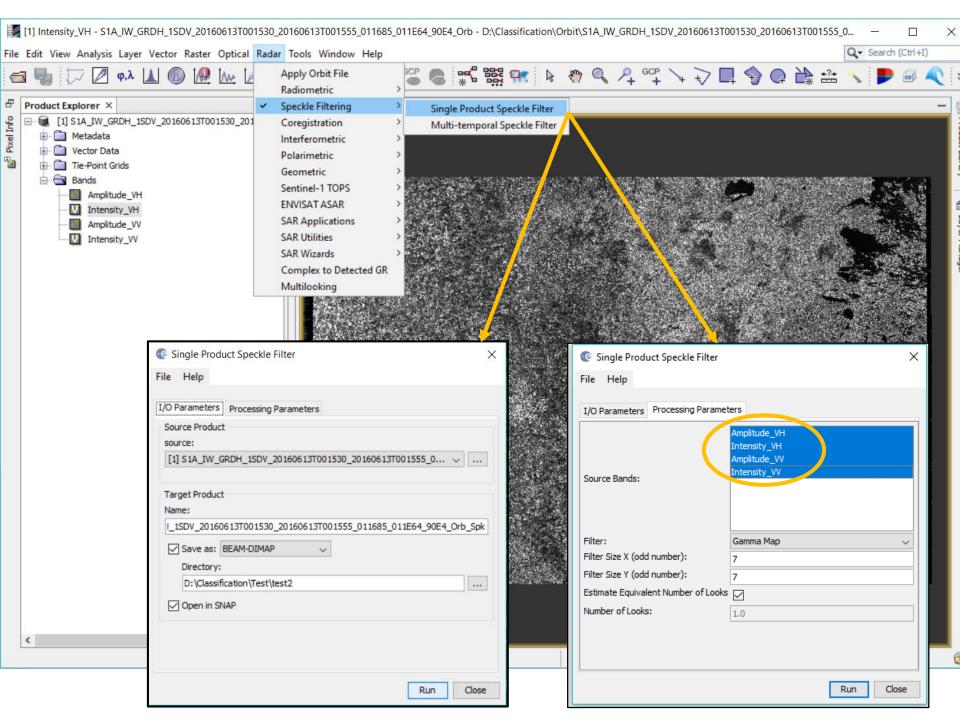
b) Processing Parameters tab: Source Bands \rightarrow Ensure all selected; Filter \rightarrow Gamma Map \rightarrow Window Size \rightarrow 7 x7

c) Click Run and Close window when completed



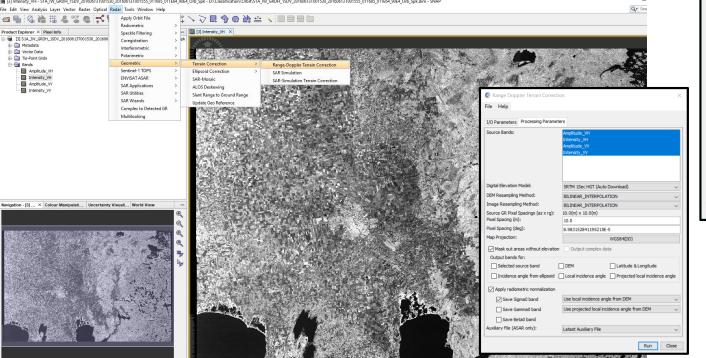


© Single Product Speckle Filter			
File Help			
I/O Parameters Processing Parameters			
Source Bands:	Ampktude_VH Intensity_VH Ampiktude_VV Intensity_VV		
Filter:	Gamma Map 🗸		
Filter Size X (odd number):	7		
Filter Size Y (odd number):	7		
Estimate Equivalent Number of Looks			
Number of Looks:	1.0		
	Run Close		

