DROUGHT

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Drought as Natural Disaster

- Drought (D) is a part of earth's climate
- D. occurs every year
- D. does not recognize borders, political & economic differences
- D. affects the largest number of people
- D. extremely costly

D. unique features

- Start unnoticeably
- Build-up slowly
- Develop cumulatively
- Impact cumulative & not immediately observable
- When damage is evident it's too late to mitigate the consequences
- Drought type: Meteorological, Agricultural, Hydrological, Socioeconomic

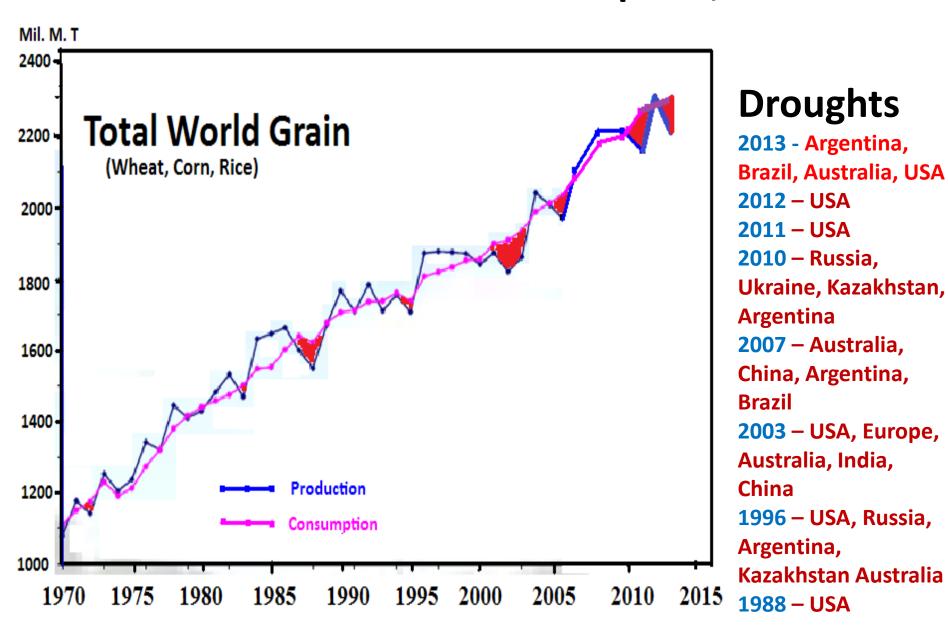
World Population Affected by Natural Disasters 1967-1991

	% _		
Disaster Type	Affected	Killed	
	Weather		
Drought	51	38	
Flood	38	9	
Hurricane etc.	8	27	
	Geological		
Earthquake	2	18	
Volcano	<1	<1	

Total People Affected: 2.8 billions

Total People Killed: 3.5 millions

World Grain Production-Consumption, 1970-2012



AFRICA: GTS (WMO-based) weather station network

On the average

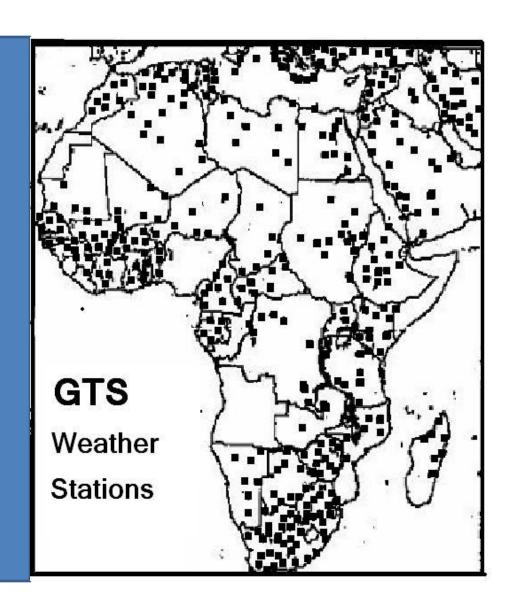
One weather station covers

31,000 sq. km

NOAA satellites cover

Current: 16 sq. km

S-NPP: 0.25 sq. km



Drought Data and Methods

- **Meteorological Data** (precipitation, temperature, humidity, soil moisture & their anomalies)
- Meteorological Indices Hydrothermal Coefficient index (GTK);
 Palmer Drought Severity Index (PDSI); Standardized Precipitation Index (SPI);
 The Standardized Precipitation (P)-Evapotranspiration Index (SPEI); The Drought

