



Vegetation Health Development and Applications Transition from AVHRR to JPSS

Presenters:

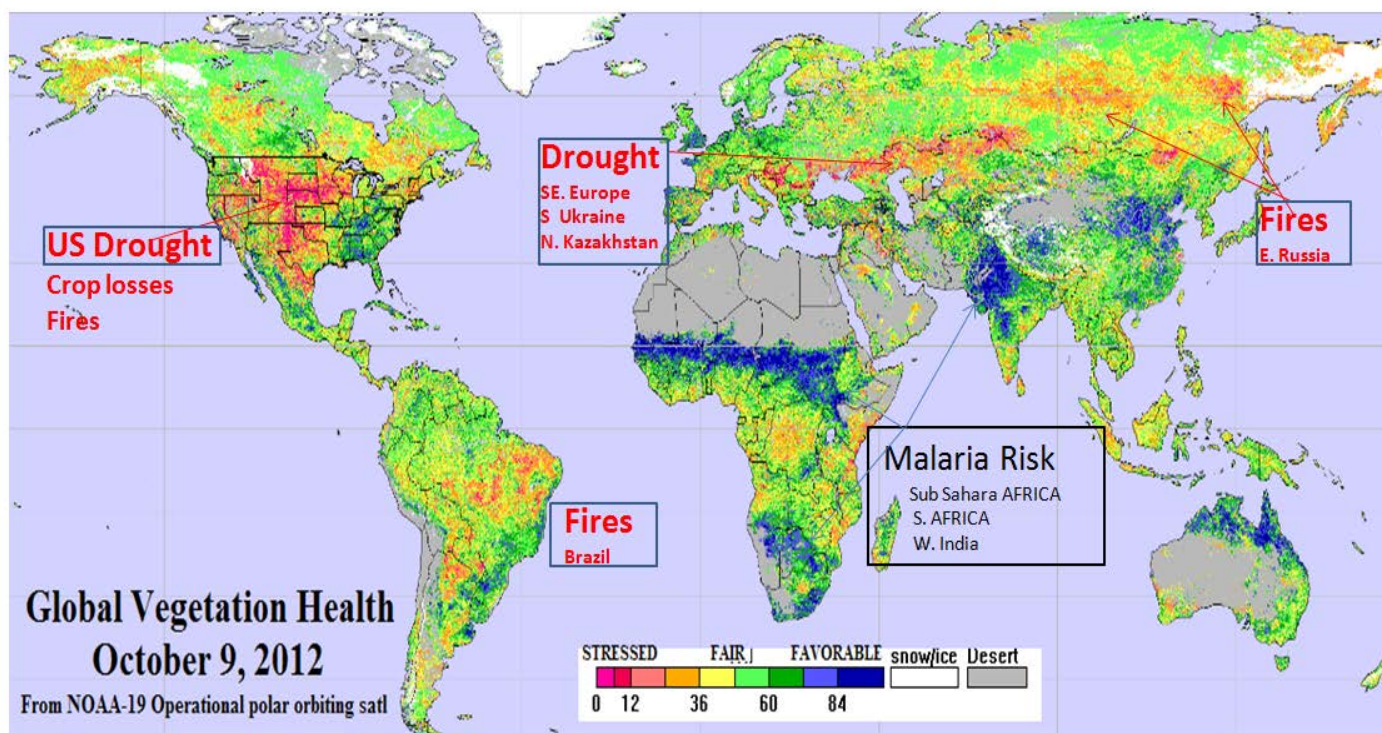
Felix Kogan, Wei Guo & Wenzhe Yang



Vegetation Health Index & Applications, Aug, 2017

Vegetation Health (VH) is an indicator of satellite-based vegetation response to weather (mostly P and T)

From AVHRR/NOAA-19 Operational Polar Orbiting Satellite



<http://www.star.nesdis.noaa.gov/smcd/emb/vci/VH/index.php>

Vegetation Health
Drought

Area

Intensity

Duration

Start/End

Impacts

Moisture Stress

Thermal Stress

Healthy Vegetation

Crop/pasture Prod.

Fire Risk

Soil Saturation

Mosquitoes Diseases

Land Greenness

Landslides

Food security

Climate Change

Land Change



GTS (WMO-based) weather station network

One weather station covers

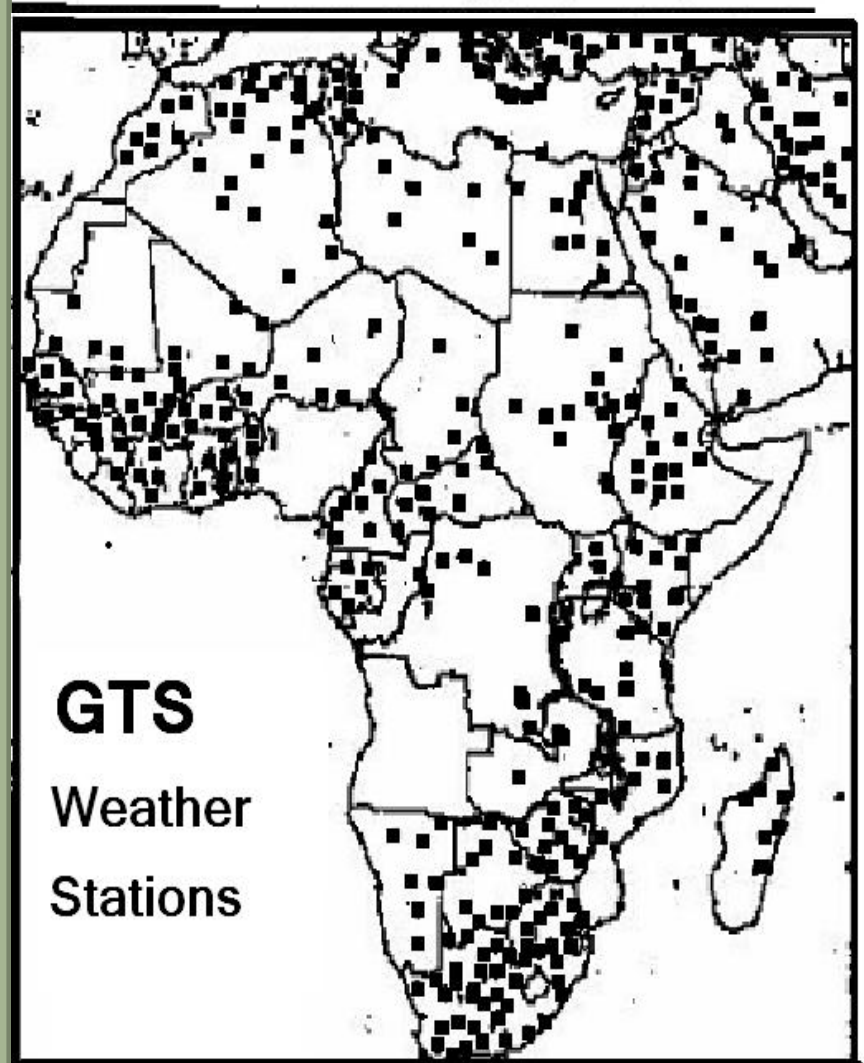
GLOBAL -10,000 sq. km

Africa - 31,000 sq. km

NOAA satellites cover

AVHRR: 4 & 16 km²

VIIRS: 0.5, 1 & 4 km²





Vegetation Health

(1) Normalized Difference Vegetation Index (NDVI)

VIS (Ch1) and NIR (CH2)

Brightness Temperature (BT)

IR (CH4)

(2) NDVI & BT Climatology



Vegetation Health (VH) Indices

Vegetation condition index (VCI), values 0 - 100

$$VCI = (NDVI - NDVI_{min}) / (NDVI_{max} - NDVI_{min})$$

NDVI_{max}, and NDVI_{min} – climatology (1981-2000 maximum and minimum NDVI for a pixel;

MOISTURE

Temperature condition index (TCI), values 0 - 100

$$TCI = (BT_{max} - BT_{min}) / (BT_{max} - BT_{min})$$

NDVI_{max}, and NDVI_{min} – climatology (1981-2000 maximum and minimum NDVI for a pixel

THERMAL

Vegetation Health Index (VHI), values 0 – 100

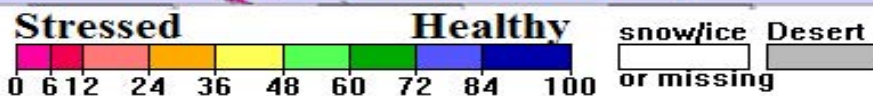
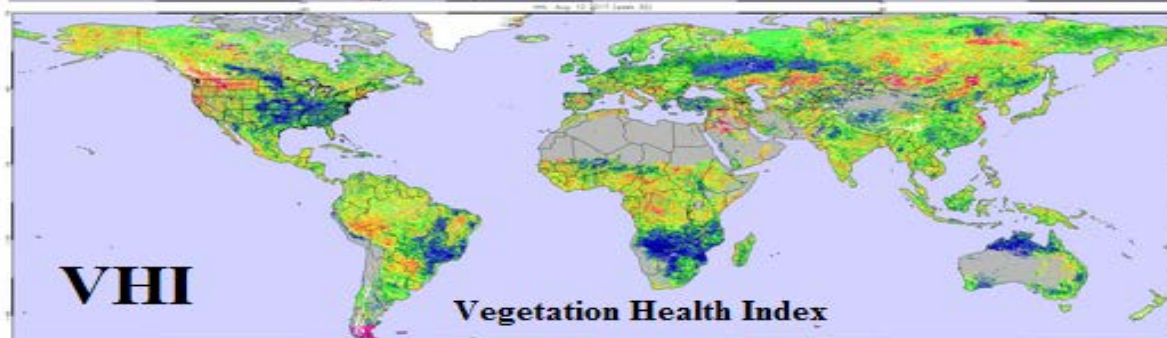
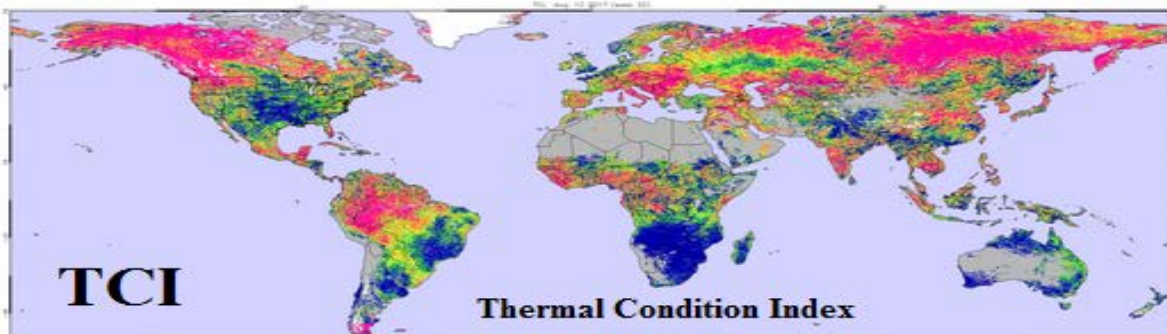
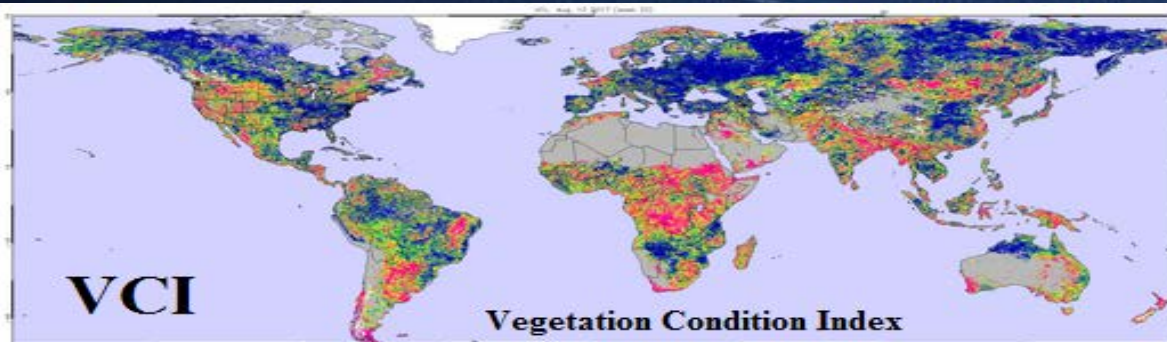
$$VHI = a * VCI + (1 - a) * TCI$$

VEG.
HEALTH

0 – indicates extreme stress; **100** – indicates favorable conditions



VCI, TCI & VHI Aug 11, 2017

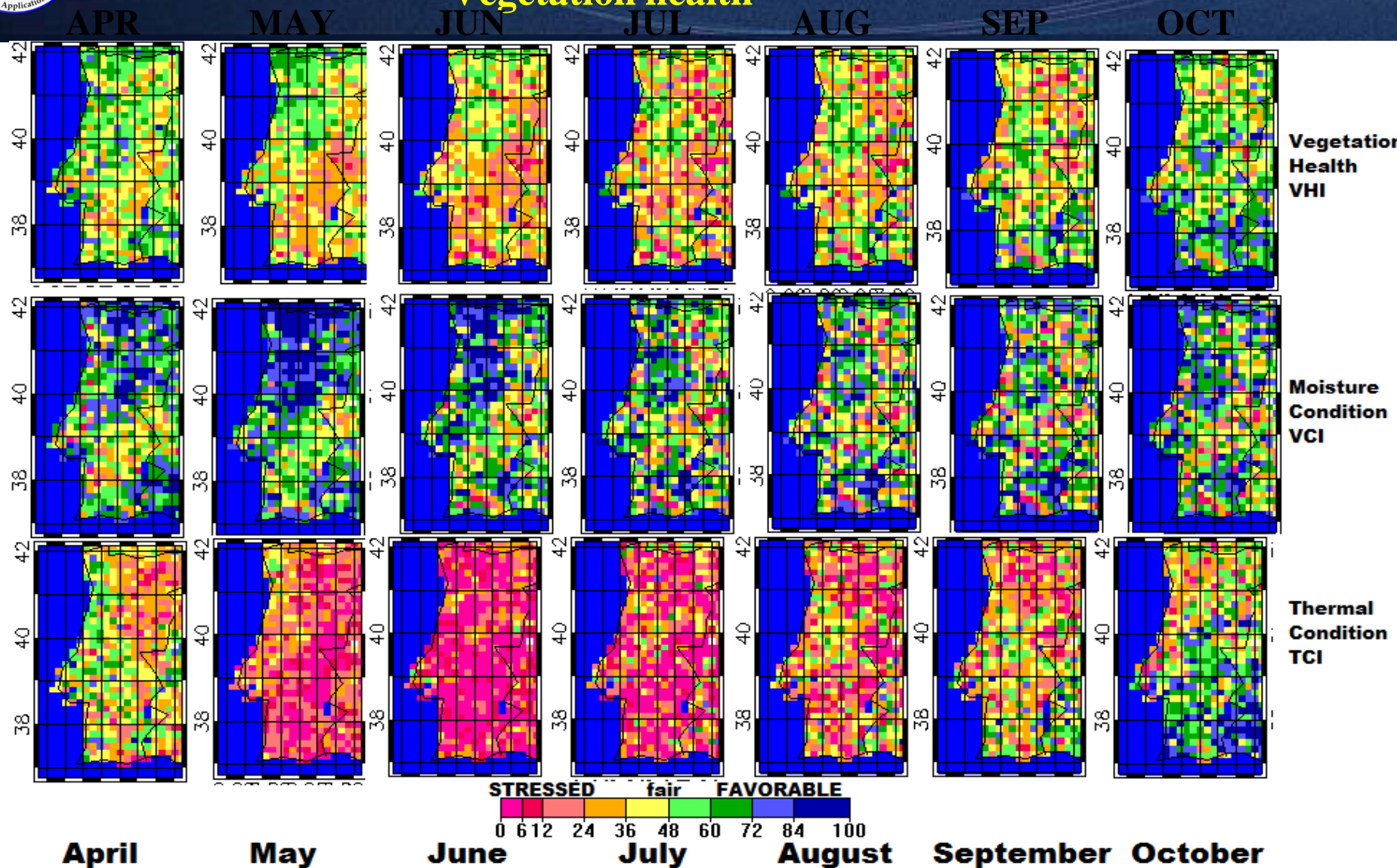


World Vegetation Health August 11, 2017



2015 PORTUGAL

Vegetation health



Vegetation Health 2015, PORTUGAL

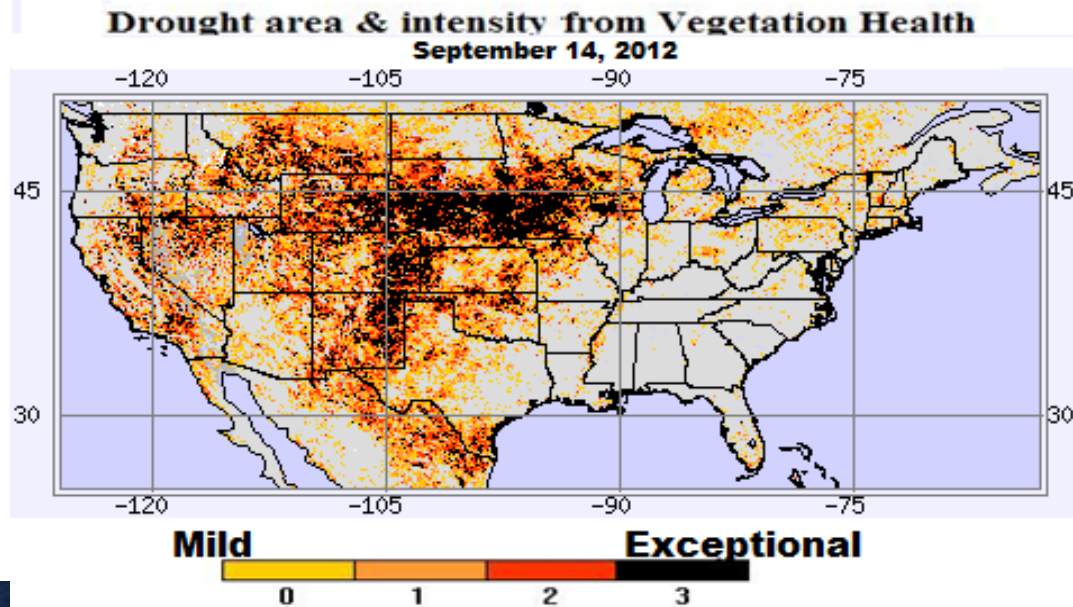
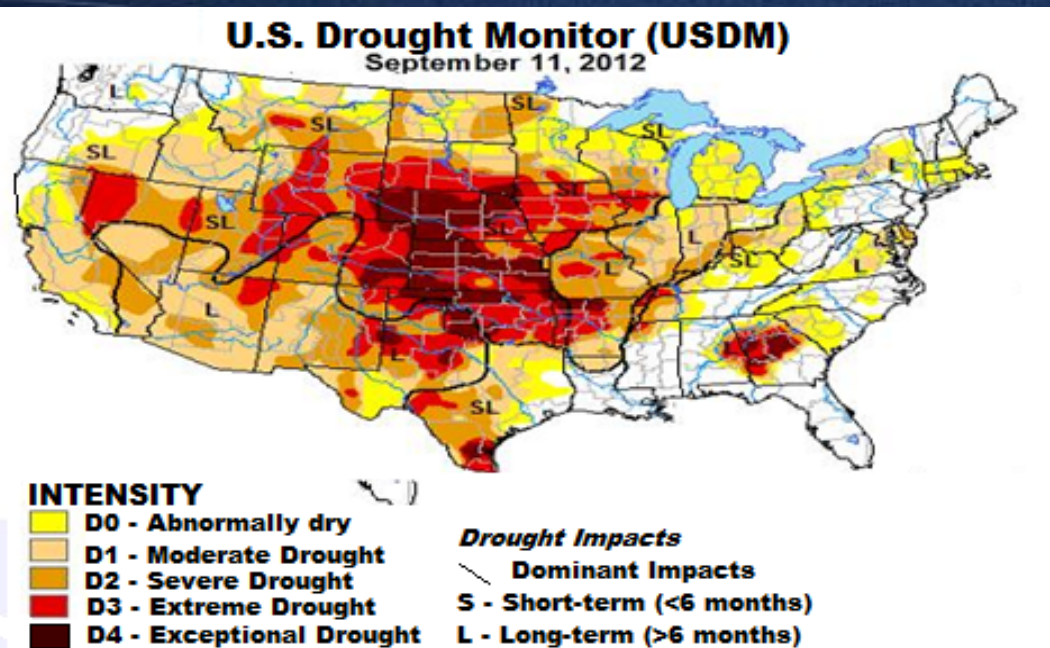