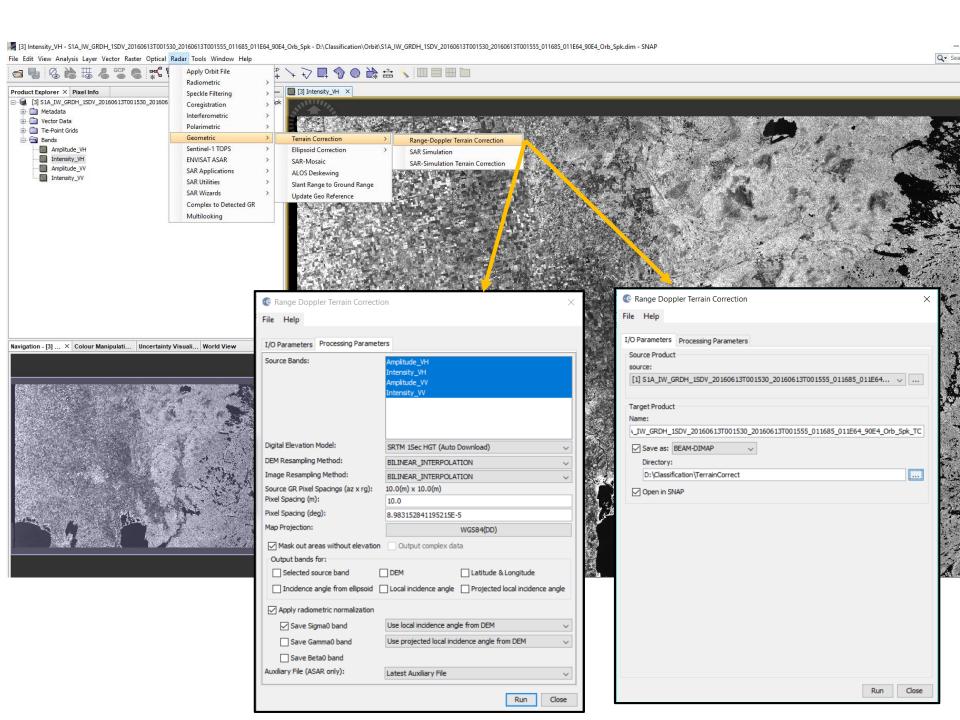


Pre-processing steps - Terrain Correction

- 1. Go to Radar Menu >> Geometric >> Terrain Correction >> Range Doppler Terrain Correction:
 - a) I/O Parameters tab: source \rightarrow Speckle image + Target product
 - b) Processing Parameters tab: Source Bands → Ensure all selected; Digital elevation model → SRTM 1Sec HGT (AutoDownload); Select Apply Radiometric Normalization → Save Sigma0 band → Use local incidence angle from DEM
 - c) Click Run and Close window when completed

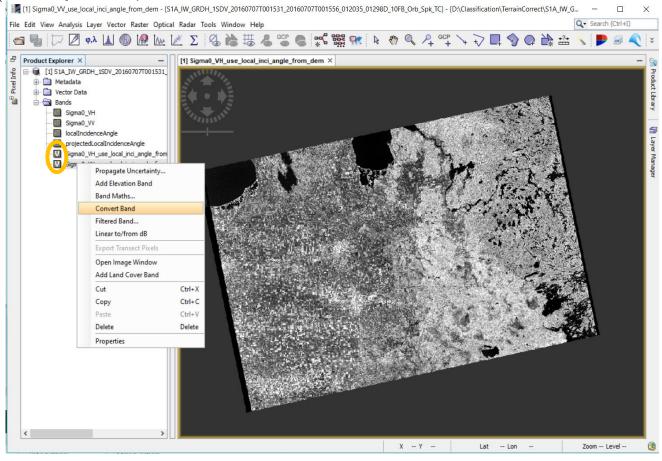


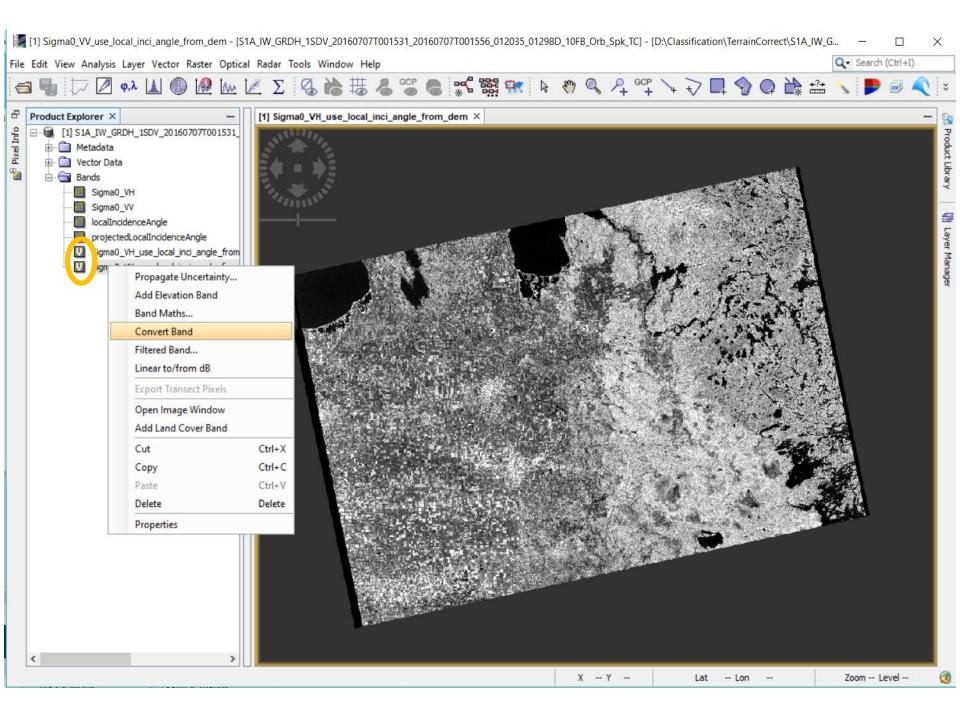




Pre-processing Steps – Convert Virtual Bands

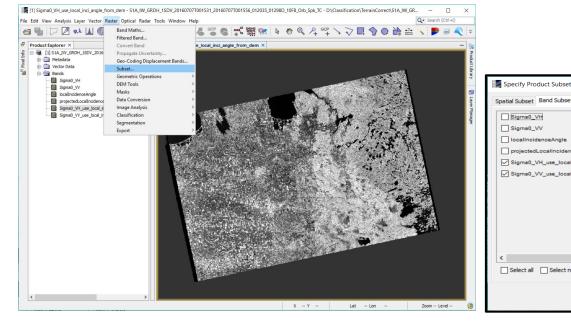
- 1) Right click on 'Virtual' band and select convert
- 2) Can then apply the Subset tool (without saving)
 - a) Saving the converted bands at this point (entire image) takes a long time
 - b) However if you do not save the file the converted band will not be saved but the virtual band (lookup table) will still be available