 Monitor; The new Forest Drought-Stress Severity Index
- **Satellite Indices** Normalized Difference Vegetation Index; Wide dynamic range vegetation index- WDRVI; Enhanced vegetation index EVI; Vegetation Health Indices (VH)
- Combination Indices: Satellite & In situ Vegetation Drought Response Index- VegDRI; Evaporative Drought Index (EDI);

What to Measure?

- Start
- End
- Intensity
- Duration
- Magnitude
- Area
- Season
- Origination
- Impacts

CHALLENGERS

- Unreliable D. forecasts
- D. Deficiency of water
- How to measure "Deficiency"?
- D. Multidimensional disaster
- Contributors: Ecosystem, Weather, Climate, Soil
- Contributors: Space, Time
- D. Difficult to validate: Indices are not validated
- D. Representation Too technical to understand
- Users need intensity, area, duration, start/end, impacts

Drought from Space: New Method

Sensors Advanced Very High Resolution Radiometer (AVHRR)

Visible Infrared Imaging Radiometer Suite (VIIRS)

Satellites NOAA: *NOAA-7, 9, 11, 14, 16, 18, 19*

SNPP/JPSS

Data Resolution Spatial - 1, 4 (GAC), 8 & 16 km (GVI);

Temporal - 7-day composite

Period 35-year (1981-2014)

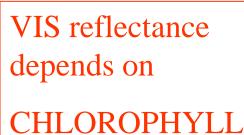
3-year (2011-2014)

Coverage World (75 N to 55 S)

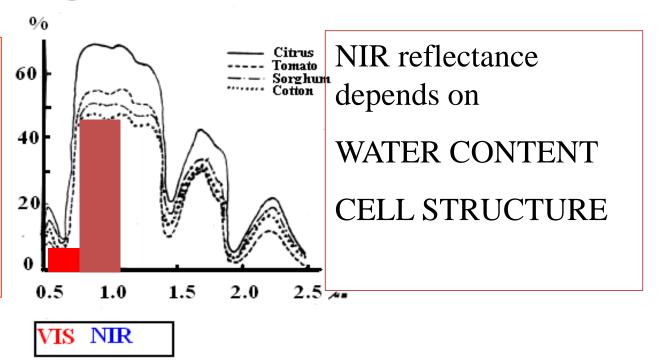
Channels VIS, NIR, Thermal

Typical Vegetation Reflectance

Vegetation Reflectance



CAROTENOID



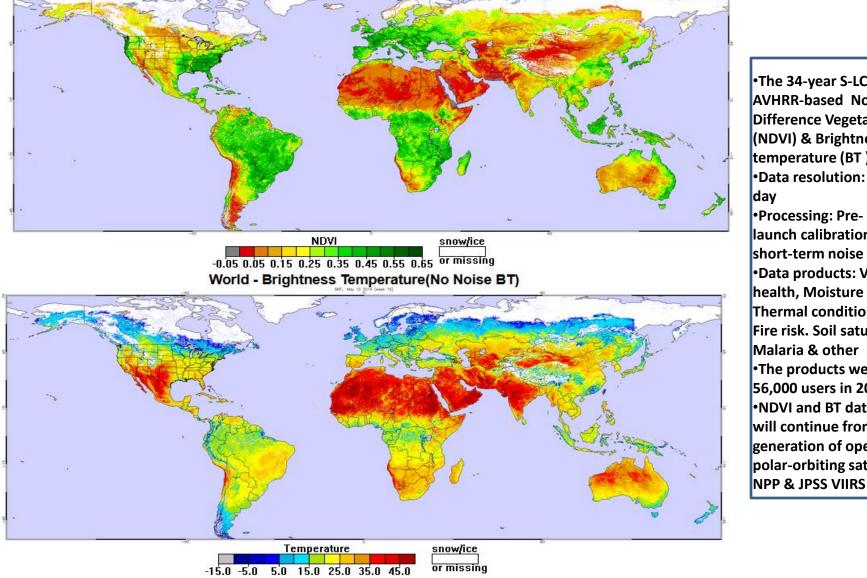
NDVI = (NIR-VIS)/(NIR+VIS)