Vegetation Health (VH) Data

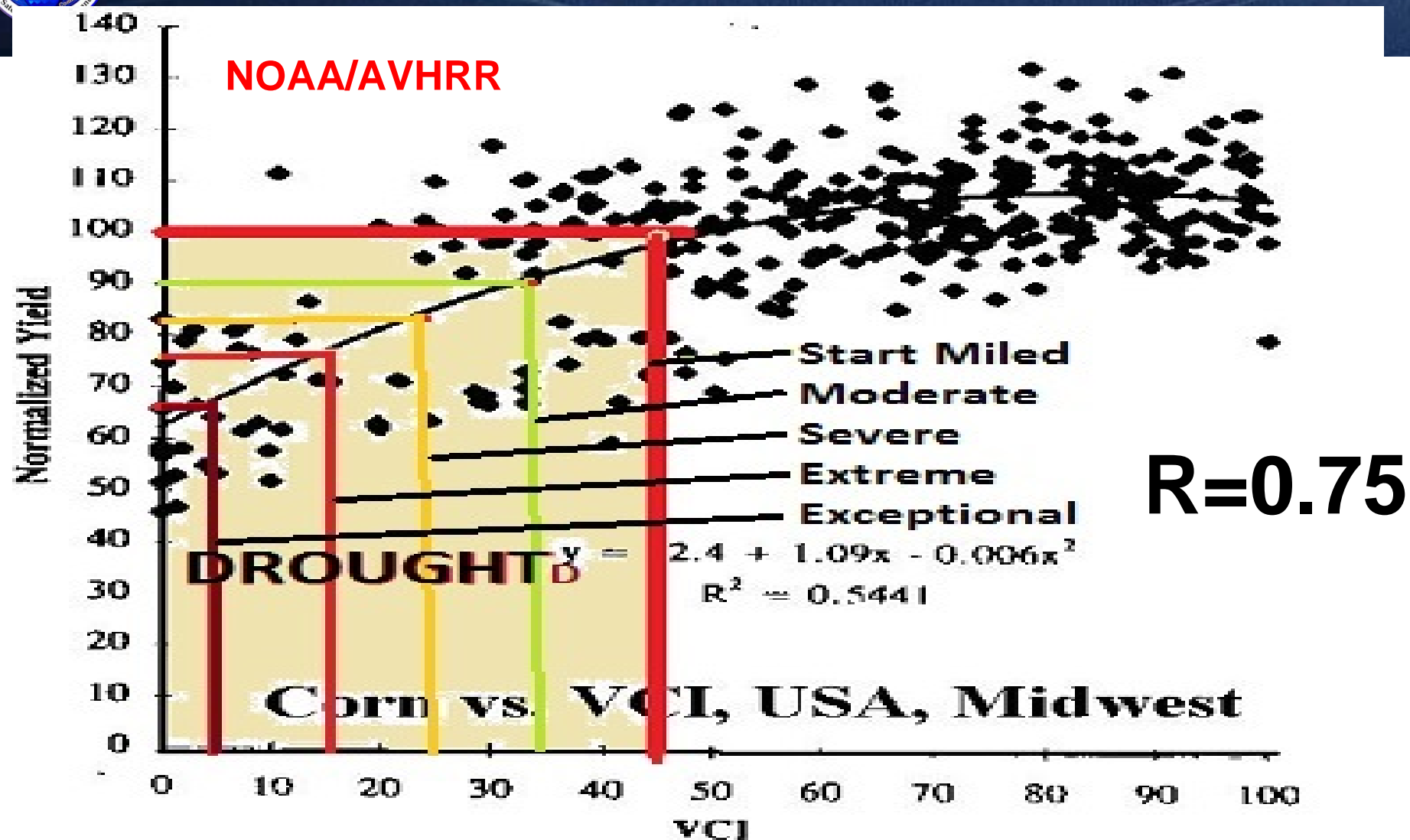
- **NOAA/AVHRR (37y)** 16 sq. km
4 sq. km
- **SNPP/VIIRS (6 y)** 4 sq. km
1 sq. km
0.5 sq. km



Drought USDM & VH



Corn Yield vs AVHRR-based Vegetation health indicator (VCI) Midwest, USA

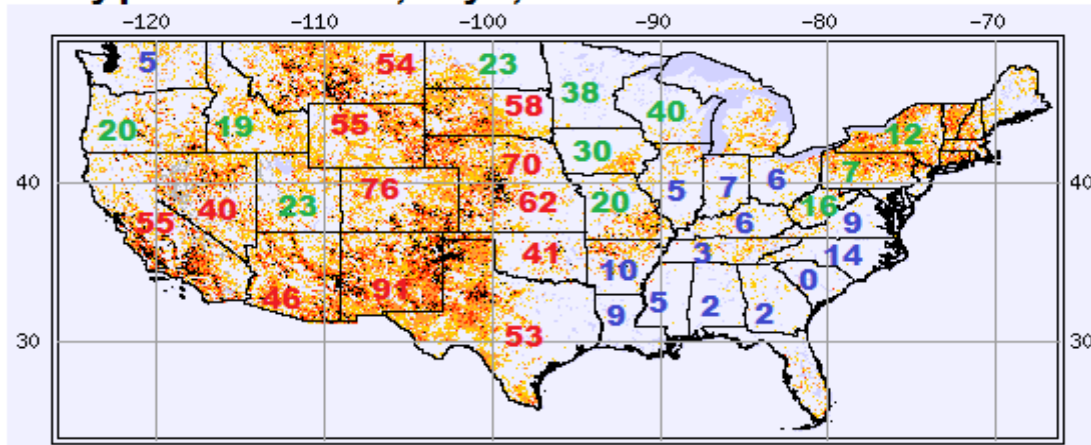


Correlation of Corn Yield Anomaly with VCI and Drought Classification

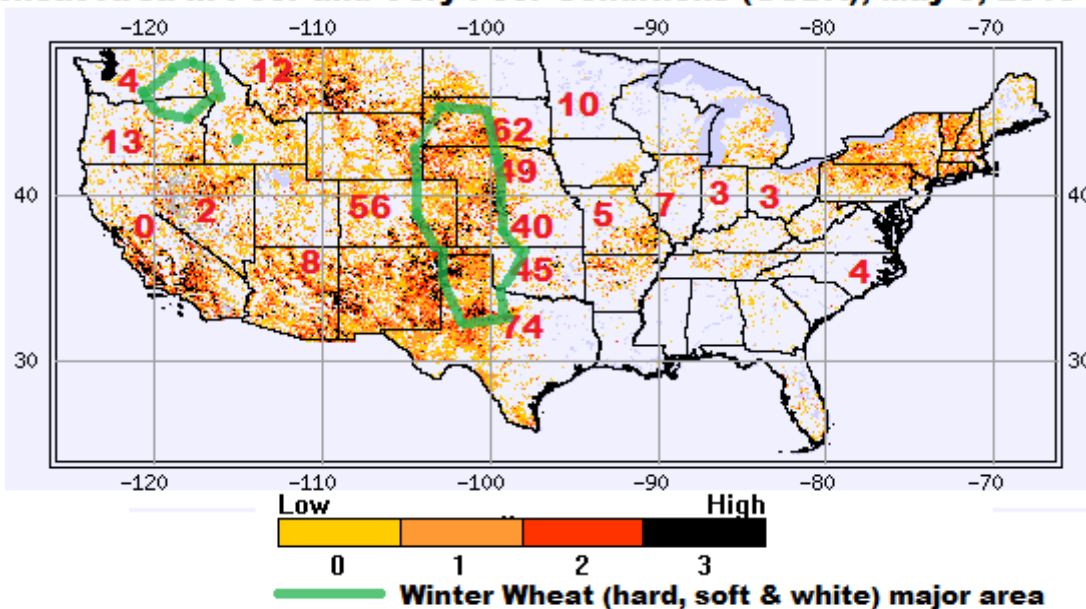
VH-based drought stress & USDA pasture & winter wheat condition, May 6, 2013



VH-based Drought Stress & % state with pasture & range land in poor & very poor conditions, May 6, 2013

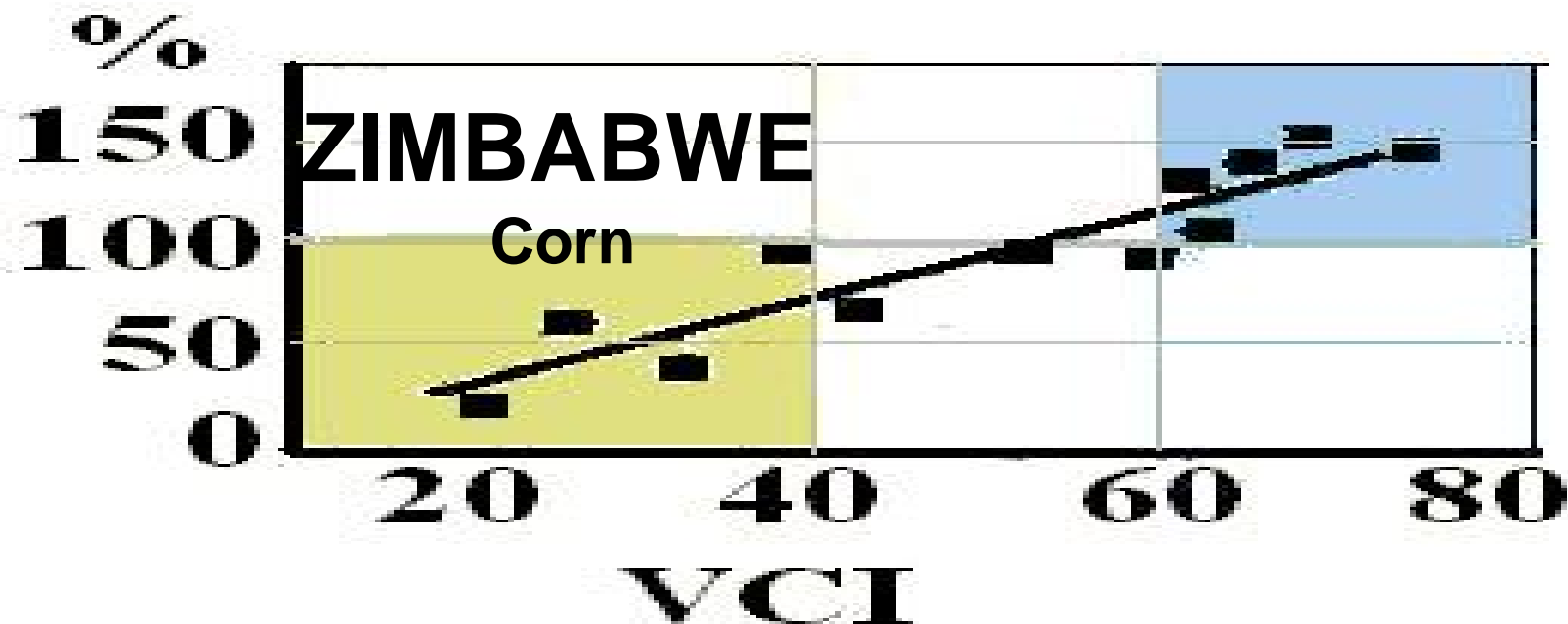
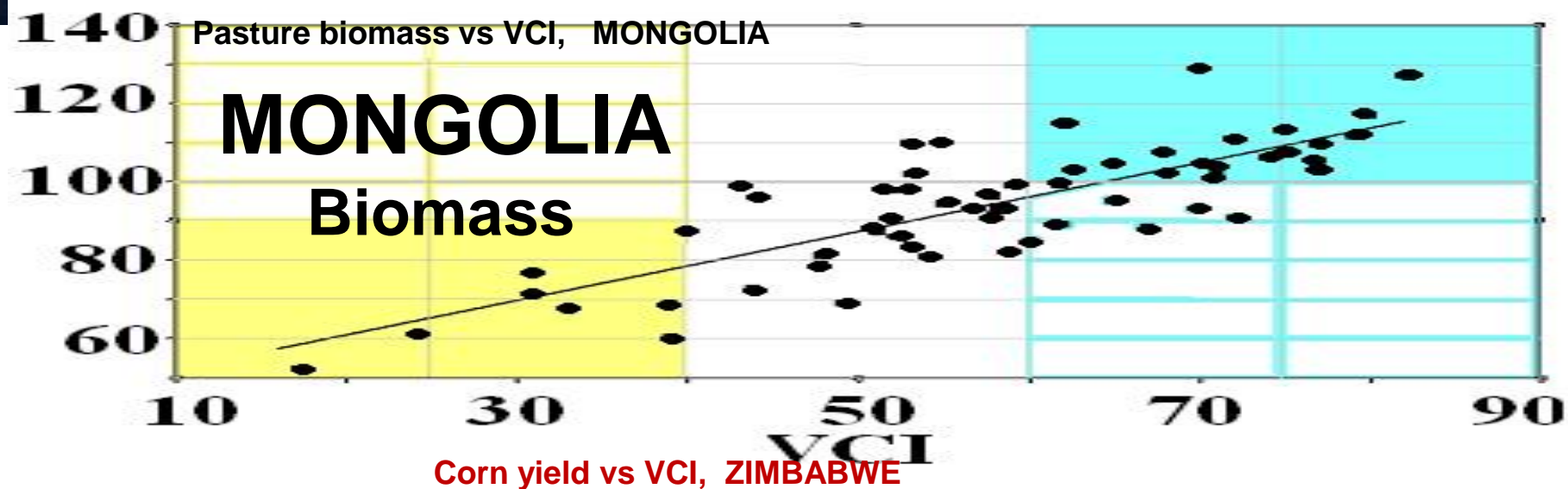


VH-based Drought Stress (NOAA), May 6, 2013 & Percent Winter Wheat Area in Poor and Very Poor Conditions (USDA), May 5, 2013





VH-Biomass & Corn Yield Modeling & Prediction

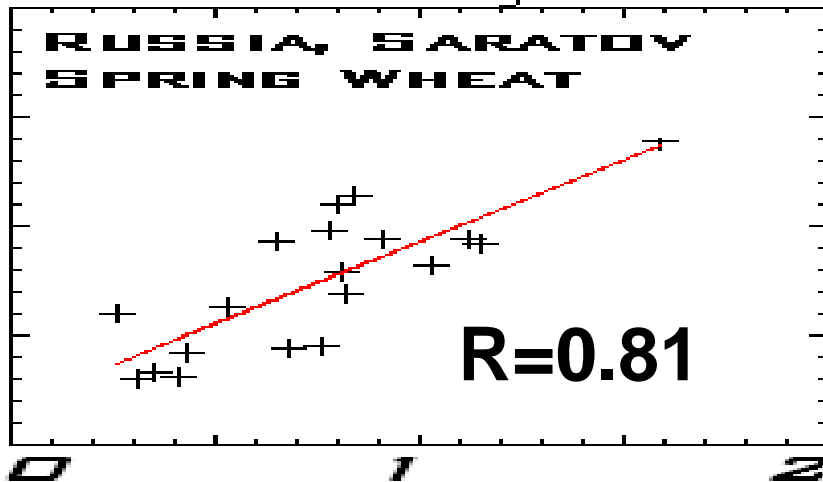




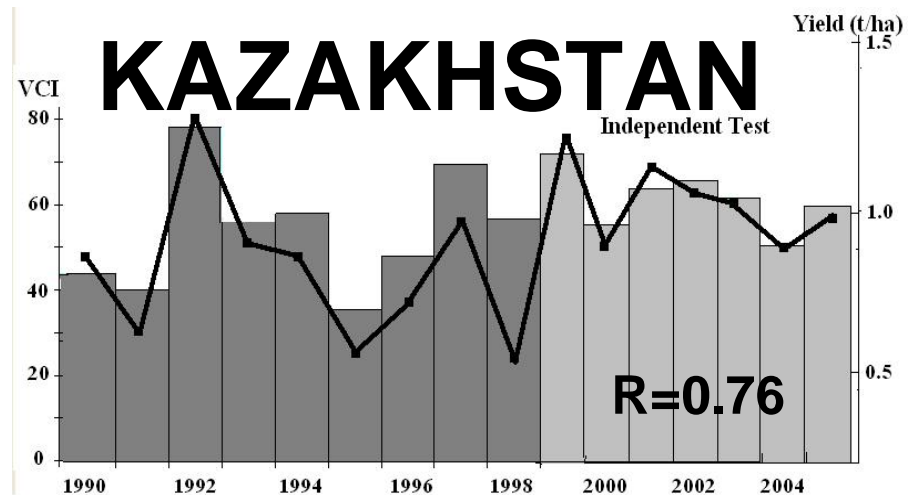
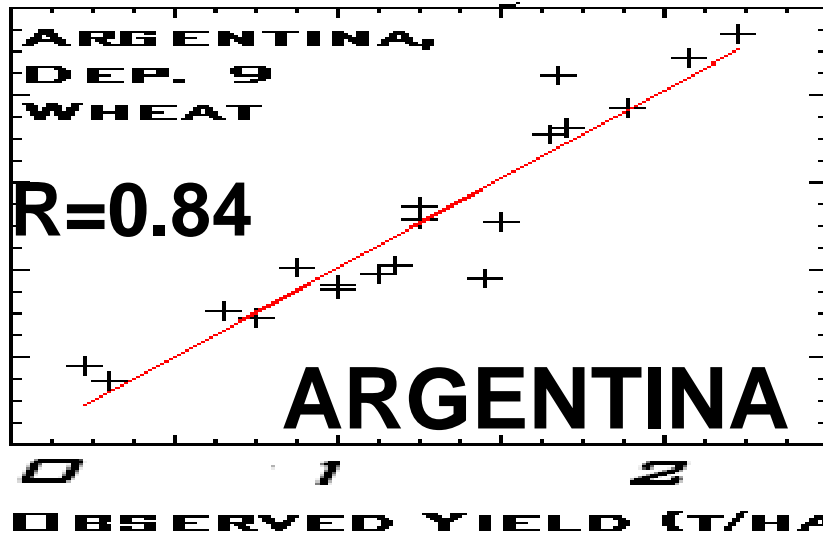
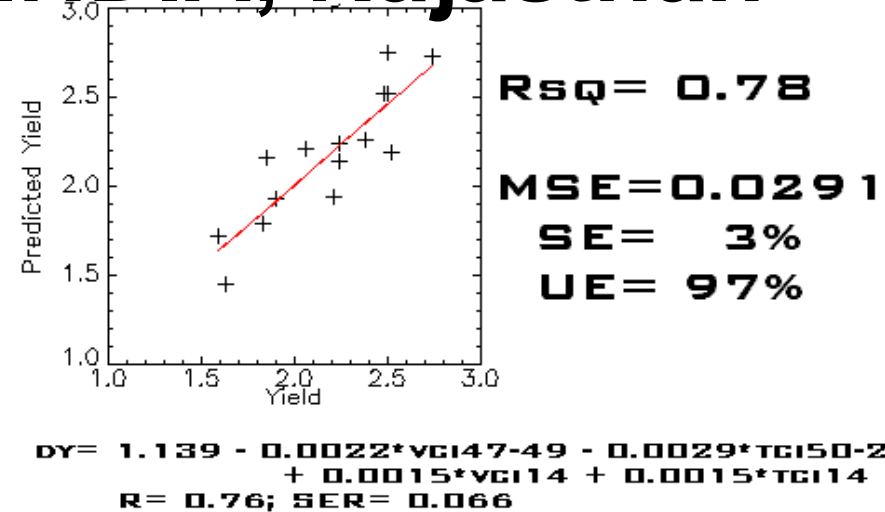
VH Predicted vs Observed Wheat Yield