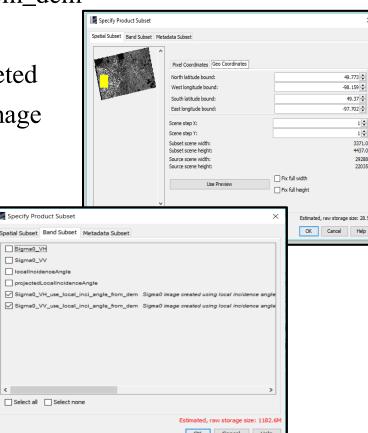


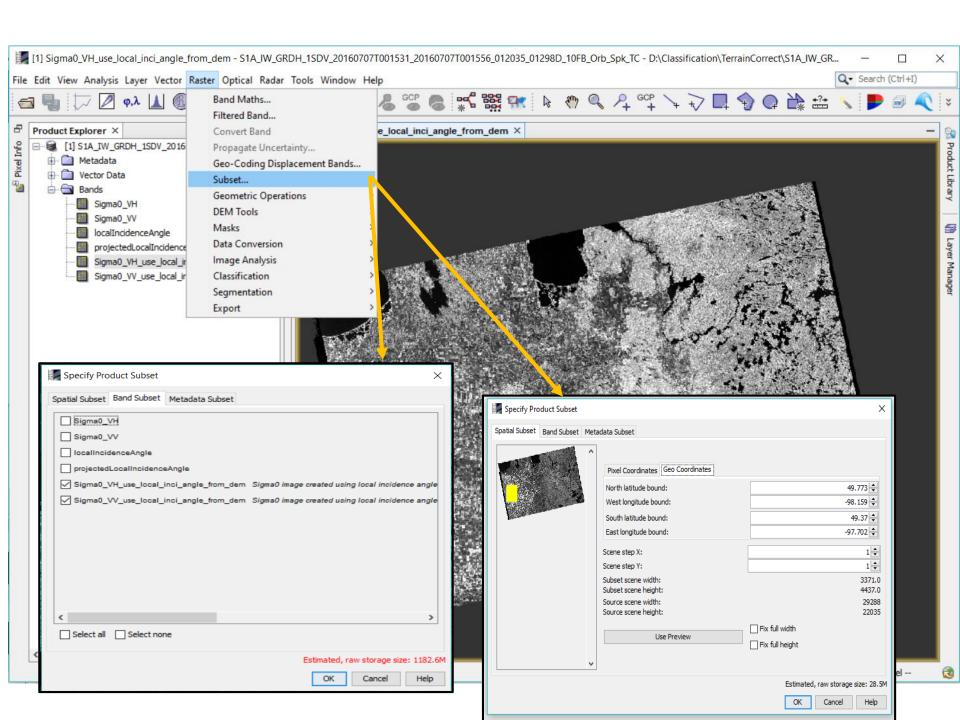


Pre-processing Steps – Subset Raster to AOI (Per Image)

- 1) Go to Raster Menu >> Subset:
 - a) Spatial Subset tab → enter the upper left and lower right coordinate under geo coordinates
 - b) Band Subset → select bands you wish to subset
 "Sigma0 VH/VV_use_local_inci_angle_from_dem"
 - c) Metadata Subset: leave as default
 - d) Click Okay and Close window when completed
 - e) You will need to "Save" the newly subset image