NDVI & Reflectance

Cover Type	Ch1	Ch2	NDVI
Vegetation: Dense	.050	.150	0.500
Medium	.080	.110	0.140
Light	.100	.120	0.090
Bare Soil	.269	.283	0.025
Clouds	.227	.228	0.002
Water	.022	.013	-0.26

NDVI = (Ch2-Ch1)/(Ch2+Ch1)

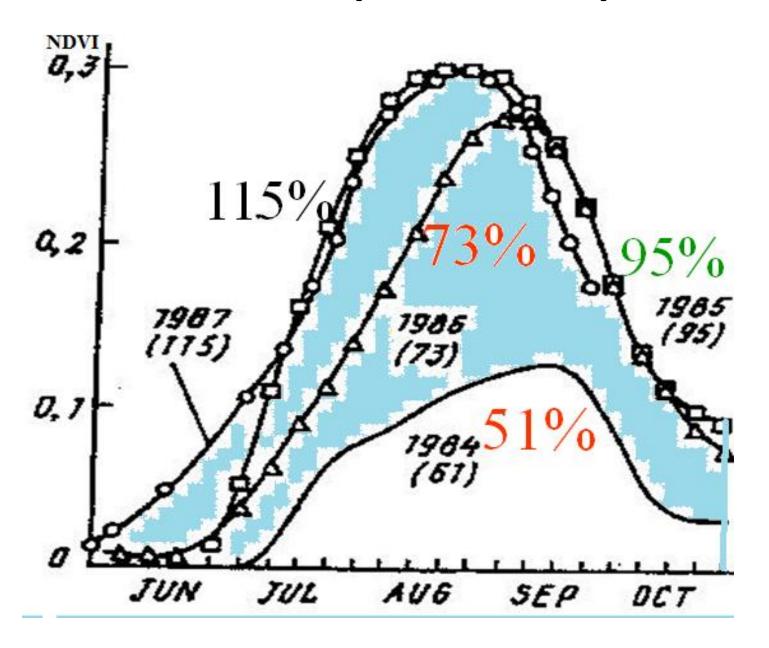
No noise NDVI and Brightness Temperature (BT) May 13, 2014



World - Greenness (No Noise NDVI)

The 34-year S-LCDR includes **AVHRR-based Normalized** Difference Vegetation Index (NDVI) & Brightness temperature (BT) Data resolution: 4 km & 7day Processing: Pre- & postlaunch calibration, long- and short-term noise removal Data products: Vegetation health, Moisture and Thermal conditions, Drought, Fire risk. Soil saturation, Malaria & other •The products were use by 56,000 users in 2013 NDVI and BT data records will continue from the new generation of operational polar-orbiting satellites: S-