R=0.81

RUSSIA



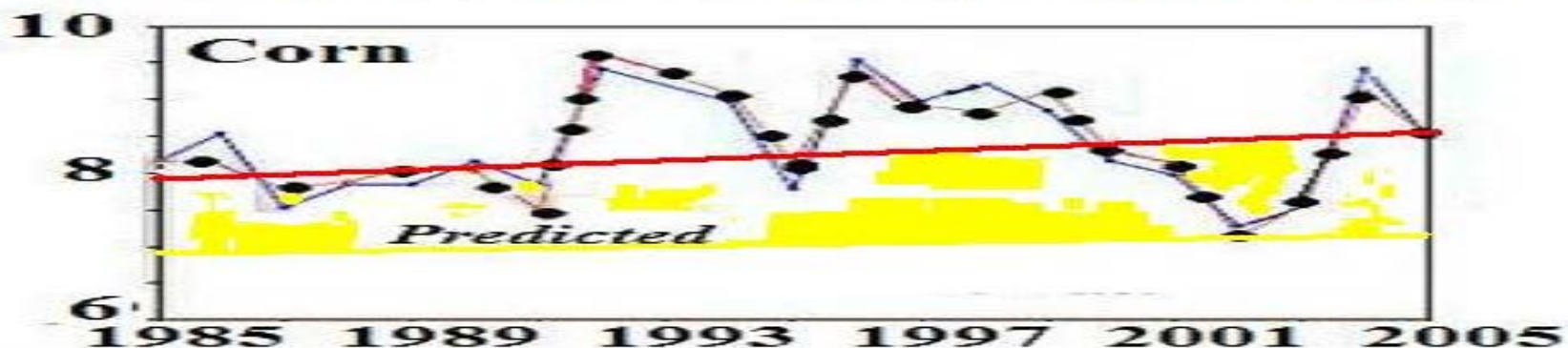
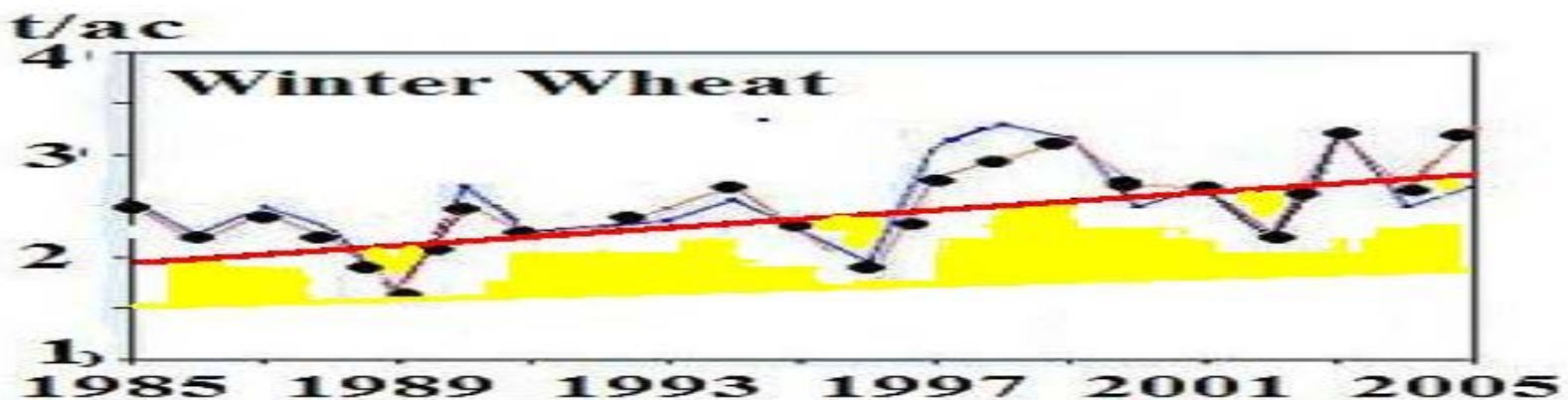
INDIA, Rajasthan



Mean Yield (bars) and VCI, Kazakhstan, Spring Wheat



VH-Crop Losses Prediction: USA, Kansas



Observed & Predicted Yield, USA, Kansas

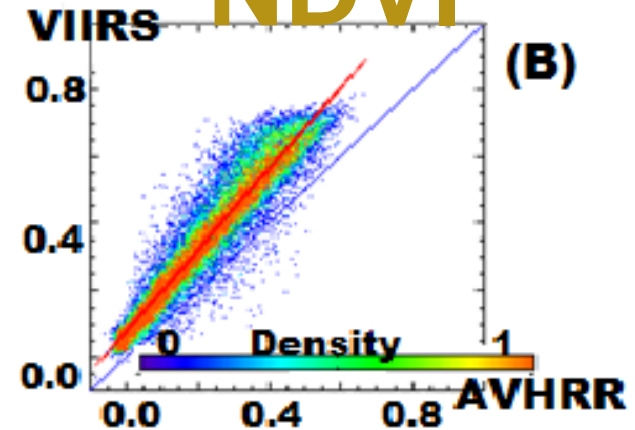
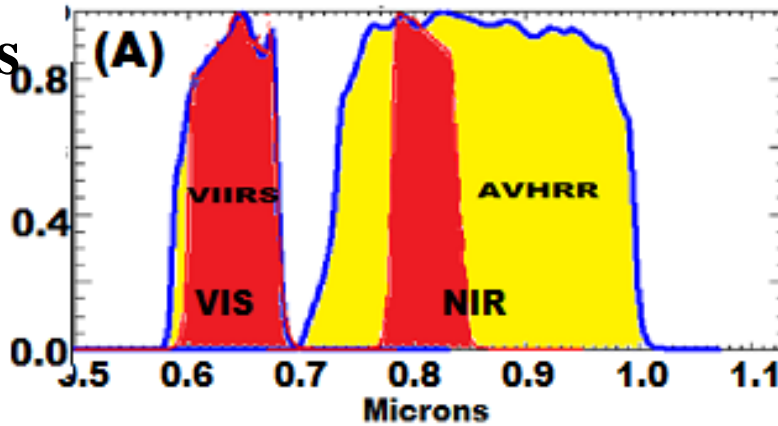


Channels: VIIRS versus AVHRR

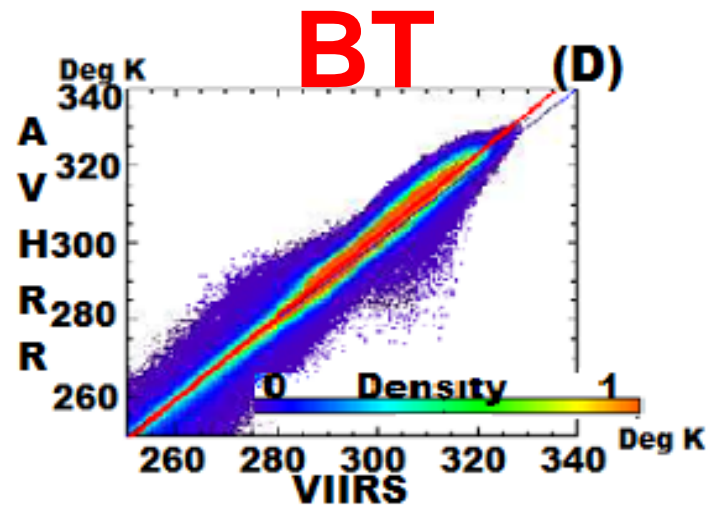
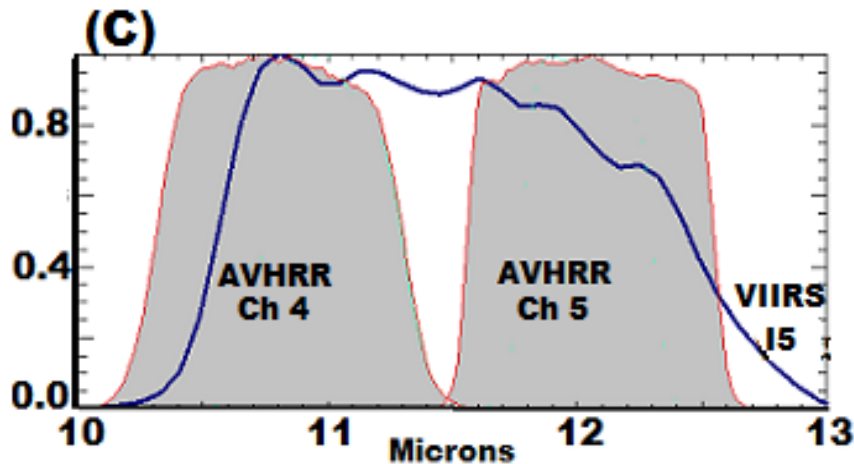
Climate data records problems

NDVI

Channels



Normalized Difference Vegetation Index (NDVI)



Brightness Teperature (BT)



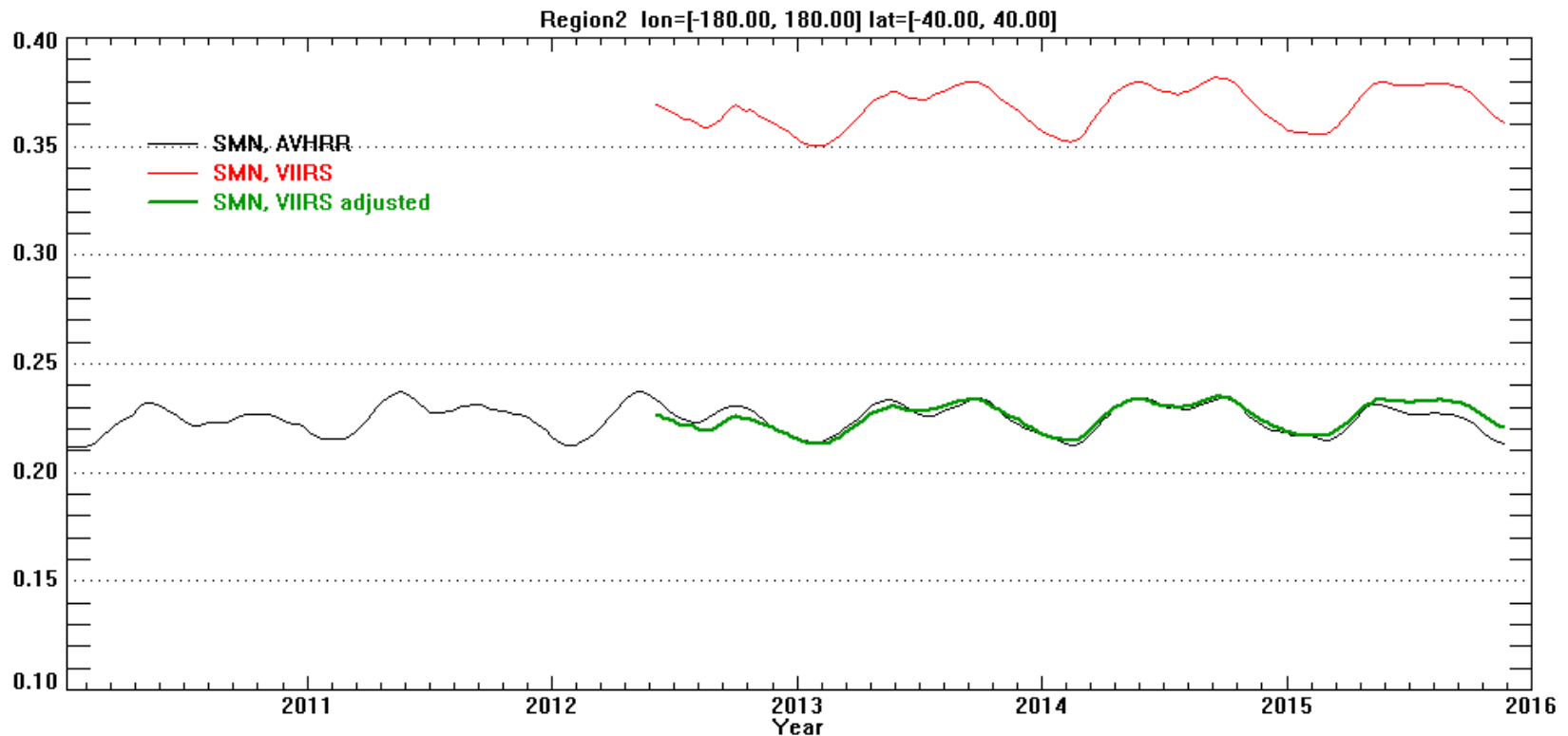
Transition to SNPP/VIIRS & J-1

- TWO methods:

1. Adjust SNPP/VIIRS weekly NDVI & BT to weekly NOAA/AVHRR
2. Adjust SNPP/VIIRS **5-year** weekly NDVI & BT **climatology** to NOAA/AVHRR **36-year** climatology

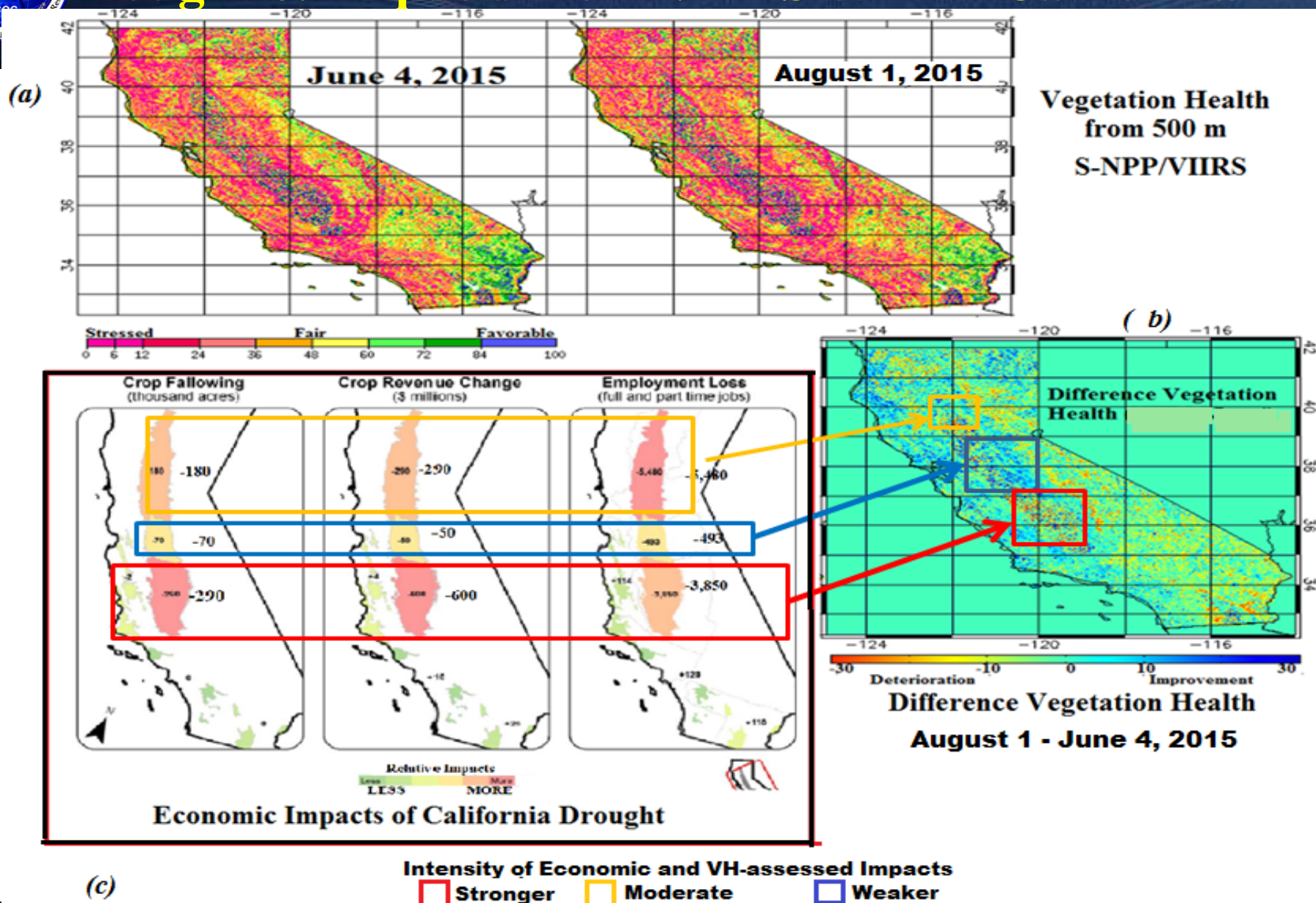


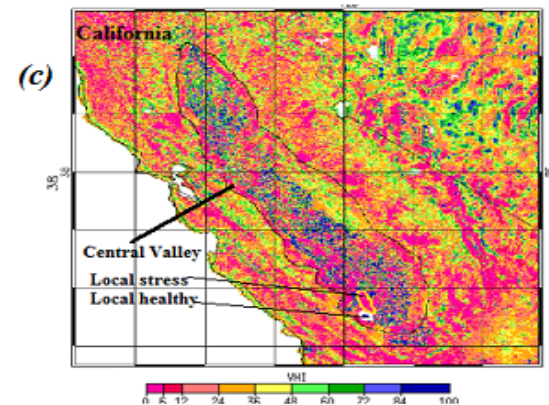
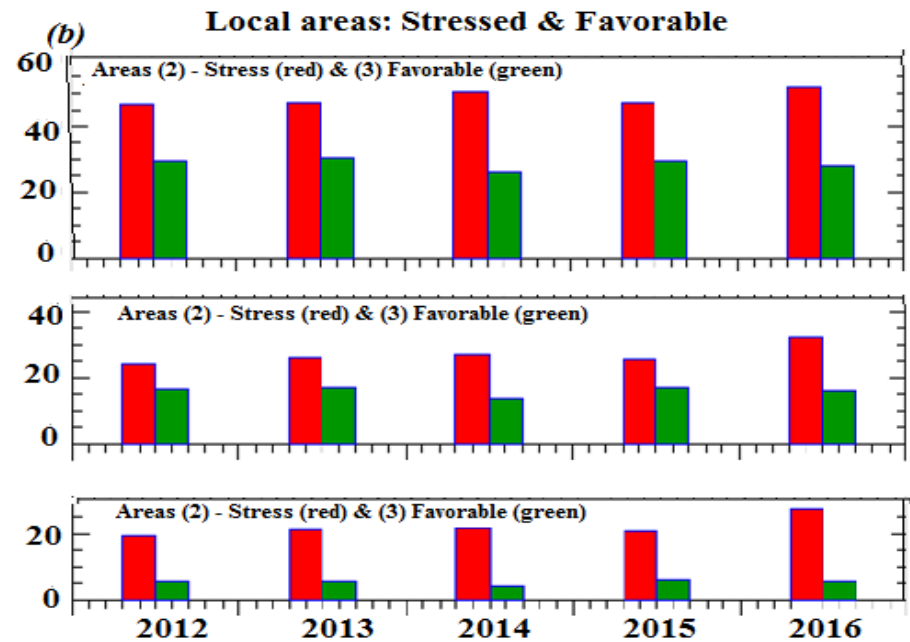
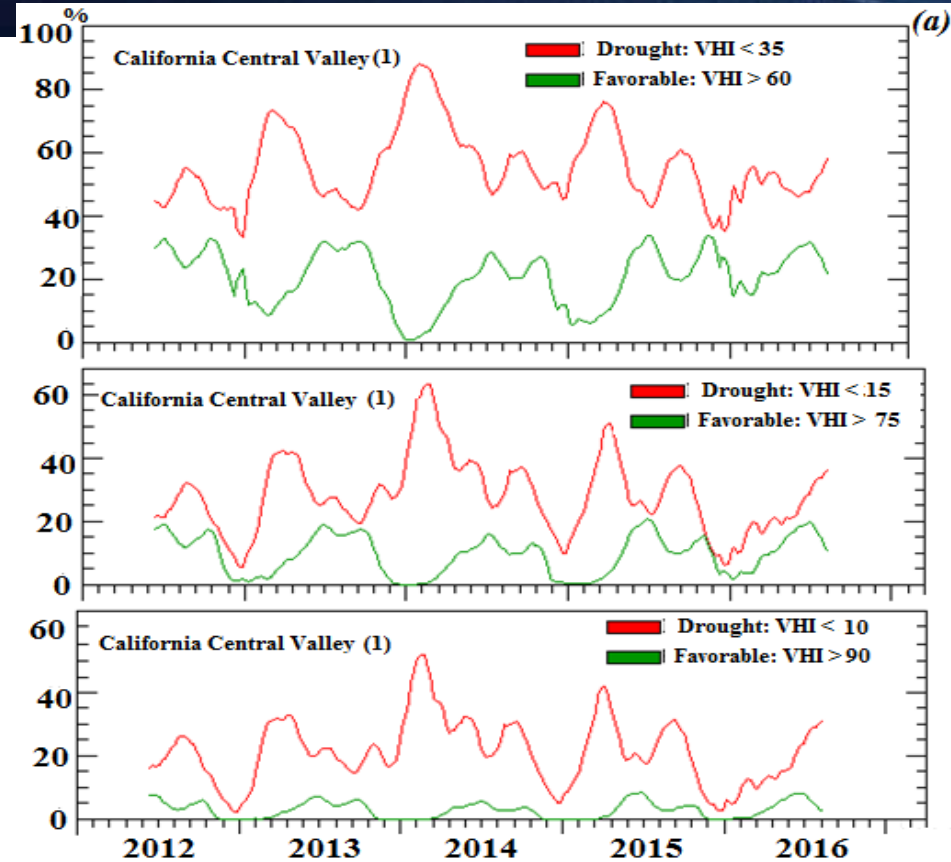
Time series of SMN from VIIRS and AVHRR



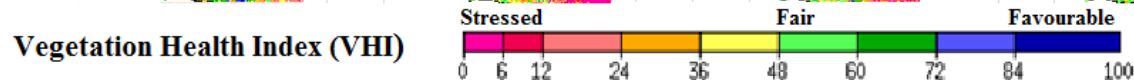
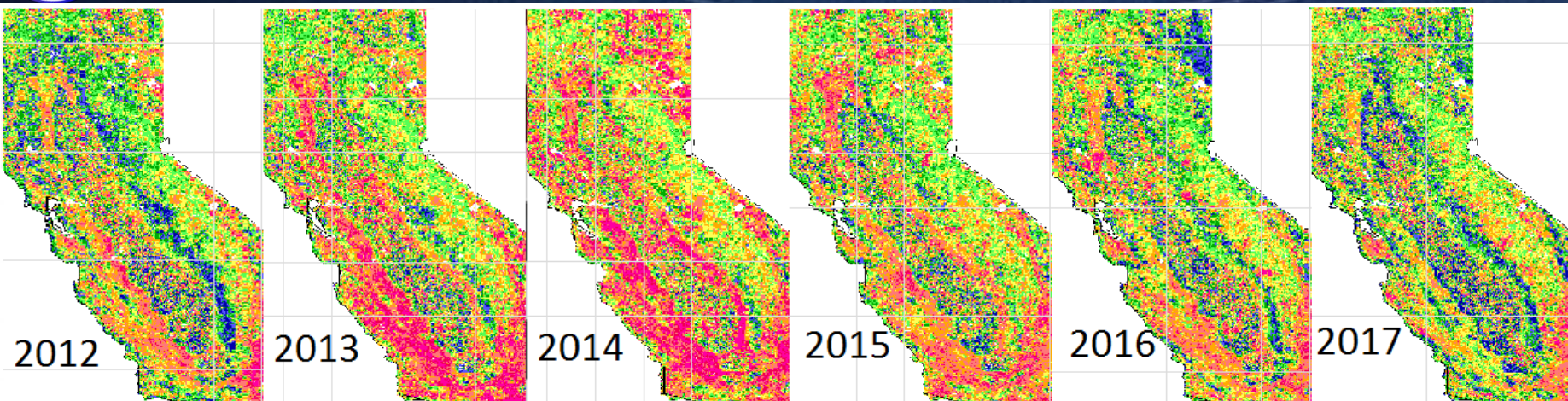
The SMN from VIIRS and AVHRR are not on the same level. Adjustment is required to make the long term time series aligned.