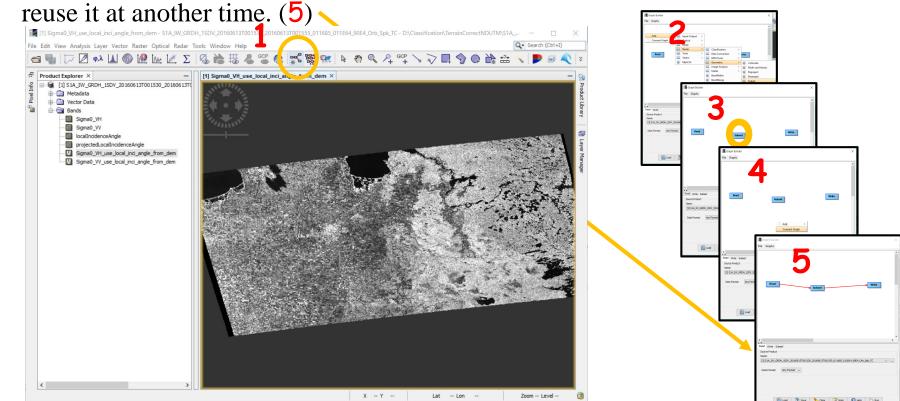


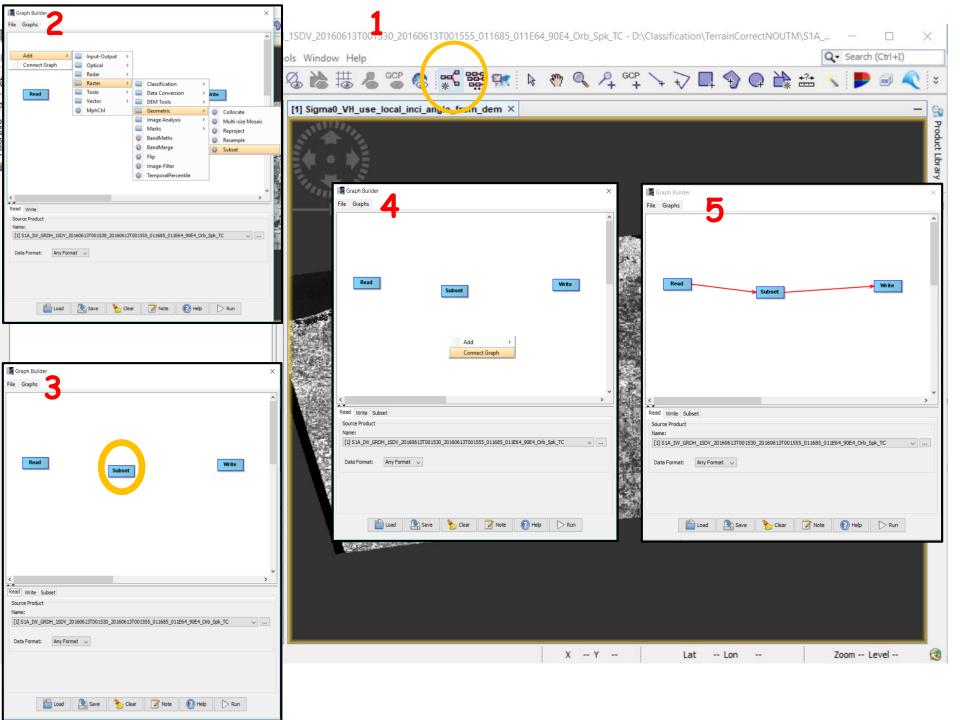


How to Build a Graph in SNAP

- 1) Go to Toolbar and select >> (1) Graph Builder:
 - a) In blank space with 'Read' and 'Write' modules, right click and select 'All' a list of the menus appears (2)
 - b) Choose a function you need (e.g. "Subset") (3)
 - c) Once you have all modules you need, right click in blank space again and select "Connect Graph" (4)

d) Red arrows appear and connect the modules of your process; Save your graph to

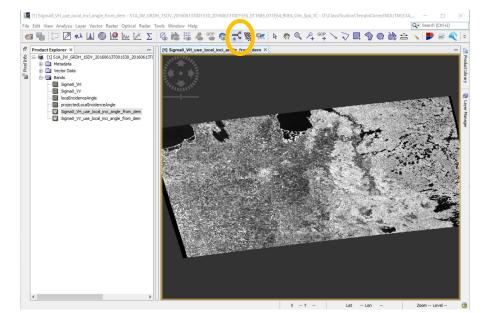


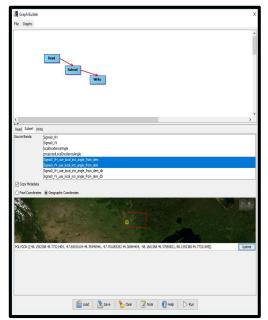


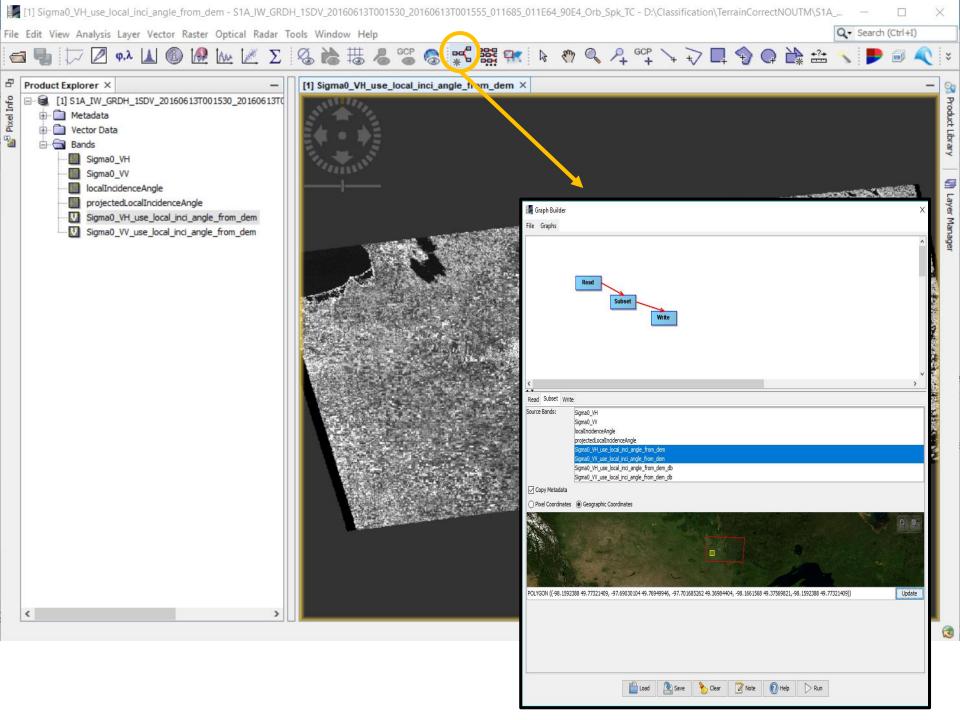
Pre-processing Steps – Subset Rasters to AOI

(in Graph)

- Subset tool in a graph format is more precise
- Can write subset to new file immediately
- Can bulk process with created graph
- In either case, smaller files will have faster processing times*
- Subset tab → select bands to be subset
- 2) Subset tab → click on Geographic Coordinates and add a well-known text polygon (5 coordinates, (last coordinate is same as first to close polygon))
- 3) Write Tab → Ensure new folder is selected for subset files to be saved to