NDVI & Rainfall (% of mean) SUDAN



NDVI-based Weather & Climate components, Ethiopia

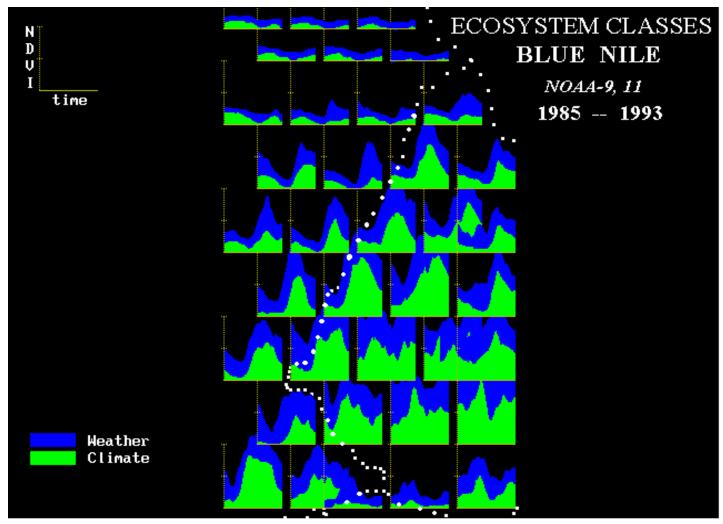


Fig. 6. Ecosystem classes from multi-year maximum and minimum NDVI in the Blue Nile basin

SATELLITE PRODUCTS

Vegetation condition index (VCI), values 0 - 100

VCI=(NDVI-NDVImin)/(NDVImax-NDVImin)

NDVImax, and NDVImin – climatology (1981-2000 maximum and minimum NDVI for a pixel;

MOISTURE

Temperature condition index (TCI), values 0 - 100

TCI=(BTmax-BTmin)/(BTmax-BTmin)

NDVImax, and NDVImin – climatology (1981-2000 maximum and minimum NDVI for a pixel