Drought & Impacts from VIIRS 0.5 km California

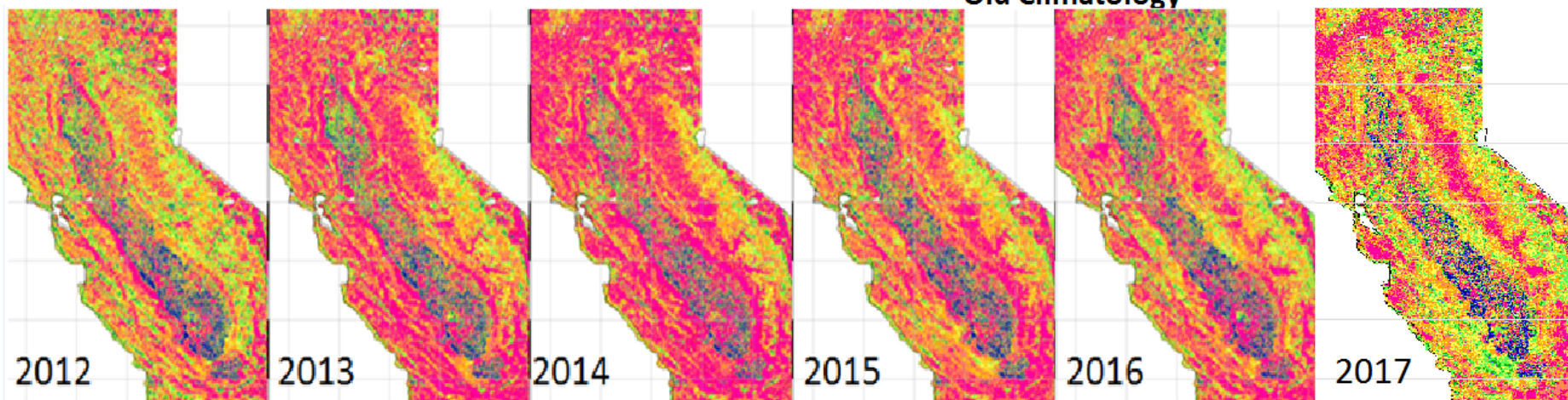




Vegetation Health 0.5 sq.km: New & old Climatology



New Climatology

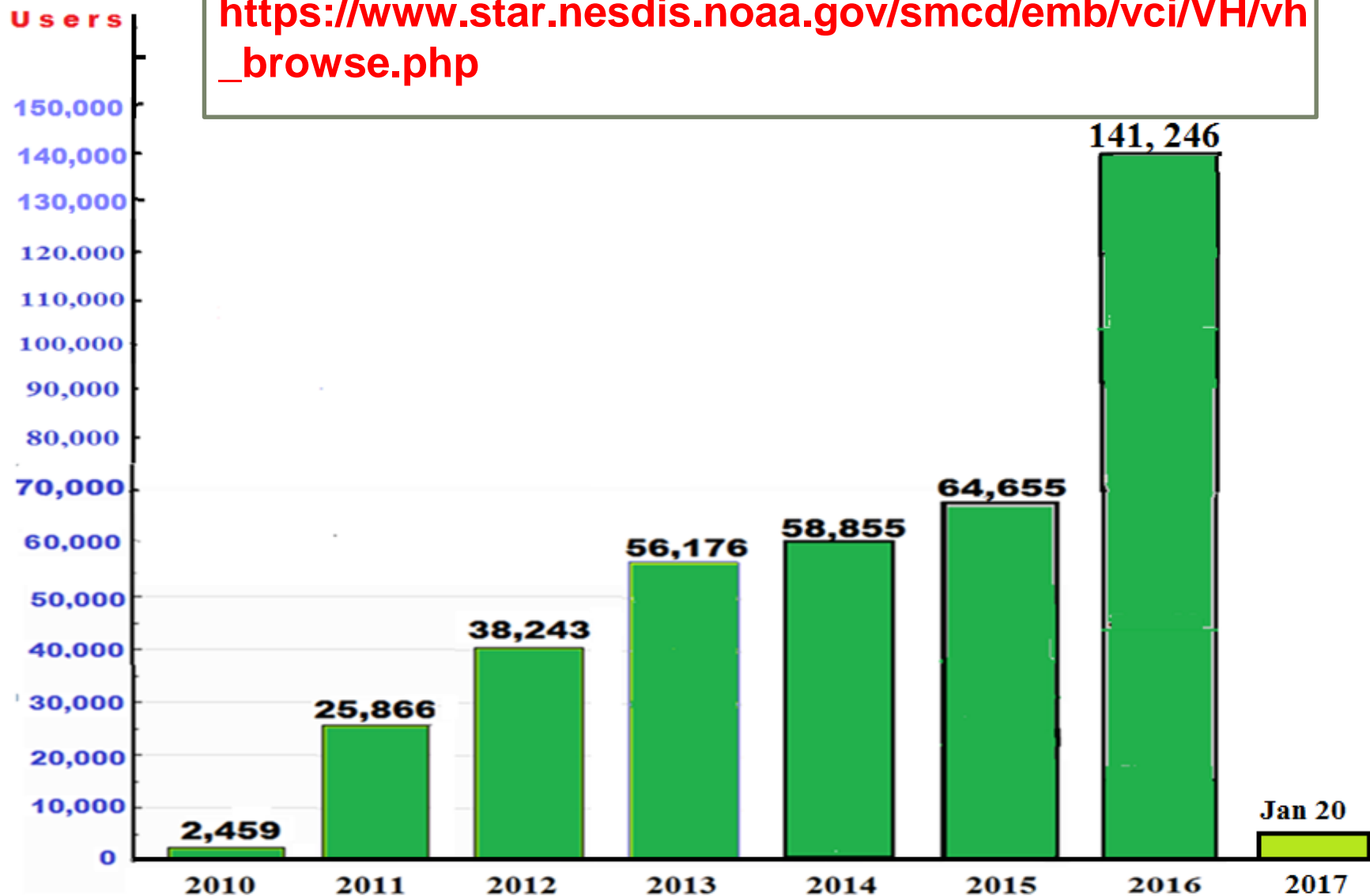


Old Climatology



Users attending Vegetation Health WEB

https://www.star.nesdis.noaa.gov/smcd/emb/vci/VH/vh_browse.php





Publication since the early 1980s

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- Kogan F, W. Guo & W. Ysang 2017. SNPP/VIIRS vegetation health to assess 500 California drought, *Geomatics, Natural Hazards and Risk*, DOI:10.1080/19475705.2017.1337654 (<http://dx.doi.org/10.1080/19475705.2017.1337654>)
- [C. Bhuiyan](#), [A. K. Saha](#), [N. Bandyopadhyay](#) & [F. N. Kogan](#) 2017. Analyzing the impact of thermal stress on vegetation health and agricultural drought – a case study from Gujarat, India. *GIScience & Remote Sensing*, <http://dx.doi.org/10.1080/15481603.2017.1309737>
- Kogan F & W. Guo 2016. Strong 2015–2016 El Niño and implication to global ecosystems from space data. *International Journal of Remote Sensing*, 38:1, 161-178, DOI: [10.1080/01431161.2016.1259679](https://doi.org/10.1080/01431161.2016.1259679)
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- Kogan, F., L. Salazar, and L. Roytman, 2012: Forecasting crop production using satellite based vegetation health indices in Kansas, United States. *International Journal of Remote Sensing*, Vol. 00, No. 0, Xxxx 2011, 1–17, http
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Publication since the early 1980s

- Kogan F.N. 1998. A typical pattern of vegetation condition in southern Africa during El Nino years detected from AVHRR data using three-channel numerical index. *International Journal Remote Sensing*, Vol. 19, No. 18, 3689-3695.
- Kogan FN. 1997. Global drought watch from space. *Bulletin of the American Meteorological Society* **78**: 621-636.
- *Kogan, F.N.J.T. Sullivan and Pu Bu Ciren, 1996. Testing post-launch calibration for the AVHRR sensor on world desert targets during 1985-1993. *Advancement in Space Research*. Vol.17, No. 1, 47-50.
- Kogan F.N. 1995a. Droughts of the late 1980s in the United States as derived from NOAA polar-orbiting satellite data. *Bulletin of the American Meteorological Society*, **76** (5), 655-667.
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- Kogan F. 1993a. Development of global drought-watch system using NOAA/.AVHRR data. *Advancement in Space Research*. Vol.13, No. 5, 219-222.
- Kogan F. 1993b. Global drought detection and impact assessment from space. *DROUGHT: A Global Assessment* (Ed. Wilhite, D.A.), Vol. 1. Routledge, London and New York, 196-209.
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- Kogan F. 1989. Remote sensing of weather impacts on vegetation in non-homogeneous areas. *International Journal Remote Sensing*, Vol. 11, No. 8, 1405-1419.
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- Kogan F. 1985a. Sun Won't Keep Shine on Soviet Agriculture. *The Wall Street Journal*. August 29, p12.
- Kogan F. 1985b. Look for Soviet buying of grain to drop this year. *The Des Moines Register*. August 13.
- Kogan F. 1983a. Perspectives for grain production in the USSR. *Agricultural Meteorology*, 28, 213-227.
- Kogan F. 1983b. Soviet grain production: Resources and prospects. *Soviet Geography*, Vol. XXIV, No 9, 631-661.



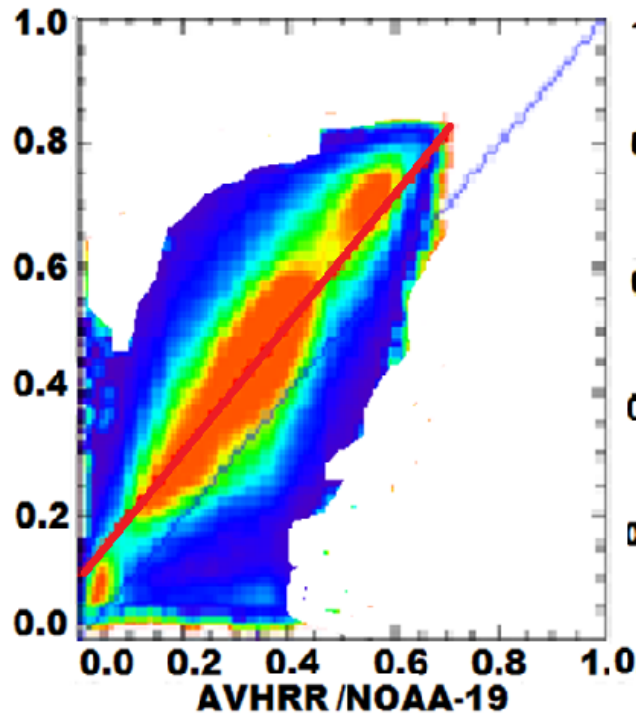
• **BACK UP**



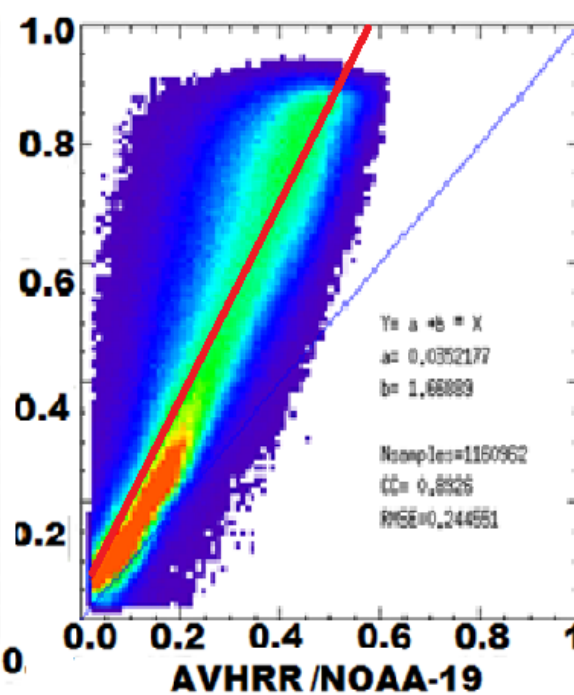
Mean NDVI (40N-40S)

VIIRS vs MODIS vs AVHRR: 2012 July

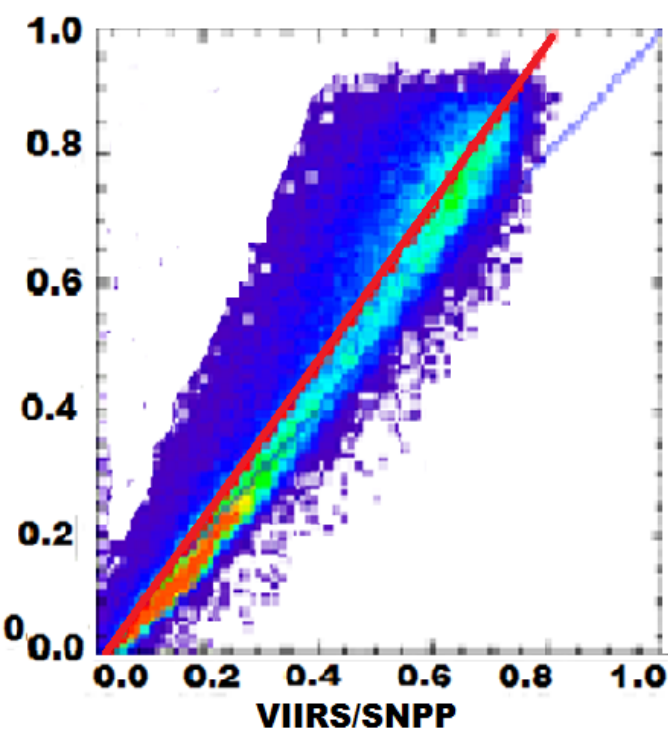
VIIRS/SNPP



MODIS/AQUA



MODIS/AQUA



ACCOMPLISHMENTS

- Data collection from SNPP/VIIRS, AVHRR/NOAA-18 & 19, MODIS/AQUA
- Algorithm Development (data sampling, gap filling, reflectance calculation, noise removal)
- Data processing (NDVI, BT, no noise NDVI & BT)
- Data Stability tests

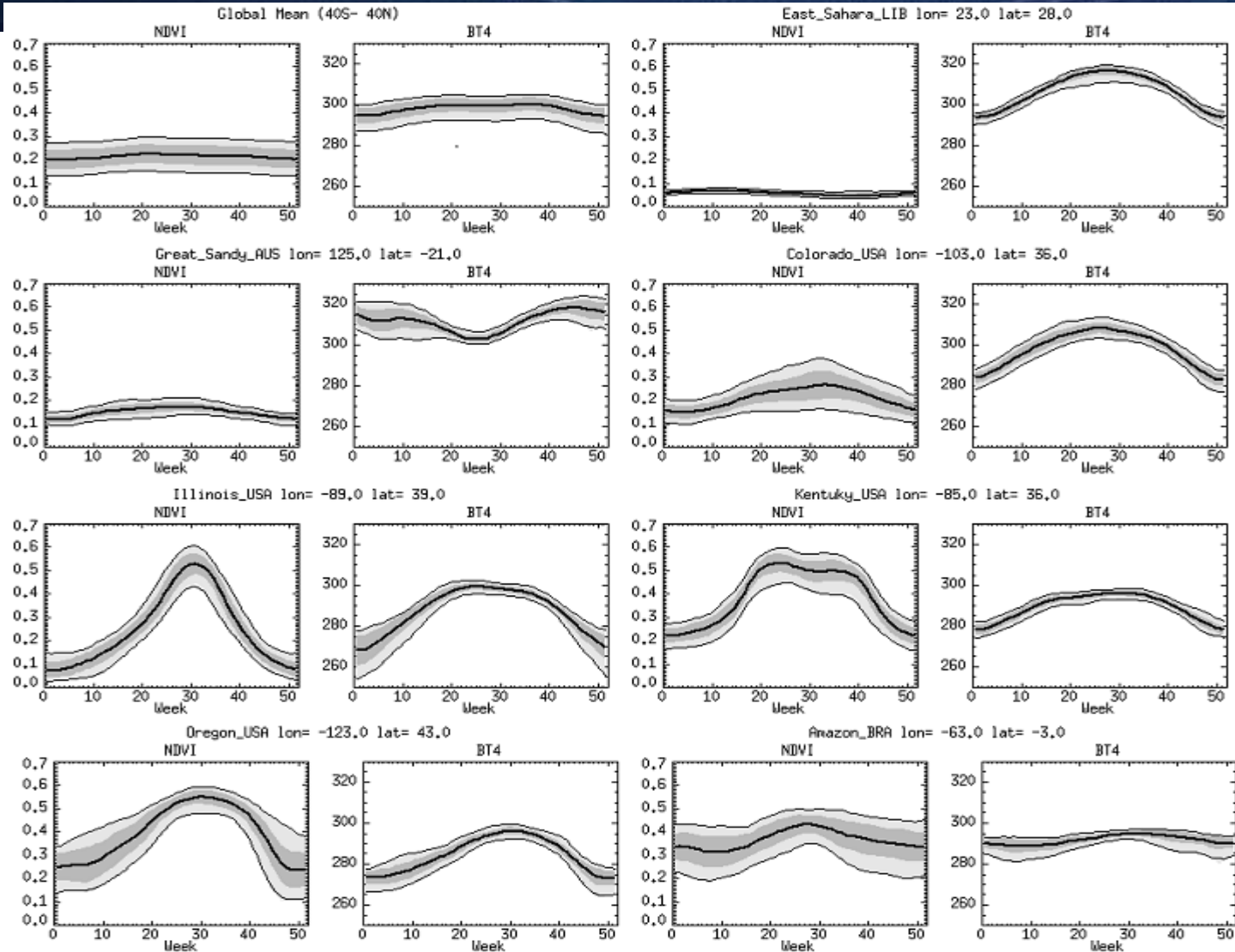


Algorithm Background

- **The SNPP & J1/VIIRS Vegetation Health (VH) Indices from EDR**
- **VIIRS/VH-1 is global weekly composite products with 1 km² resolution**
- **The VH-EDR consists of three indices, Vegetation Condition Index (VCI), Thermal Condition Index (TCI) and Vegetation Health Index (VHI)**
- **The algorithm is based on three bio-physical laws: Liebig's Law-of-Minimum, Shelford's Law-of-Tolerance & Carrying Capacity**
- Daily **SNPP/VIIRS-1** is derived from the TOA radiances: VIS, NIR - image bands: & Thermal bands
- The radiances are pre-launch & post-launch calibrated, checked for accuracy
- Weekly composite raw Normalized Difference Vegetation Index (NDVI) and raw Brightness temperature (BT) are calculated after adjustment for sun angle, observation time & atmospheric effects
- Noise removal from weekly raw NDVI & BT and converting them to no noise NDVI (SMN) and BT (SMT)
- Climatology derivation from multi-year SMN and SMT weekly data, using three Bio-physical Laws
- Two SNPP/VIIRS & J1/VIIRS climatology conversions to longer data sets & tests:
 - NDVI & BT real-time weekly from 5-year SNPP/VIIRS to 36-year NOAA/AVHRR
 - NDVI & BT climatology weekly from 5-year SNPP/VIIRS to 36-year NOAA/AVHRR
- Modeling SNPP/VIIRS-1 Vegetation condition index (VCI), Temperature condition index (TCI) & Vegetation health index (VHI)
- Derivation of VH-EDR's per-pixel quality flags (e.g., land/water snow, desert, cloud mask and random noise)
- Test SNPP VIIRS/VH-1 performance with in situ data