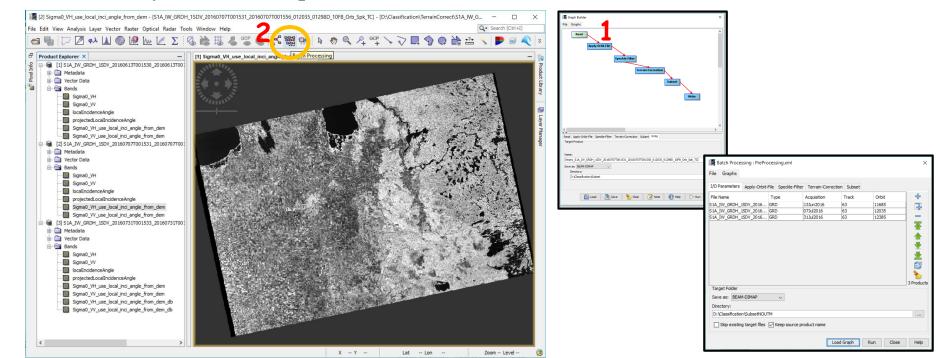


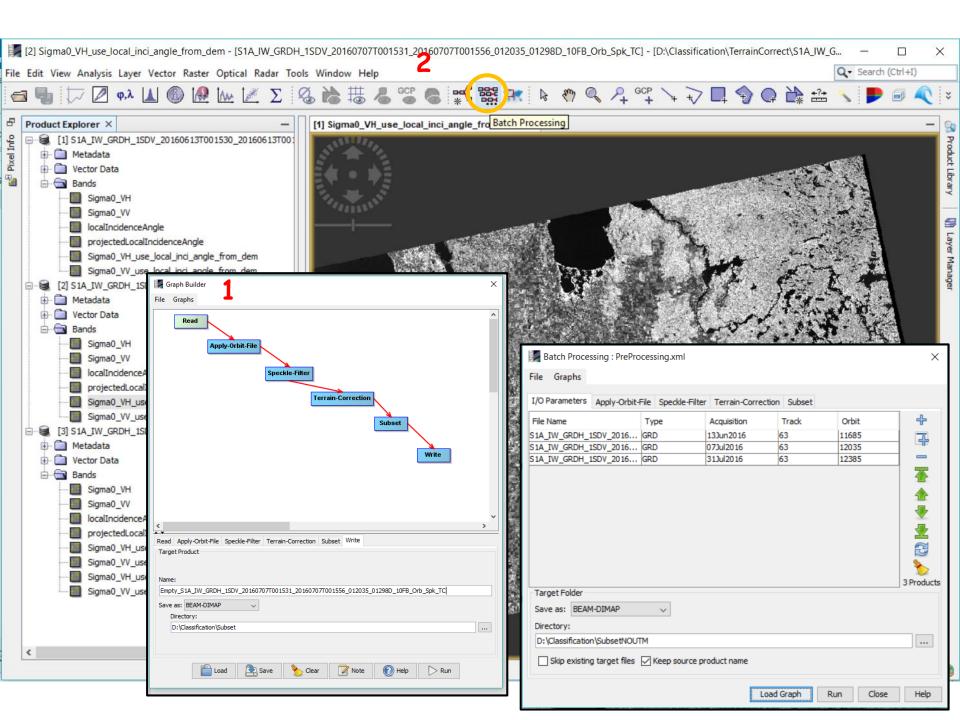




Bulk processing – Create a graph and Pre-process All Images to Coregistration

- Can process multiple files (bulk processing) and run over night
- 1) Create a processing graph (this example contains all pre-processes to 'Subset') (1)
- 2) Batch Processing (2)
 - I/O Parameters tab: source → Click Plus sign with line over top adds all open imagery → Click Revolving Arrows refreshes metadata → Load Graph to load created processing graph
 - Once you load the graph, tabs of all processing modules will appear
 - Ensure you select all parameters under each tab





Pre-processing Steps - Co-registration

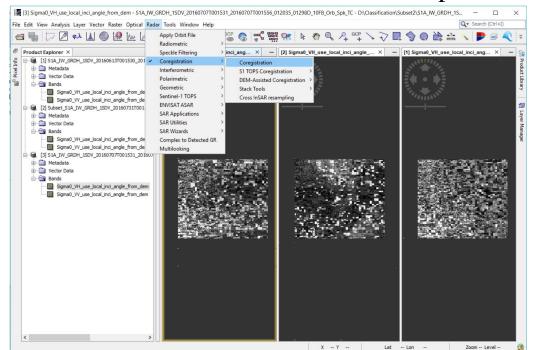
Spatial alignment of images acquired on June 13th, July 7th and July 31th, 2016