THERMAL

Vegetation Health Index (VHI), values 0 – 100

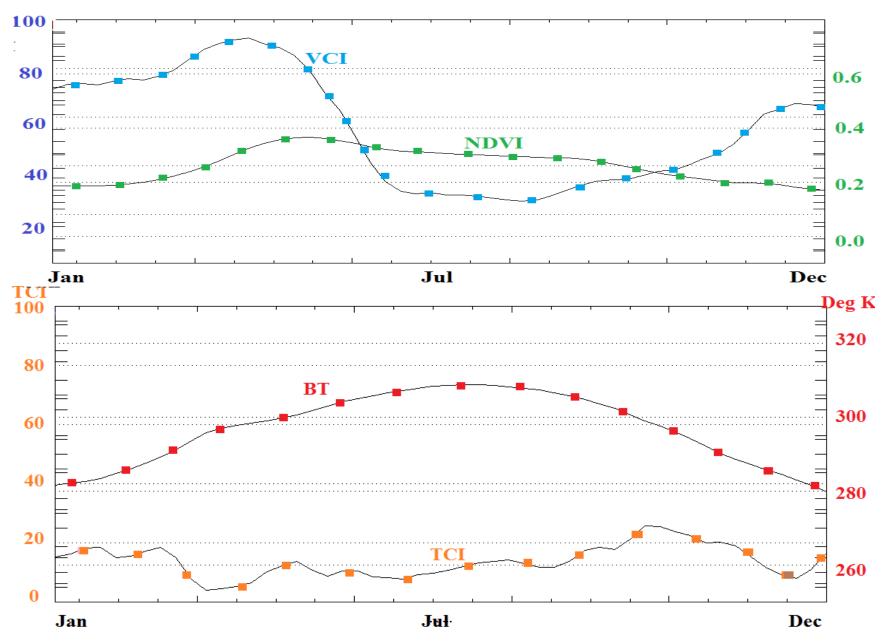
VHI=a*VCI+(1-a)*TCI

VEG. HEALTH

0 – indicates extreme stress

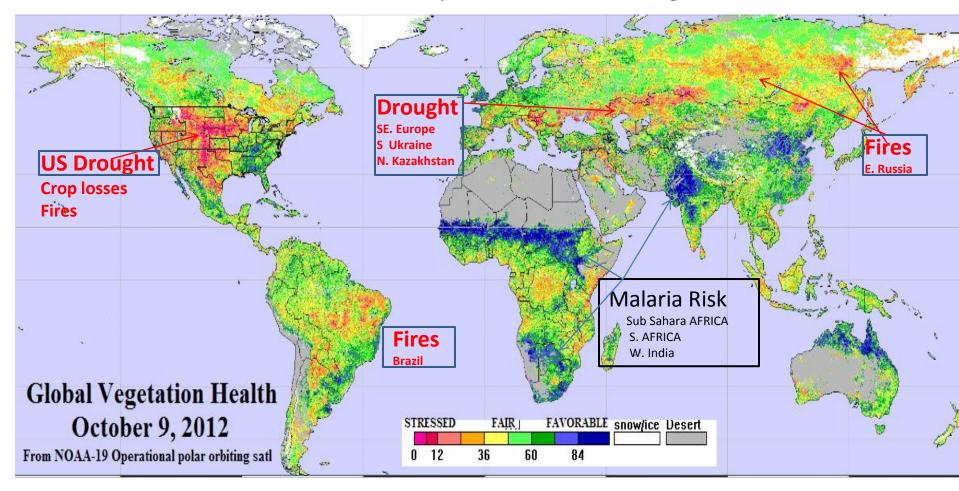
100 – indicates favorable conditions

VCI vs NDVI & TCI vs BT, 2012, Kansas



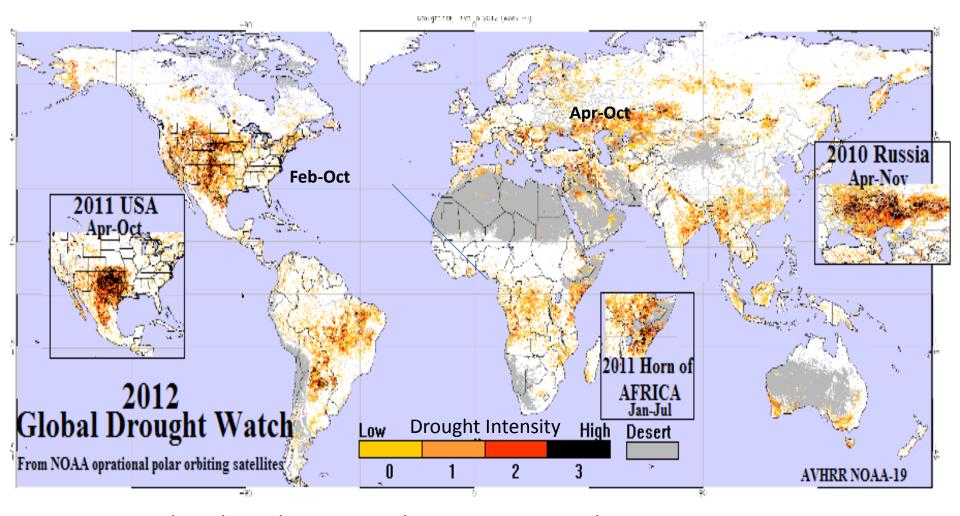
2012 Global Vegetation Health (VH)

From AVHRR/NOAA-19 Operational Polar Orbiting Satellite



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

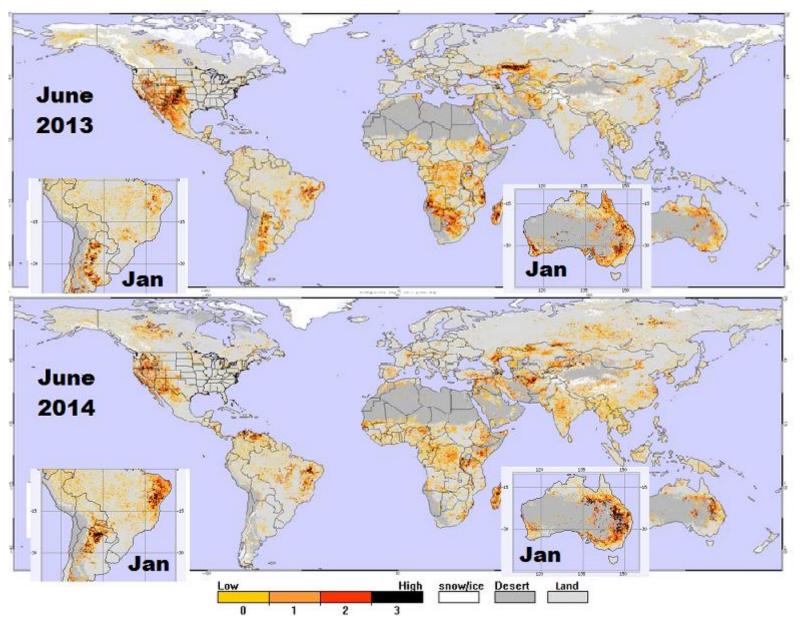
Drought-related Vegetation Stress: 2010-2012