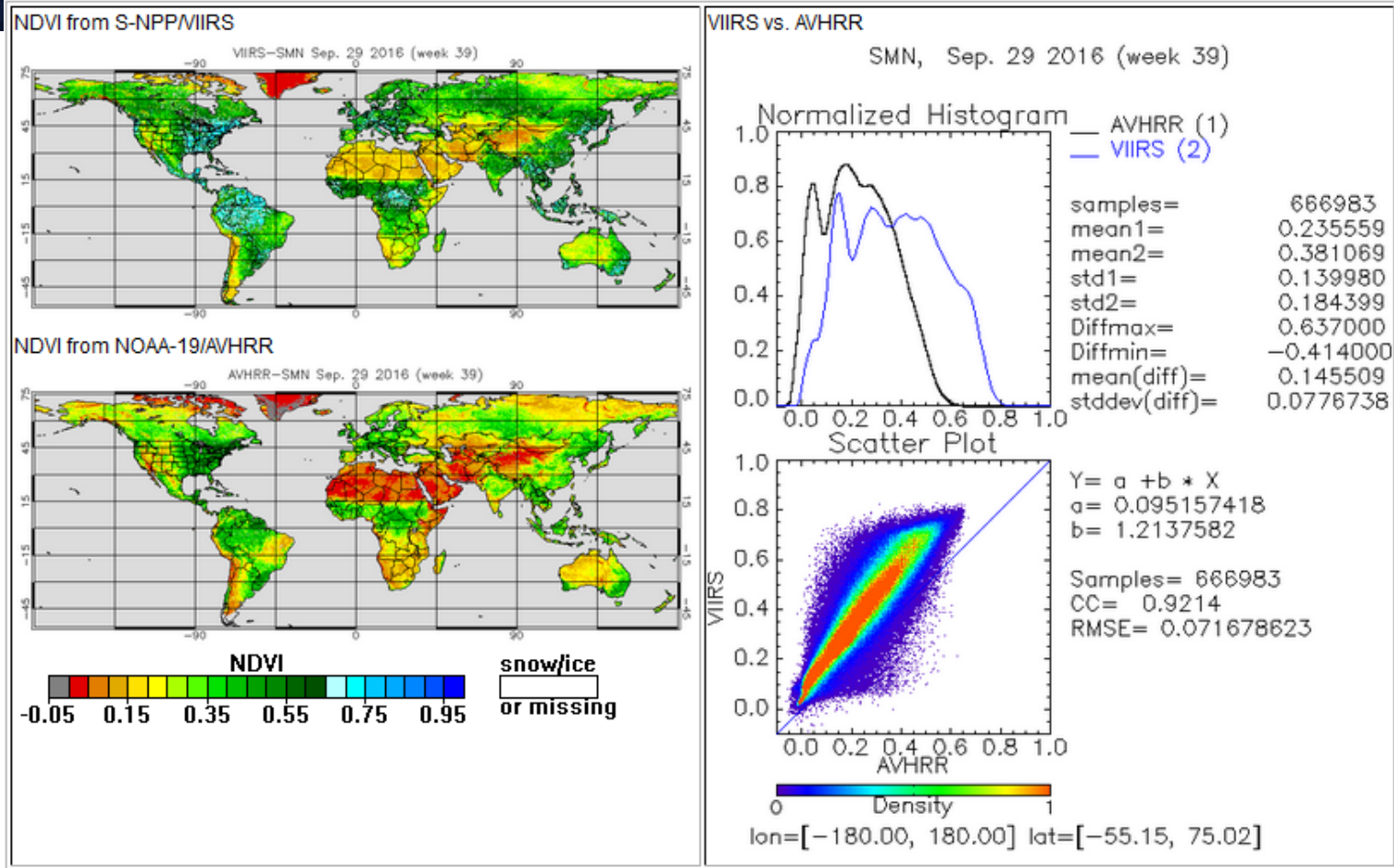


36-year Climatology: Max, Min, Mean





Expl: NDVI/SMN – VIIRS vs AVHRR



The SMN from VIIRS and AVHRR are not on the same level. Adjustment is required to make the long term time series aligned.



Vegetation Health Milestones Update

June 10, 2017

- Calculate 5 year climatology (max and min) from S-NPP VIIRS (Apr, 2017)
- Calculate same 5 year climatology from AVHRR (Apr 2017)
- Similar/dissimilar for the two climatology (May, 2017)
- Similar/dissimilar for 5-year & 36-year climatology
- Calculate VIIRS/VH (VCI, TCI, VHI) (Jun, 2017)
- Compare VIIRS/VH to AVHRR/VH for 2012-2017 (Jan, 2018)
 - If there are issues with the VIIRS climatology re-calculate it each year with J1 data.



Algorithm Improvements

Algorithm Improvements

- 4 km Climatology from AVHRR
- 4 and 1 km Climatology from VIIRS
- Cal/Val Climatology for evaluating algorithm performance
 - a. Comparison with AVHRR algorithms
 - b. Comparison with *in situ* data



SNPP/VIIRS-1 Climatology

- Directly generate 5-year VIIRS 1km climatology (V5);
- Generate pseudo 36-year VIIRS 1km climatology (V36) based on VIIRS 4km V36 climatology and 1km V5 climatology:

- Get mean value over 4km grid cell for climatology mean:

From V5 1km MEAN at specific week ($MEAN_{viirs1km}$), calculate mean values over 4km grid ($MEAN_{viirs4km}$).

- Calculate adjustment:

$$\text{Adjustment} = MEAN_{viirs1km} - MEAN_{viirs4km}$$

- Interpolate 4km VIIRS climatology to 1km resolution ($MXN_{viirs4km} \rightarrow MXN_{viirs1km}$)

$$MXN_{viirs4km} \rightarrow MXN_{viirs1km}$$

- $MXN_{new} = MXN_{viirs1km} + \text{Adjustment}$

V5
1km

V36
4km

V5 4km
Mean

V36 1km
Grid

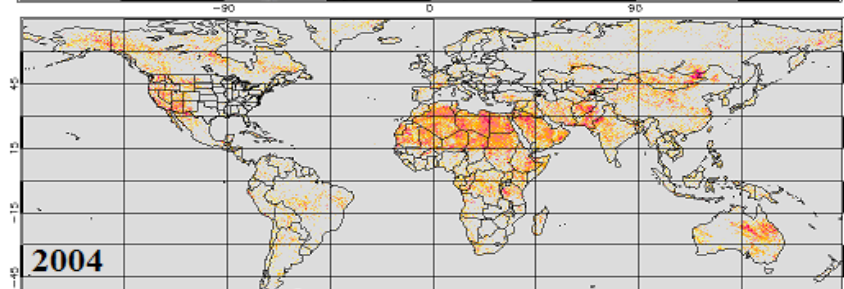
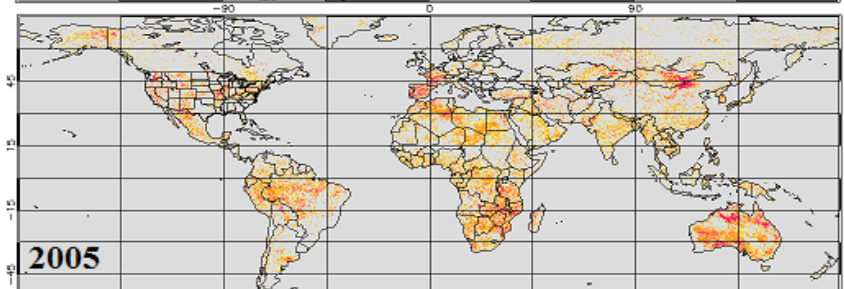
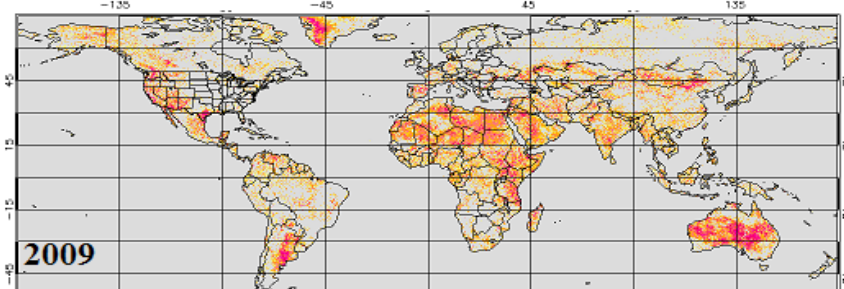
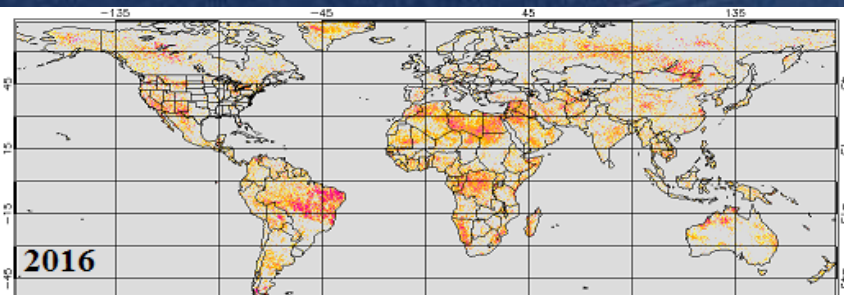
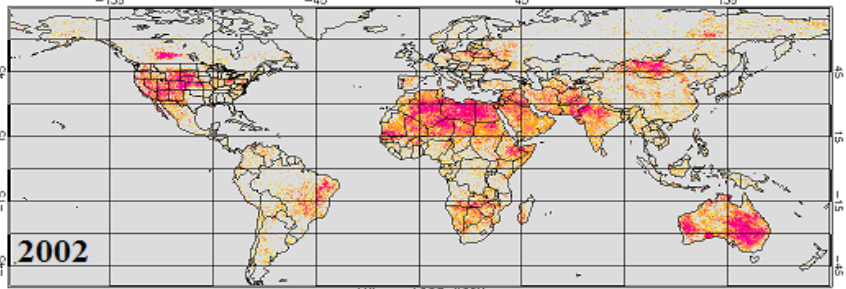
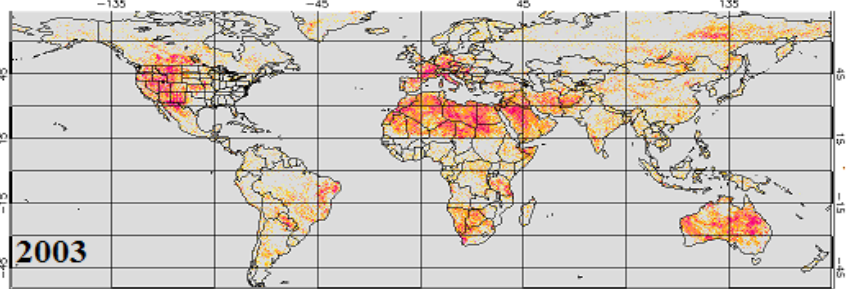
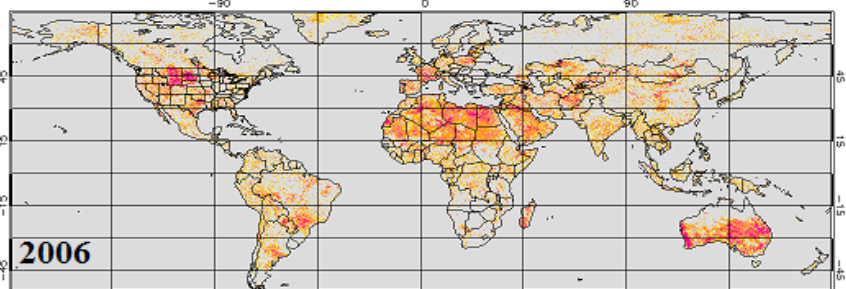
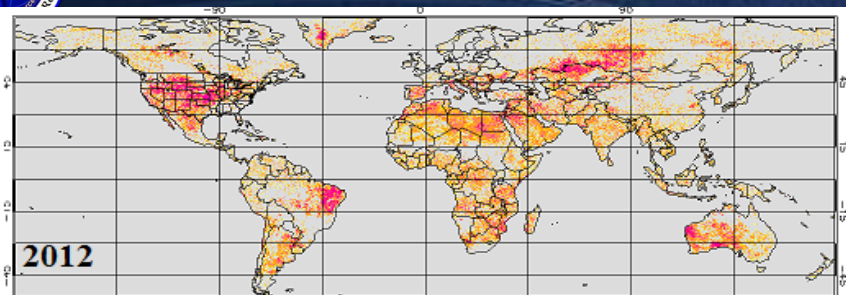
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+

V36
1km



Back Up



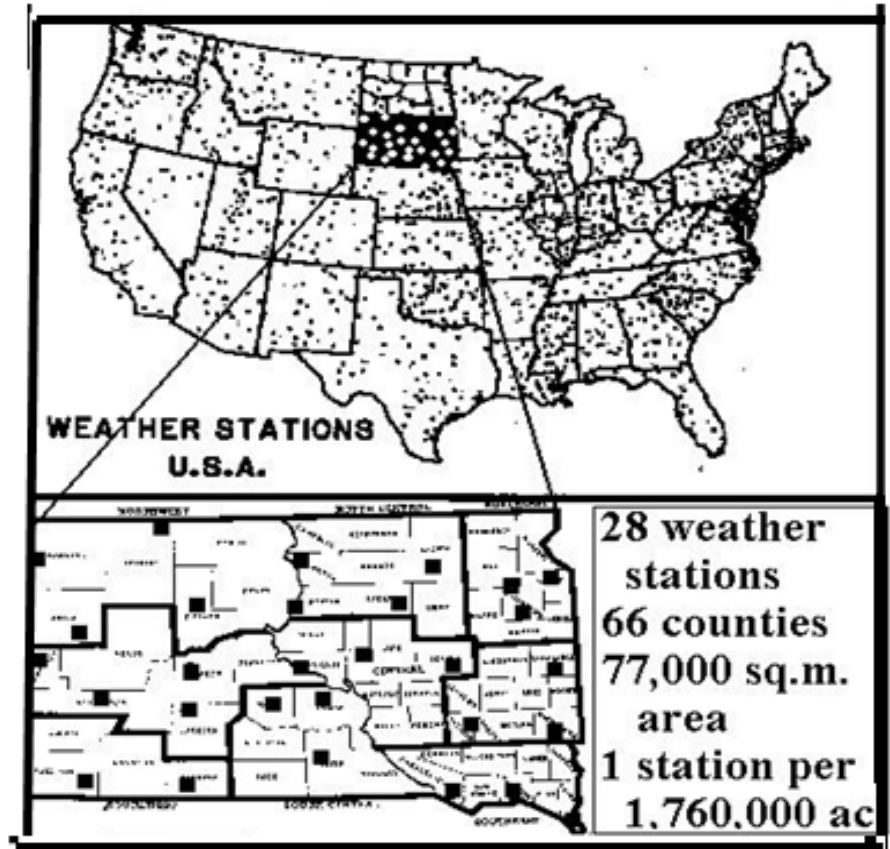
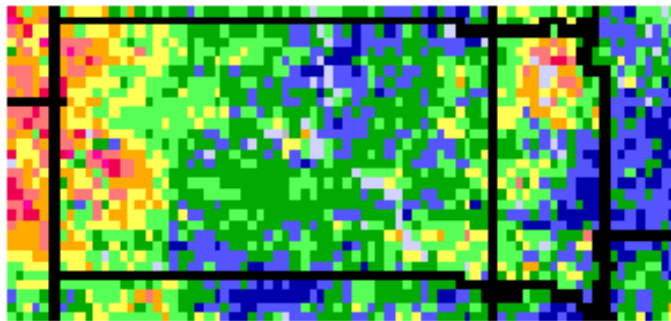
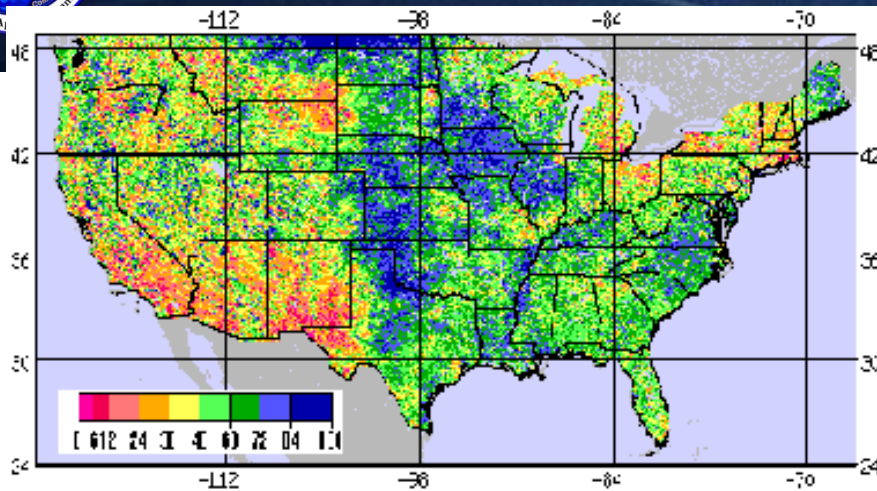
(a)

Exceptional Moderate Mild
 0 6 12 24 36 40

(b)