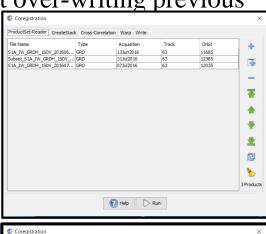
- 1. Go to Radar Menu >> Coregistration >> Coregistration:
 - a) ProductSet-Reader: Click Plus sign with line over top adds all open imagery
 → Click Revolving Arrows refreshes metadata
 - b) Create Stack: Resampling Type → Bilinear_Interpolation → Click Find Optimal Master

c) Other tabs: leave as default; ensure Write folder is not over-writing previous

files

d) Click Run and Close window when completed

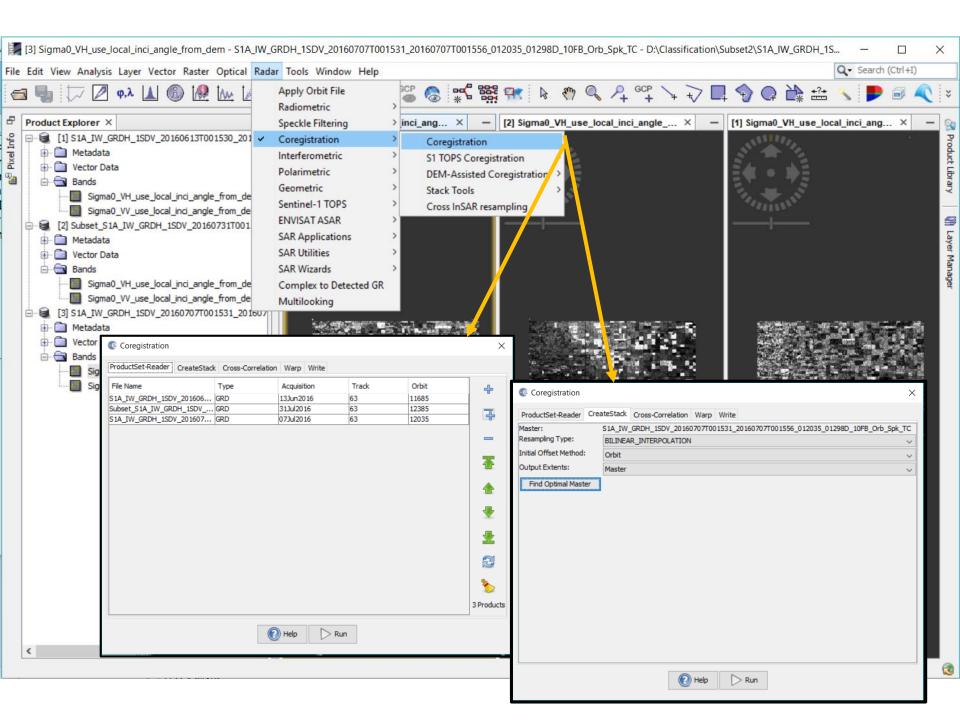




S1A_IW_GRDH_1SDV_20160707T001531_20160707T001556_012035_01298D_10FB_Orb_Spk

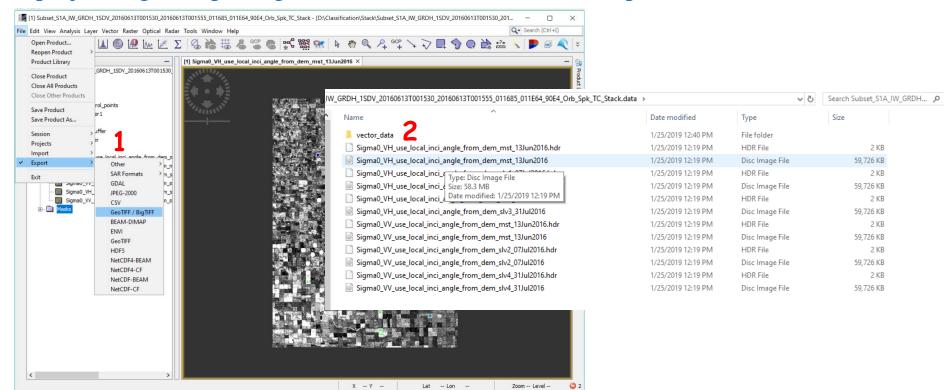
ProductSet-Reader CreateStack Cross-Correlation Warn Write

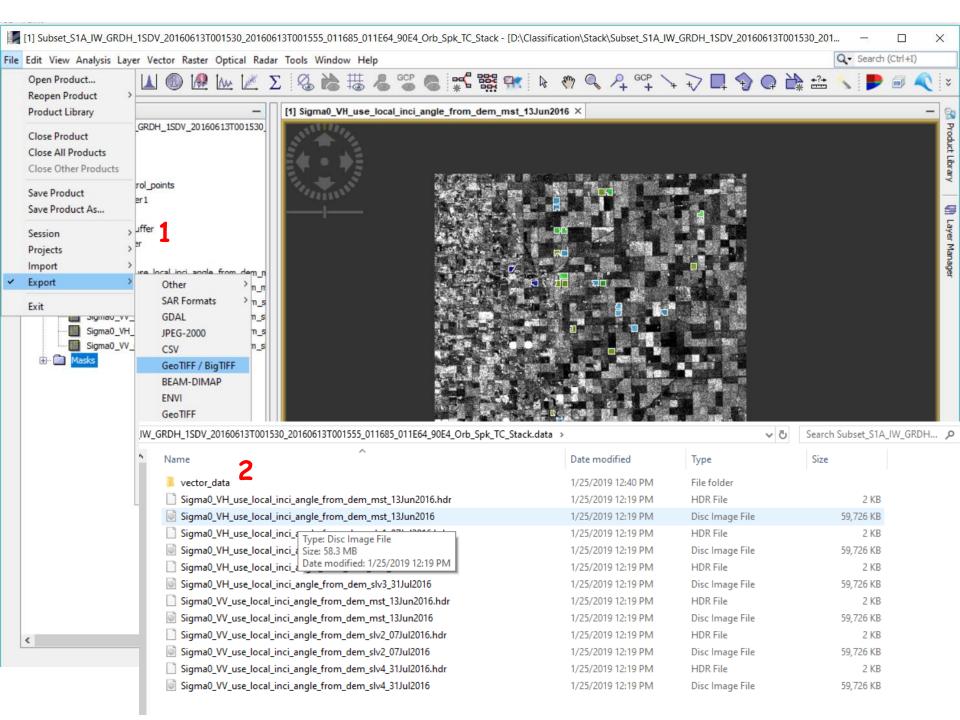
Find Optimal Master



Exporting Data Stacks Out of SNAP

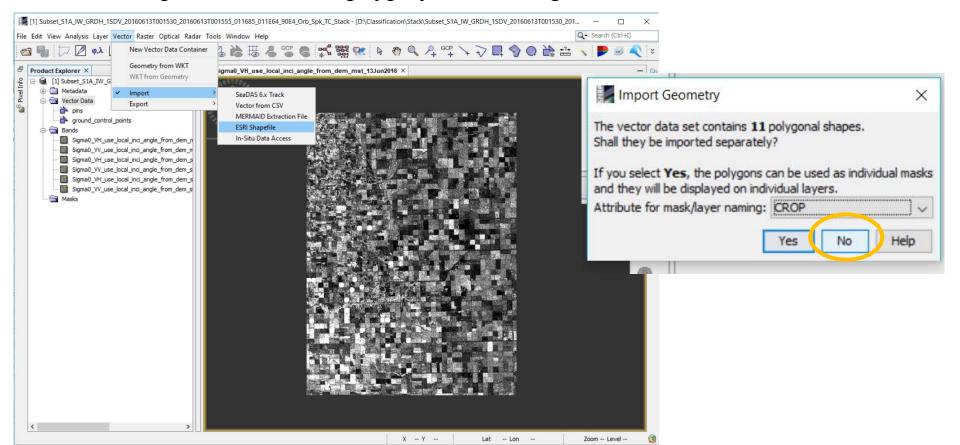
- Limited functionality in SNAP; opportunity to customize classifiers
- Two ways to 'export'
 - $SNAP \rightarrow File \rightarrow Export (1)$
 - Windows File Explorer → use the .img files in the associated BEAM DIMAP folder (2)
- Can use .tifs in R, Python etc. (RandomForest, R https://cran.r-project.org/web/packages/randomForest/randomForest.pdf)

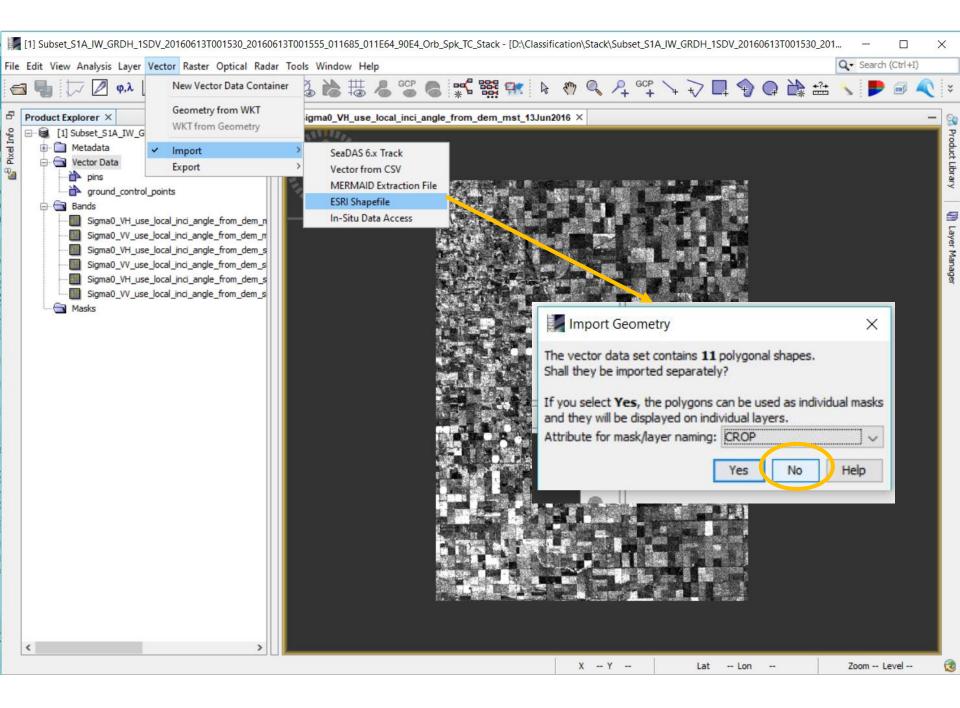




Adding Field Data in SNAP

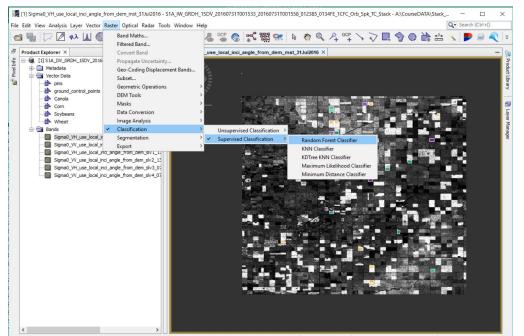
- Must import each crop type individually
- 1) Go to Vector Menu >> Import >> Esri Shapefile:
 - a) Import Geometry: Choose attribute for mask/layer naming→
 - b) When asked to import as individual masks NO
 - c) Complete for each crop type you are using