Weather Stations & Satellite Data Coverage



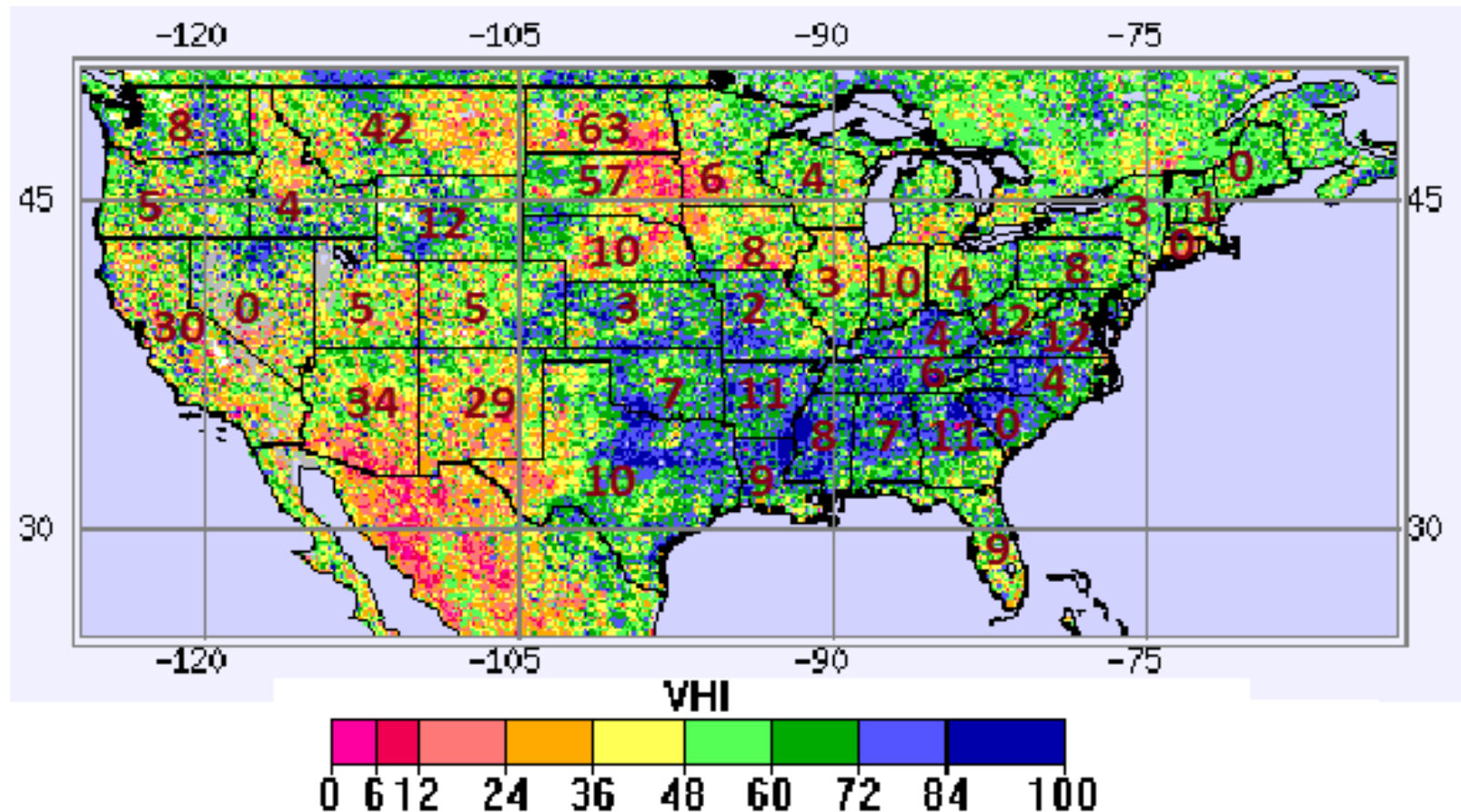
LAND COVERAGE

Satellite:

Every 4, 1 and 0.25 sq.km

Weather: One Station

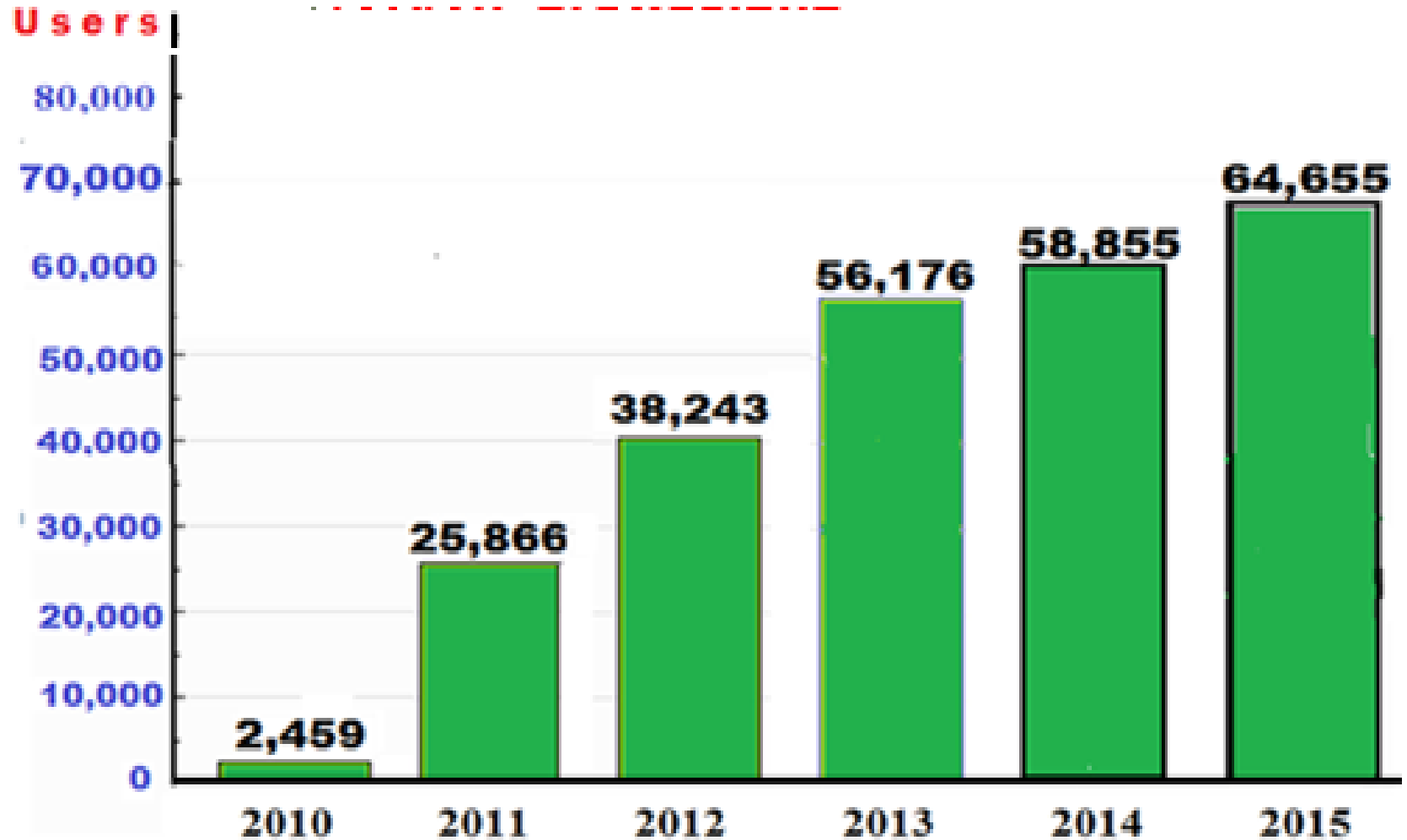
Every 7,140 sq.km
(1,760,000 acres)



Vegetation Health Jul 8, 2017 and % Pasture/Range Land in POOR & Very POOR Conditions, Jul 2, 2017 (USDA)



Number of VH Users per Year





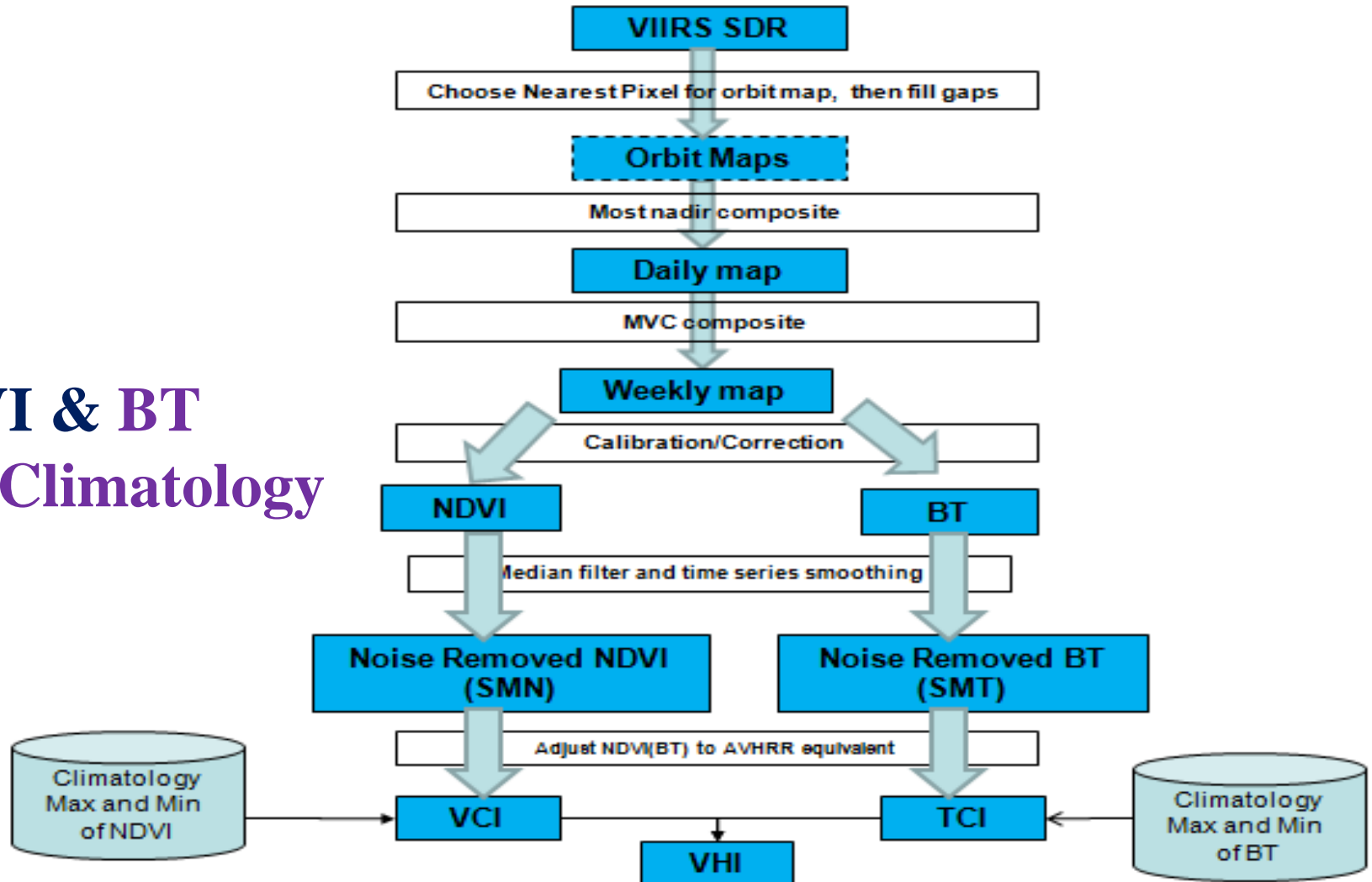
Vegetation Health data sources

Sensors	<i>Advanced Very High Resolution Radiometer (AVHRR)</i> <i>Visible Infrared Imaging Radiometer Suite (VIIRS)</i>	
Satellites	NOAA/AVHRR: NOAA-7, 9, 11, 14, 16, 18, 19 S-NPP/VIIRS, J-1 (JPSS)/VIIRS	
Data Resolution	<i>Spatial</i>	0.5, 1 , 4 (GAC), 8 & 16 (GVI) km;
	<i>Temporal</i>	7-day composite
Period	37-year	(1981-2017), future?
	5-year	(2012-2016) currently
		(2017 – 2030) future
Coverage	World (75 N to 55 S)	
	Channels: VIS , NIR , IR ; Indices: NDVI & BT	



AVHRR Algorithm Architecture (SA)

1. NDVI & BT
2. N-B Climatology

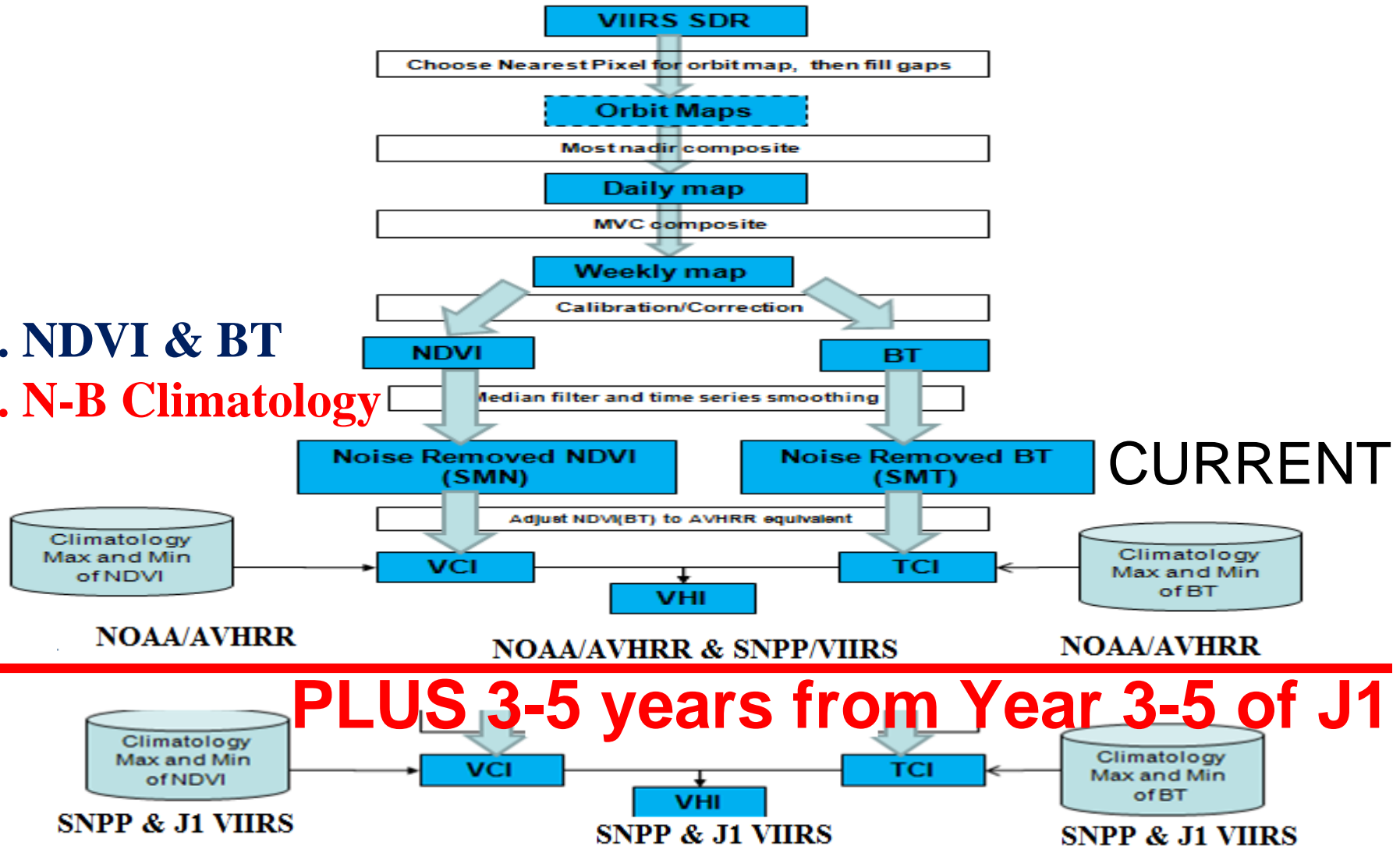




Transition to SNPP/VIIRS & J1/VIIRS

1. NDVI & BT

2. N-B Climatology



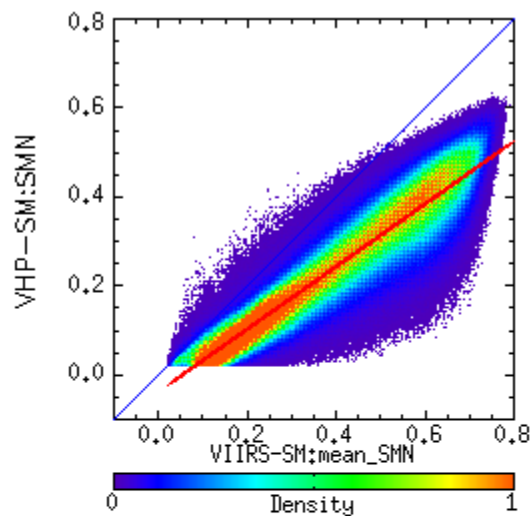


Transition to SNPP/VIIRS & J-1

- From the TWO methods the **first** one is used:
 1. Adjust SNPP/VIIRS weekly NDVI & BT to weekly NOAA/AVHRR



Expl: NDVI (SMN): AVHRR-VIIRS time series



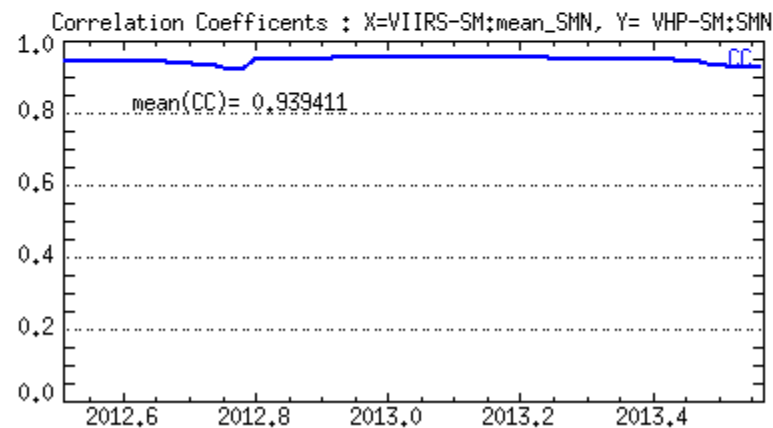
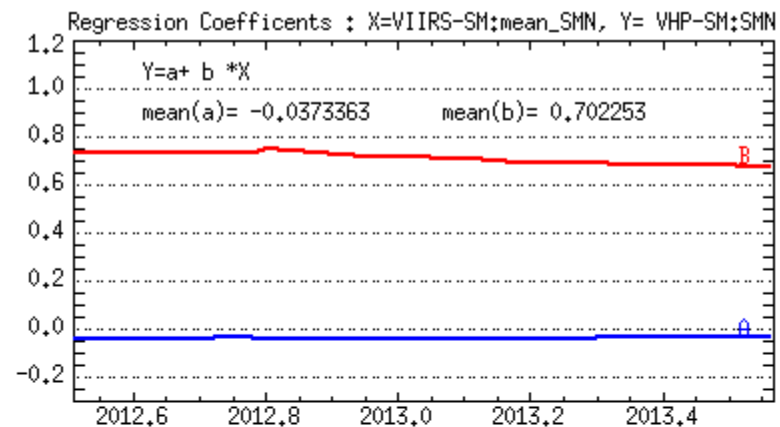
$$Y = a + b * X$$

$$a = -0.0374925$$

$$b = 0.703031$$

Nsamples=4963003
CC = 0.9374
RMSE = 0.0486750

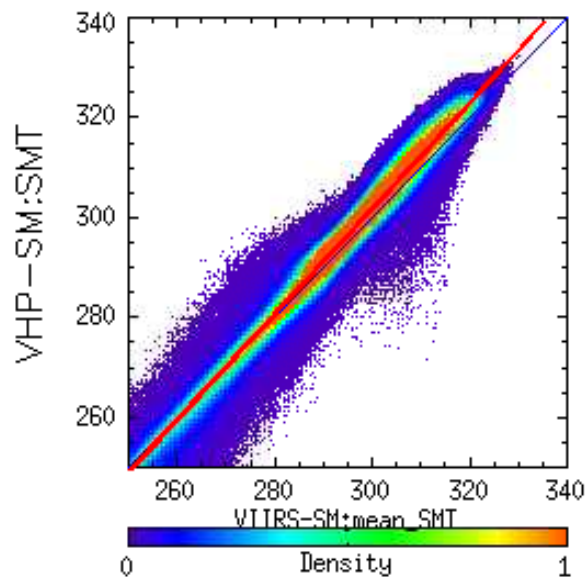
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Data used: 2012 week 27 to 2013 week 30