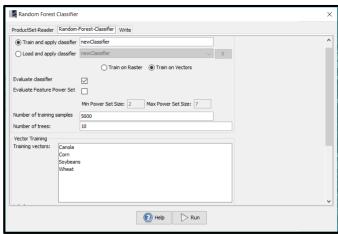


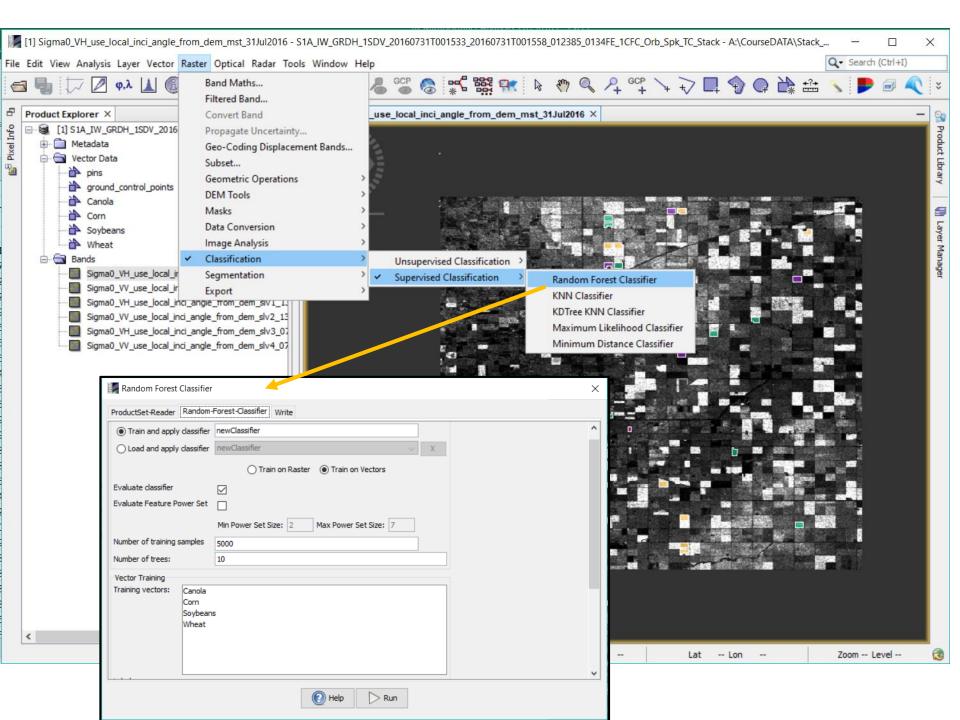


Classification in SNAP

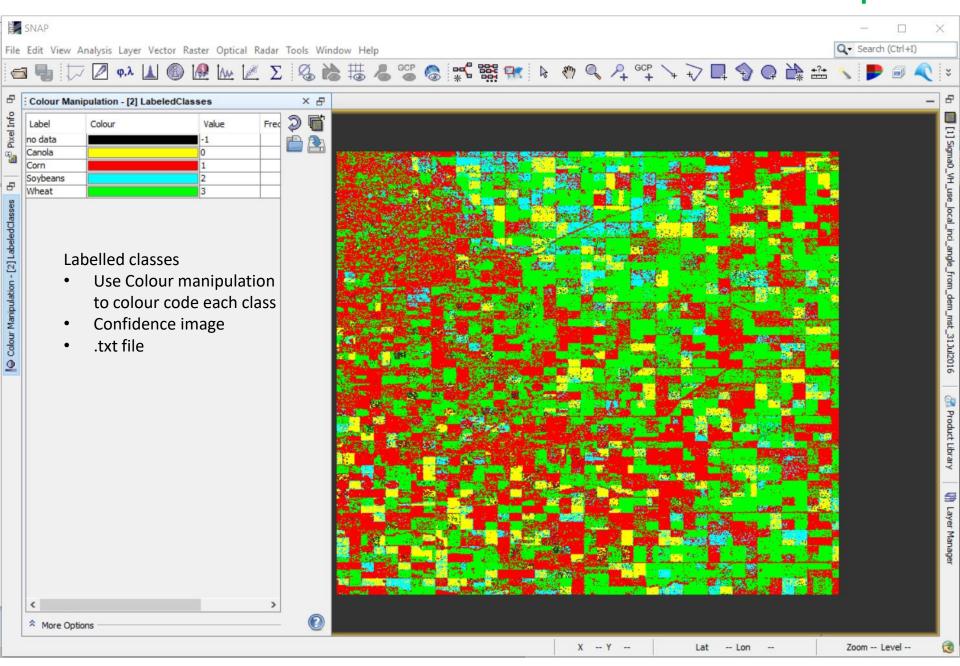
- Using the stack of EO data along with added vectors of field data to classify data using Random Forest
- 1) Go to Raster Menu >> Classification >> Supervised Classification >> Random Forest Classifier:
 - a) Random Forest Classifier: Can leave all as default
 - b) During research you can assess how changing the parameters changes results
- 2) Number of training samples are number of training pixels to include from class masks
- 3) Number of trees are the number of trees to be grown with each individual tree similar to Decision Tree classifier







Classification in SNAP – Random Forest Outputs



Classification in SNAP – Random Forest Output – Confidence Image