Expl: BT (SMT) AVHRR-VIIRS COR



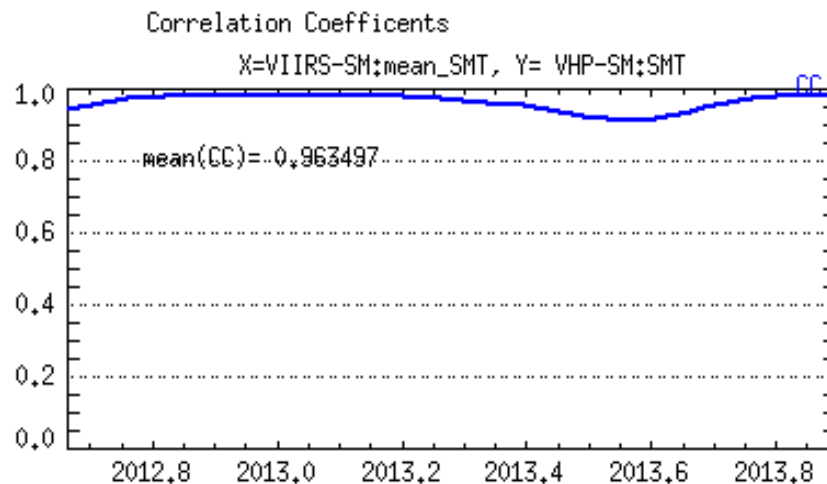
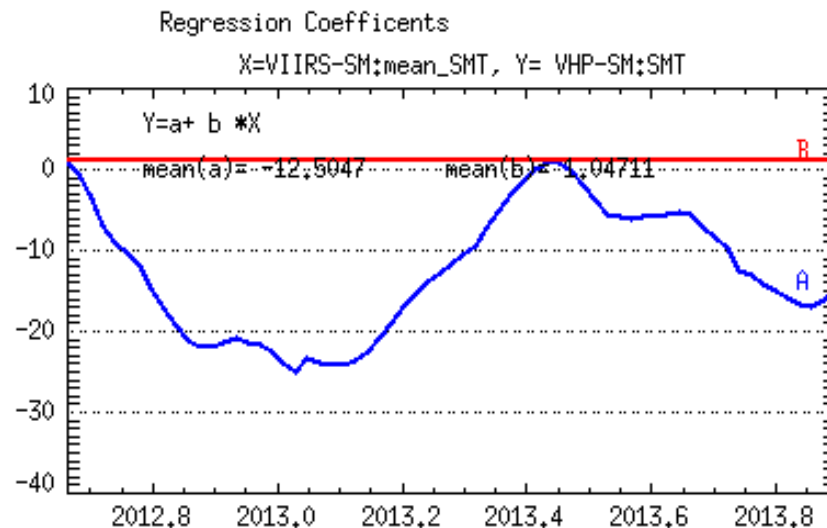
$$Y = a + b * X$$

$$a = -15,3288$$

$$b = 1,05696$$

Nsamples=8638092
CC = 0,9610
RMSE =3,51405

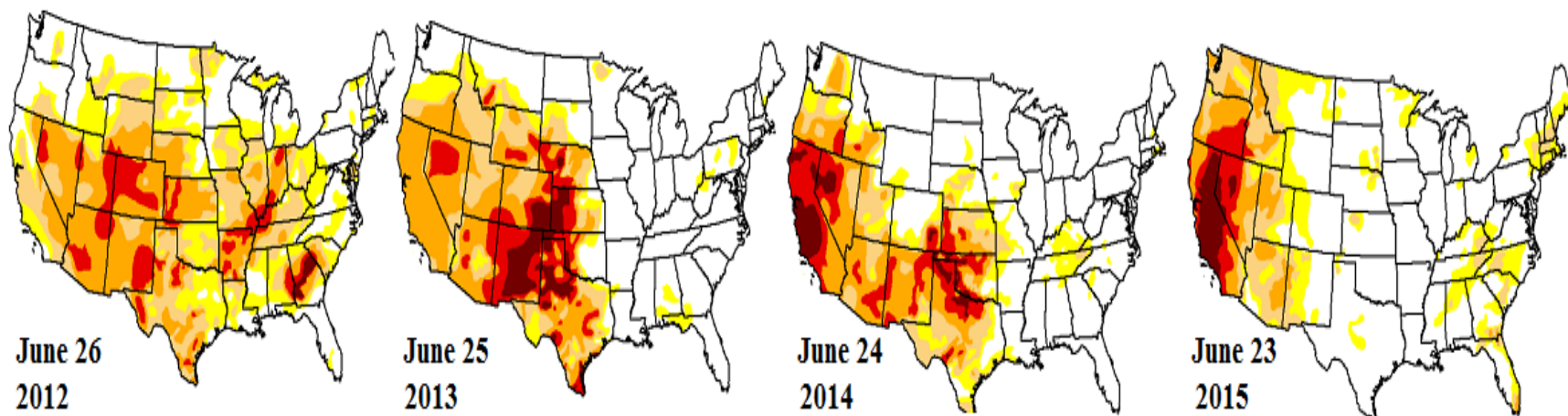
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Data used: 2012 week 35 to 2013 week 47

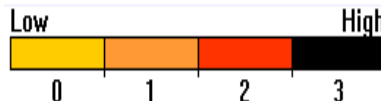
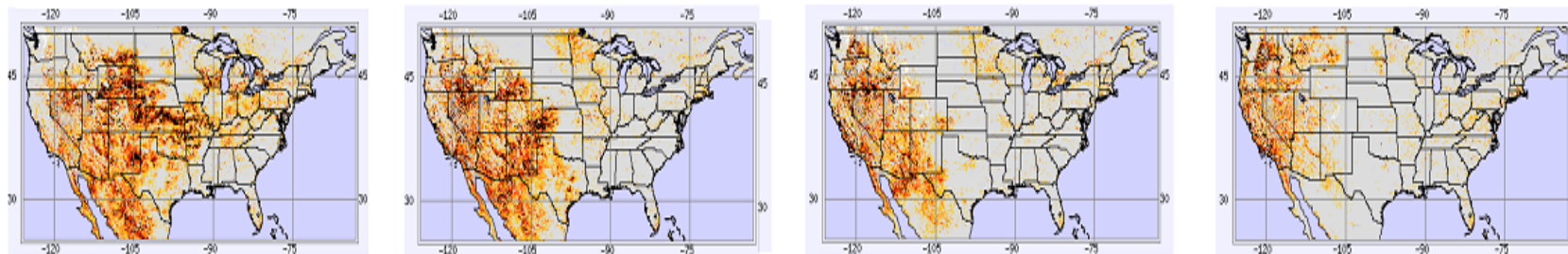


USA Drought from USDM & VHI



D0 Abnormally Dry
 D1 Moderate Drought
 D2 Severe Drought
 D3 Extreme Drought
 D4 Exceptional Drought

Drought from US Drought Monitor (USDM)



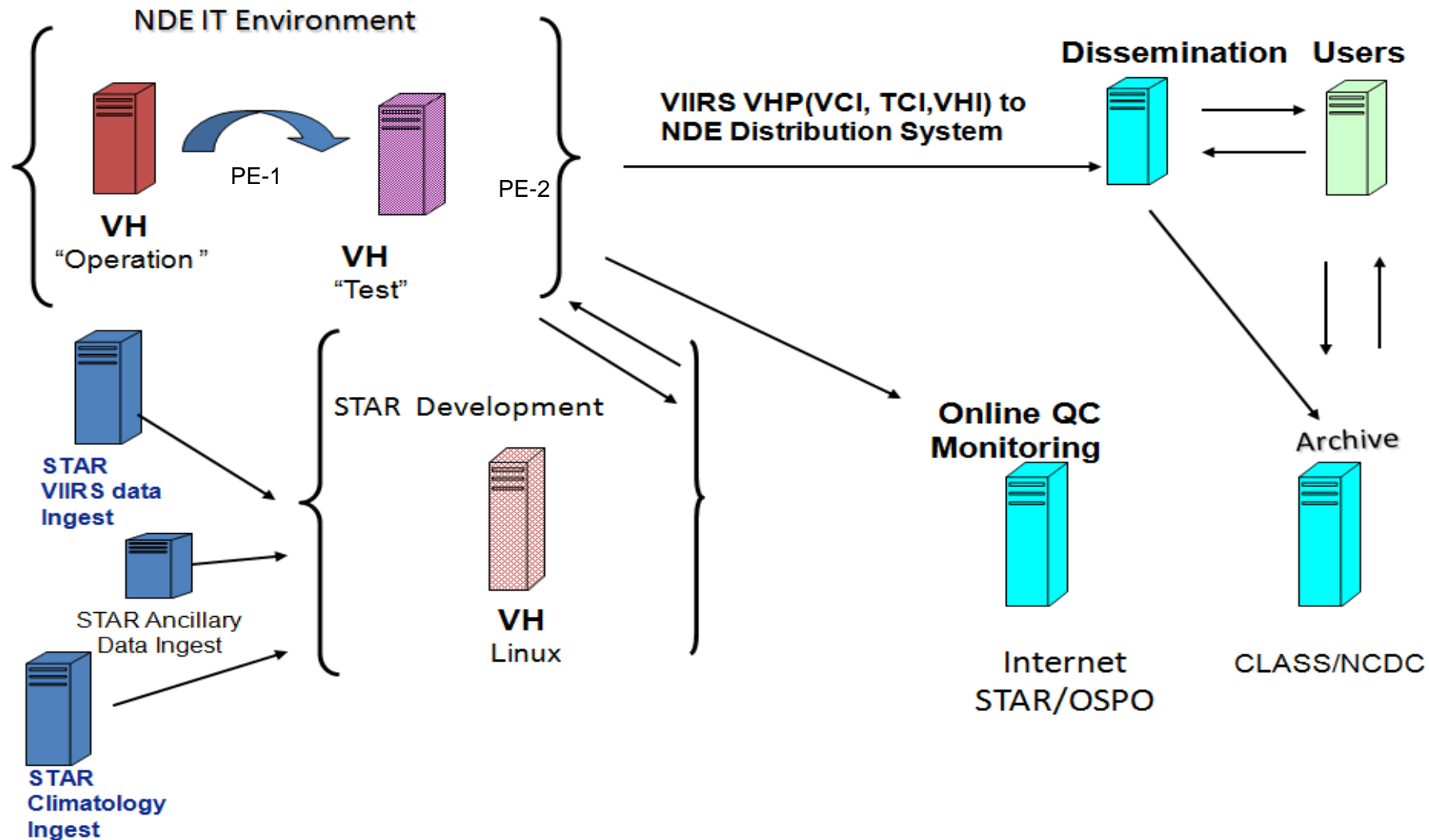
Drought from NOAA-19/AVHRR Vegetation health index (VHI)



VIIRS VHP IT Architectures: Apply #1

Adjust SNPP/VIIRS weekly NDVI & BT to weekly NOAA/AVHRR

VIIRS VHP IT Architectures

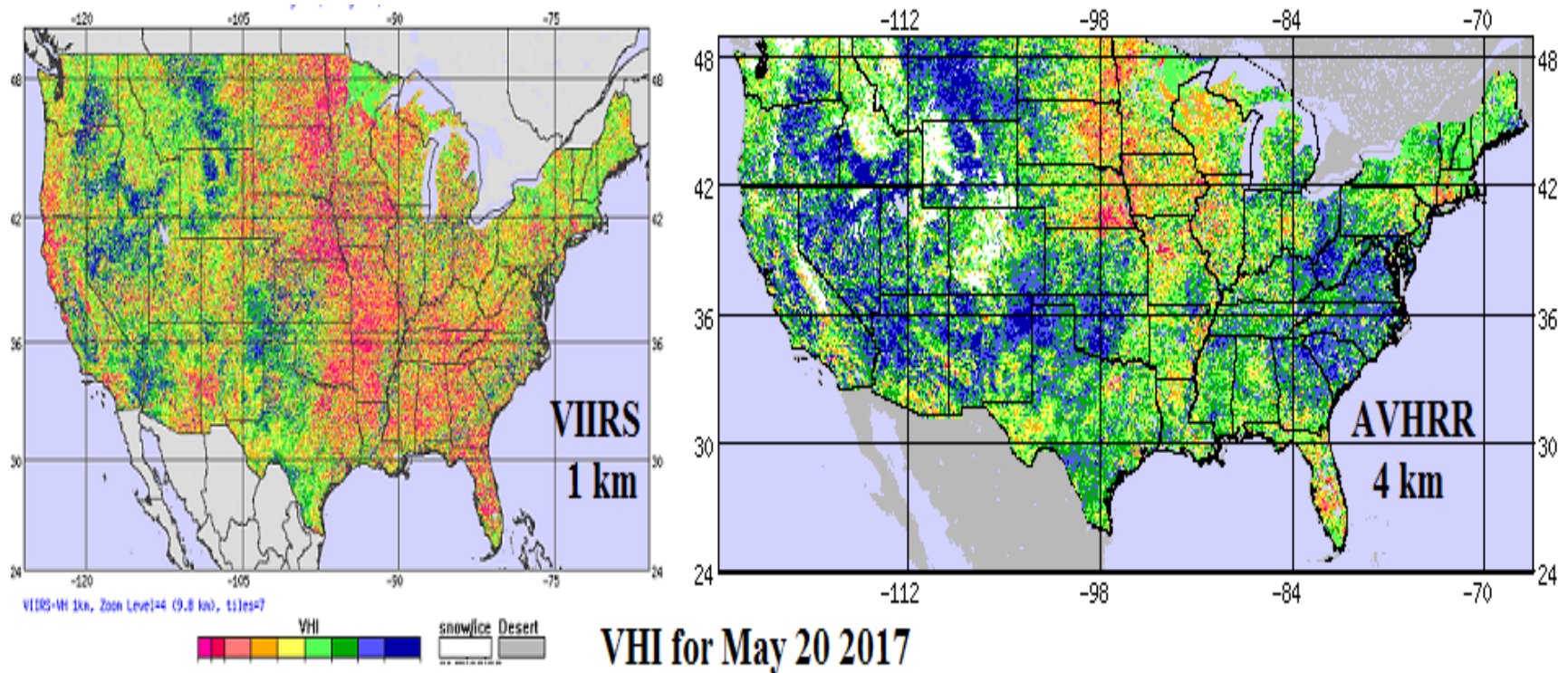




VHI: VIIRS-1km vs AVHRR-4 km

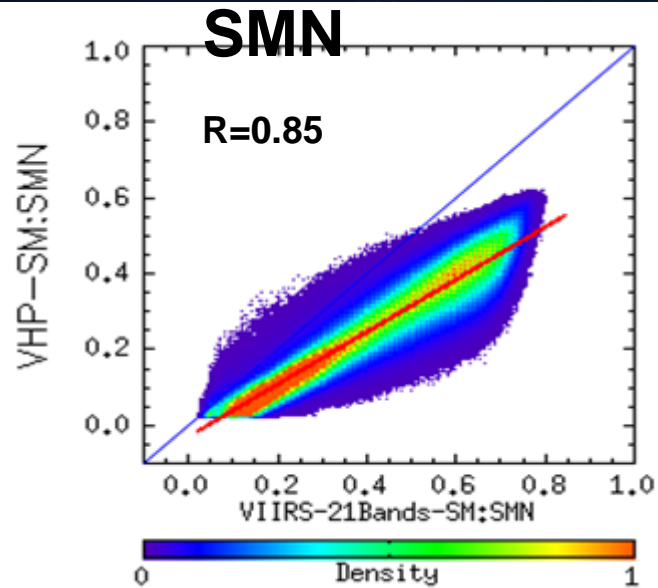
May 20, 2017

•SNPP/VIIRS-1 NOAA/AVHRR-4



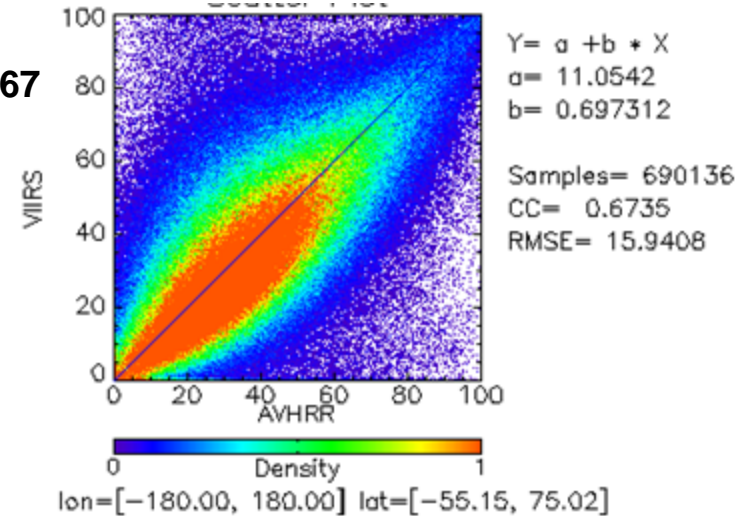


Expl: Error reduction is required



VCI

$R=0.67$



1. Raw data collection
2. VIS, NIR & IR
3. Climatology

problem
problem
problem

TRUE ????

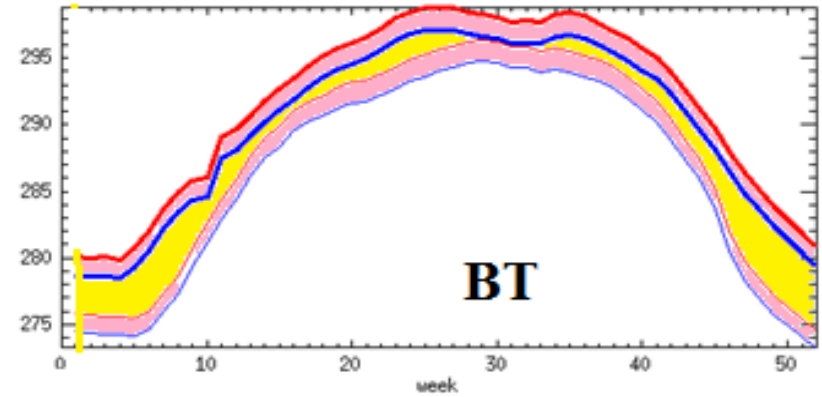
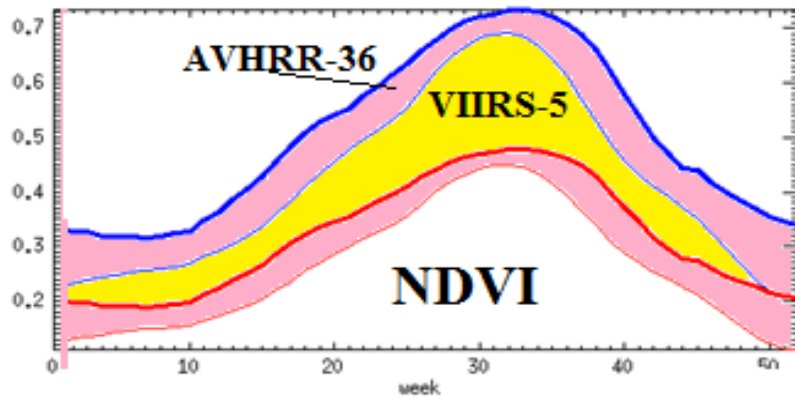


Transition to SNPP/VIIRS & J-1

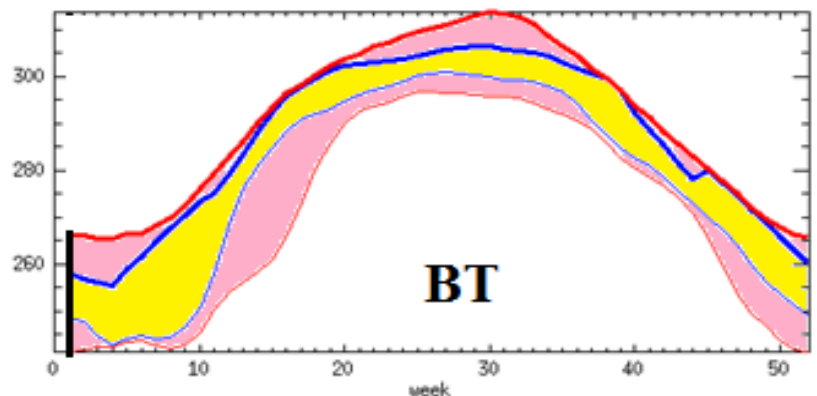
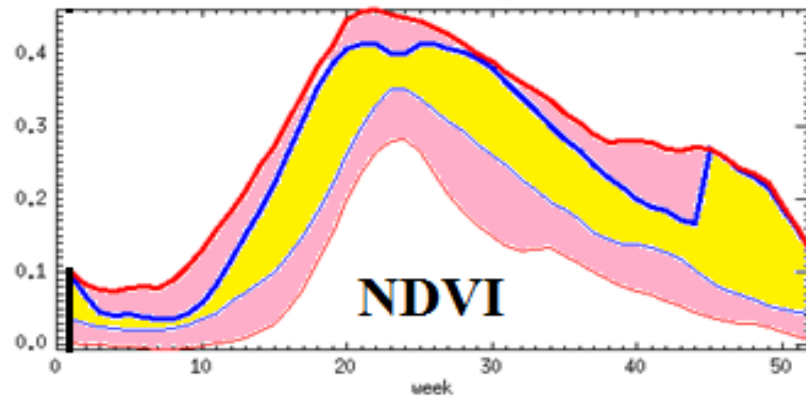
- From the TWO methods the **second** one is used:

Adjust SNPP/VIIRS **5-year** weekly NDVI & BT **climatology** to NOAA/AVHRR **36-year** climatology

Climatology A-36 vs V-5



ILLINOIS, USA (89W, 39N) Grassland



Saratov, RUSSIA (43E, 50N) Steppe

NDVI — Min
— Max
BT — Min
— Max

Climatology AVHRR-36 (1981-2016) & VIIRS-5 (2012-2016)



SNPP/VIIRS & NOAA/AVHRR 4km Climatology (Wz)

- Directly generate 5-year VIIRS 4km climatology (V5);
- Directly generate 5-year AVHRR 4km climatology (A5) at the same period with V5;
- Directly generate 36-year AVHRR 4km climatology (A36);
- Generate pseudo 36-year VIIRS 4km climatology (V36) based on the calculation from V5, A5 and A36 climatology:

$$V36 = V5 \times \frac{A36}{A5}$$

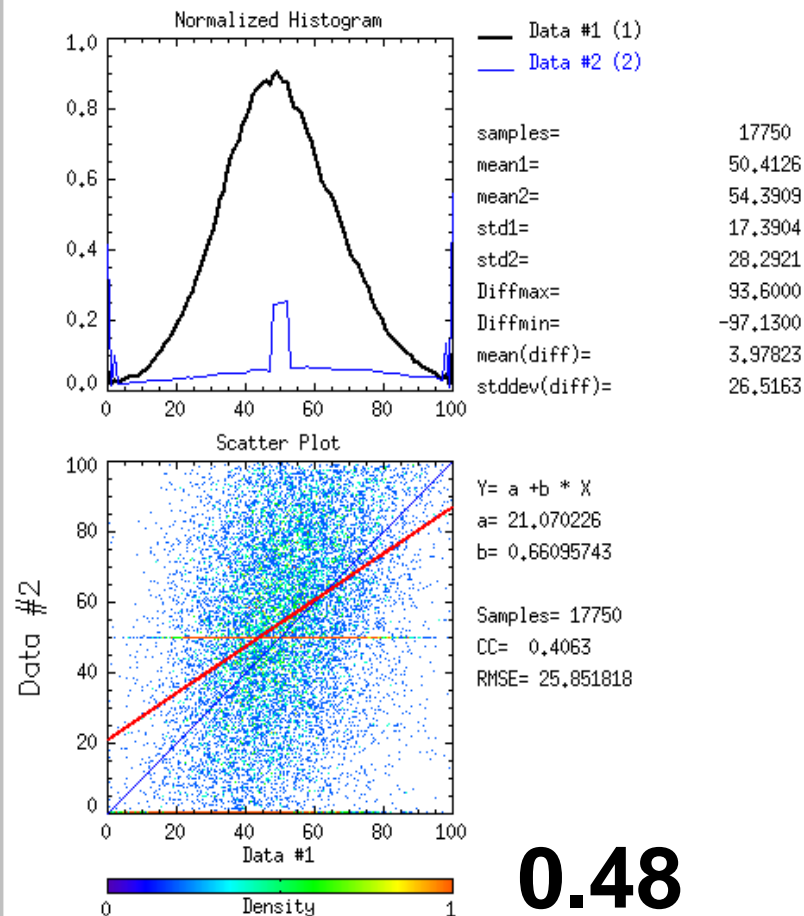


Test: VHI 4 km Product (Wz)

A36 vs. V5

Data #1: /data/home002/wyang/id1_GVI/prog/data_yang/AVHRR_VHP/4km/VH_AVHRR_3I_VHP.G04.C07.NP.P2016030.VH.nc ; VHI
Data #2: /data/home002/wyang/id1_GVI/prog/data_yang/VIIRS/4km/VH_VIIRS_5/VGVI_21Bands.G04.C07.npp.P2016030.VH.nc ; VHI

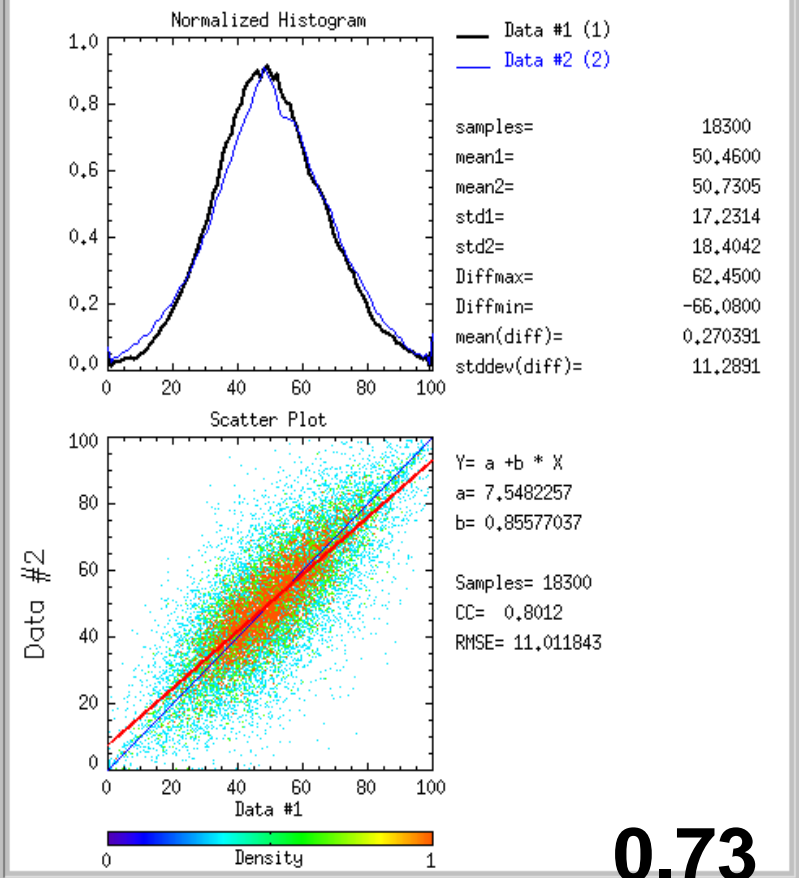
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A36 vs. V36

Data #1: /data/home002/wyang/id1_GVI/prog/data_yang/AVHRR_VHP/4km/VH_AVHRR_3I_VHP.G04.C07.NP.P2016030.VH.nc ; VHI
Data #2: /data/home002/wyang/id1_GVI/prog/data_yang/VIIRS/4km/VH_VIIRS_36_v2/VGVI_21Bands.G04.C07.npp.P2016030.VH.nc ; VHI

lon=[-179.98, 179.23] lat=[-55.10, 74.28]



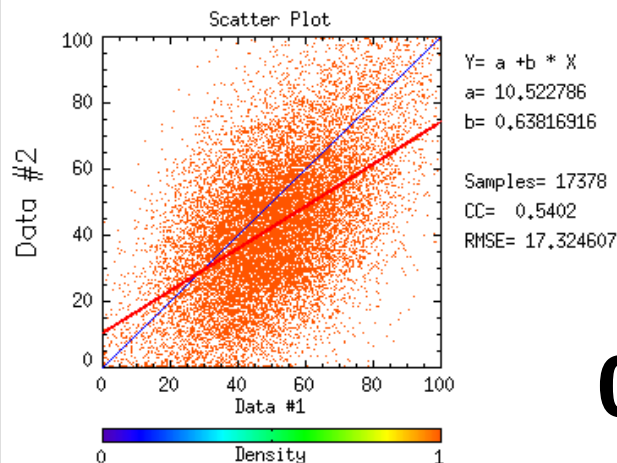
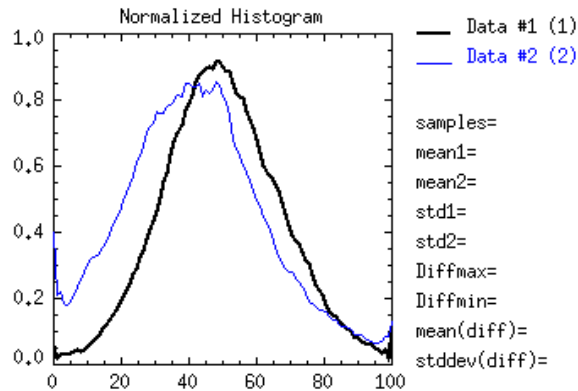


Test: VHI 1km Product

A36 vs. Former Scheme

Data #1: /data/home002/wyang/id1_GVI/prog/data_yang/AVHRR_VHP/4km/VH_AVHRR_30/VHP.G04.C07.NP.P2016030.VH.nc : VHI
Data #2: /data/data044/DATA/VIIRS/1km/VH/VGVI.G1000m.C07.npp.P2016030.VH.nc : VHI

lon=[-179.98, 179.23] lat=[-55.10, 74.28]

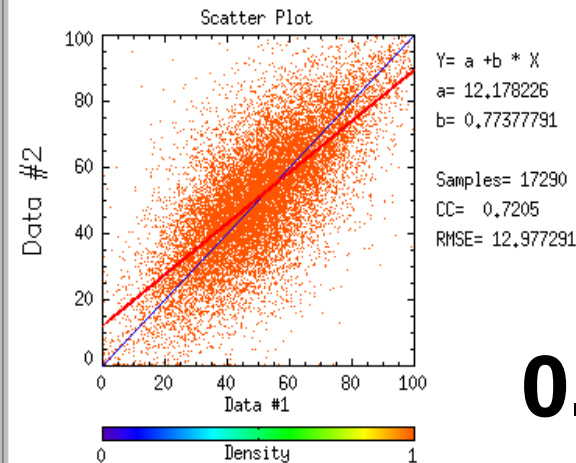
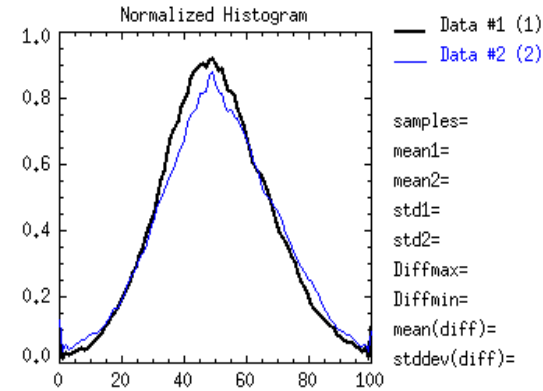


0.54

A36 vs. V36

Data #1: /data/home002/wyang/id1_GVI/prog/data_yang/AVHRR_VHP/4km/VH_AVHRR_30/VHP.G04.C07.NP.P2016030.VH.nc : VHI
Data #2: /data/home002/wyang/id1_GVI/prog/data_yang/VIIRS/1km/VH_VIIRS_36/VGVI.G1000m.C07.npp.P2016030.VH.nc : VHI

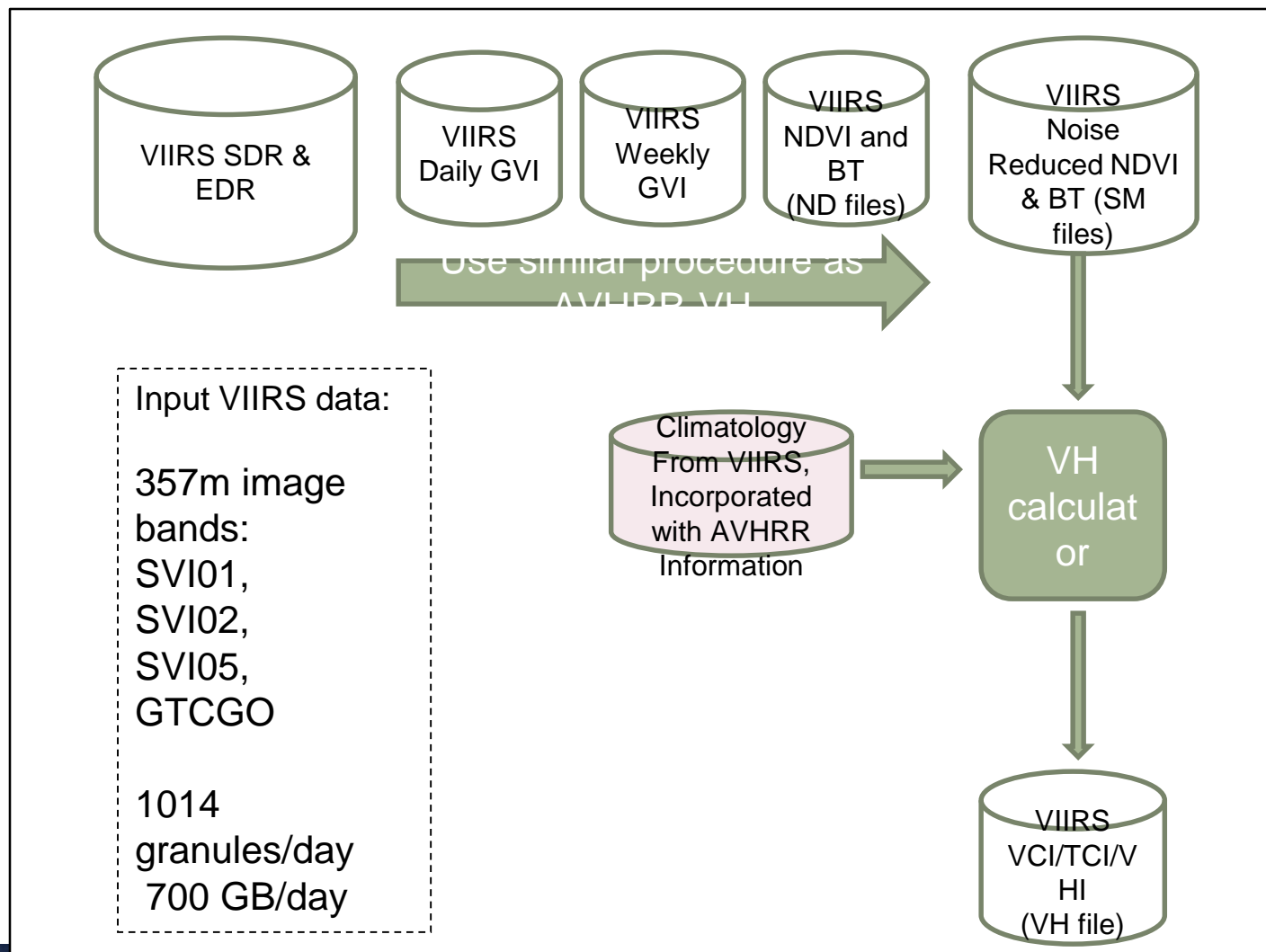
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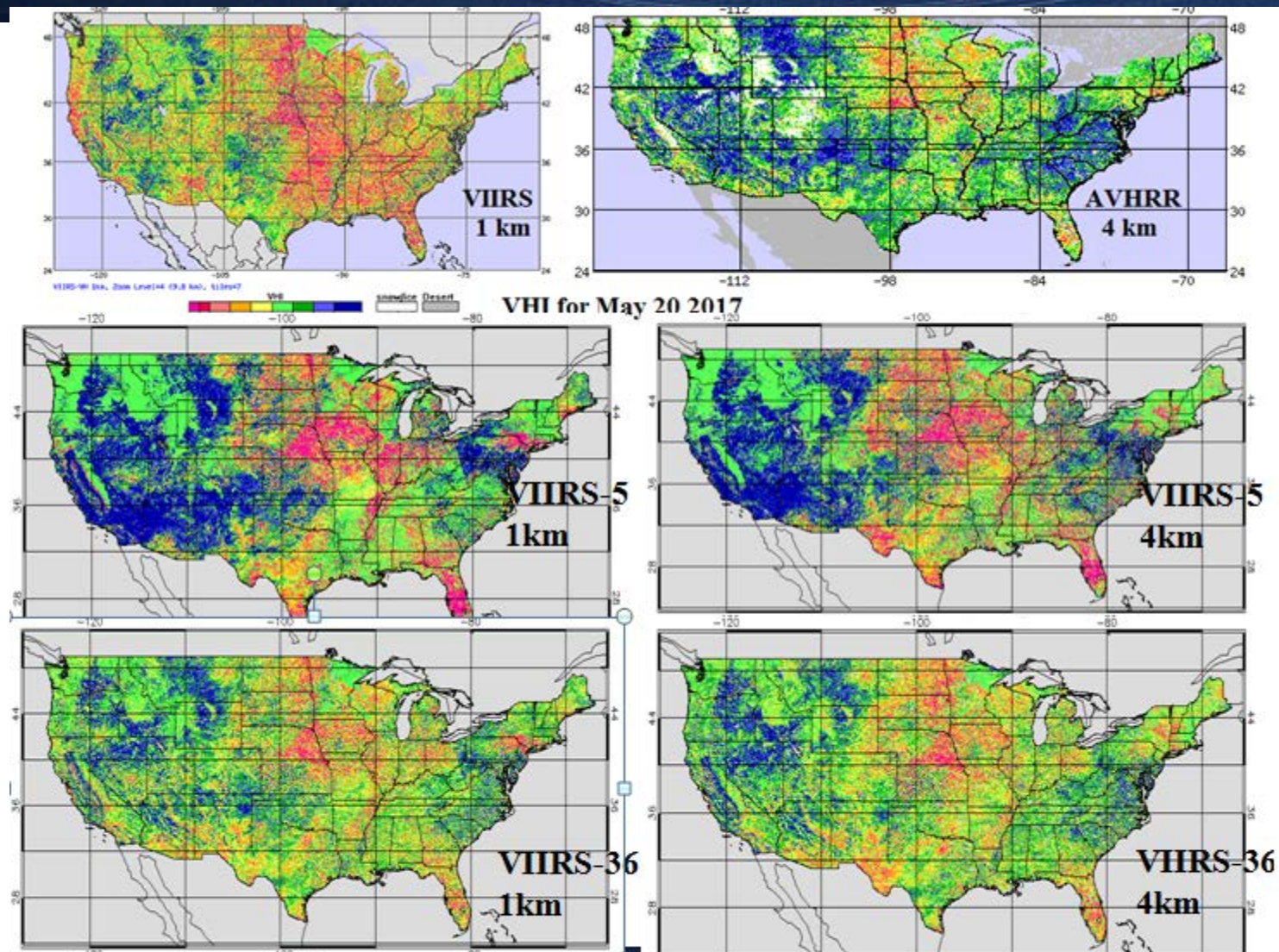
0.70



New VIIRS-VH system: Data Flow (Wz)



VHI from V-36, V-5 & A-36 climatology



VHI calculated with the V-36 climatology

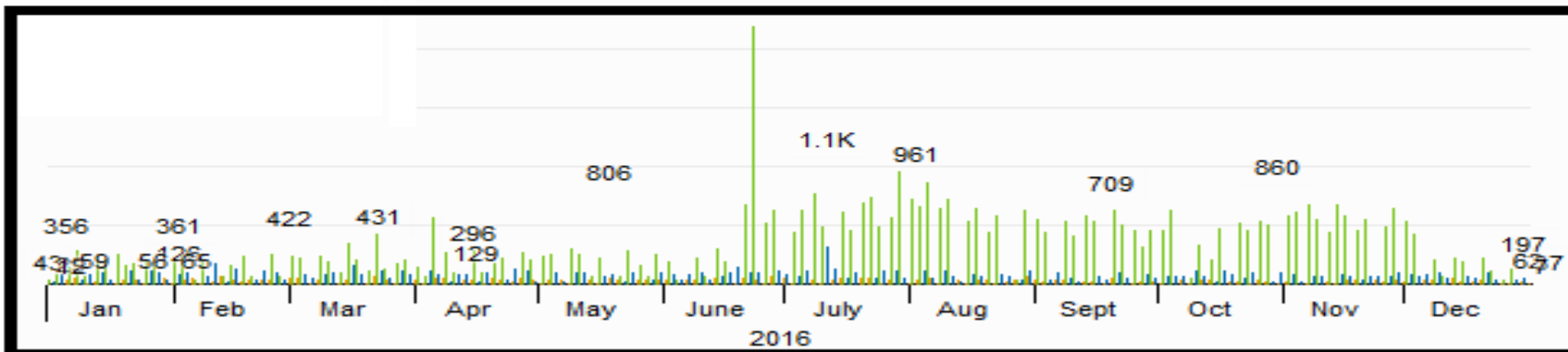


VH-WEB Users in 2016

Daily | [Weekly](#) | [Monthly](#) | [Quarterly](#) | [Yearly](#)

3.3K

- ☒ Page Views
- ☒ Unique Visits
- ☒ Returning Visits
- ☒ Labels on Chart



Date Range: ☒ Last Year ☐ 1 Jan 2016 - 31 Dec 2016

Chart Type: ☐ Line Chart ☐ Area Chart ☒ Bar Chart ☐ Save As Default



	Page Views	Unique Visits	First Time Visits	Returning Visits
Total	<u>140,163</u>	30,868	17,849	13,019
Daily Average	<u>383</u>	85	49	36

Vegetation Health (VH) Total Users in 2016 and Daily