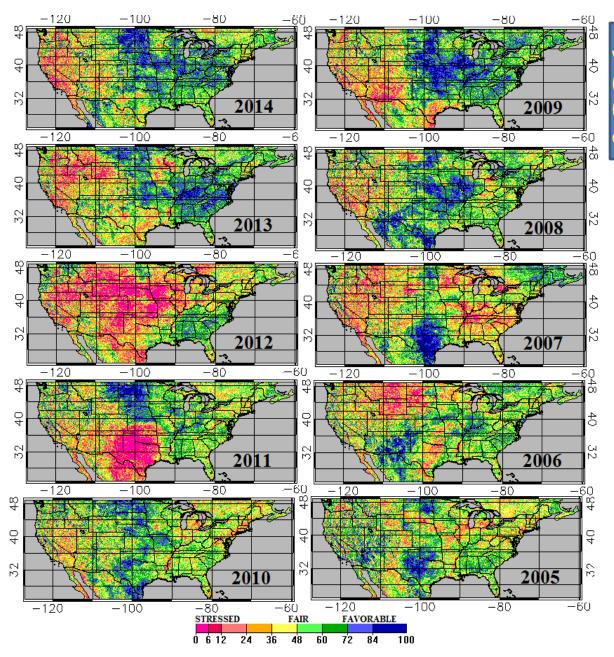
- 2012- Extreme drought in the USA, southern UKRAINE, northern KAZAKHSTAN,
 - Severe drought in eastern INDIA & southern China, Kenya & South America
- 2011 Exceptional drought in Texas (USA) and the Horn of AFRICA
- 2010 Exceptional drought in RUSSIA and UKRAINE

Drought-related Vegetation Stress 2013-2014

from satellite-based Vegetation Health

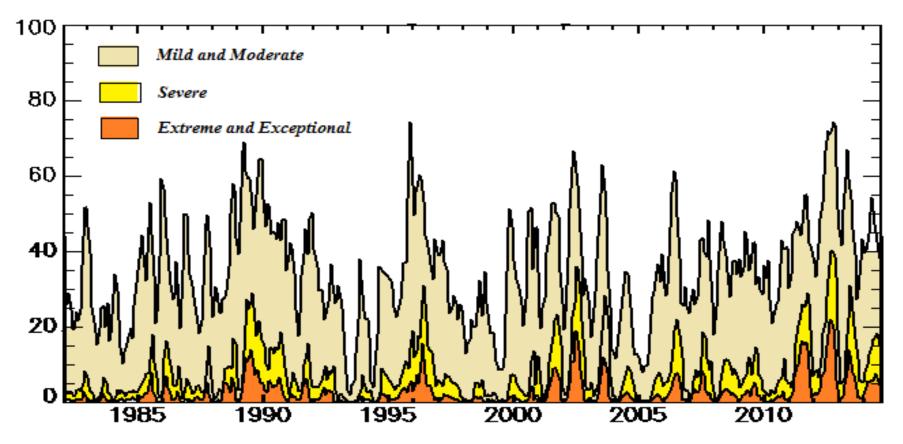


9-year Drought in Western USA from Vegetation health (VH)



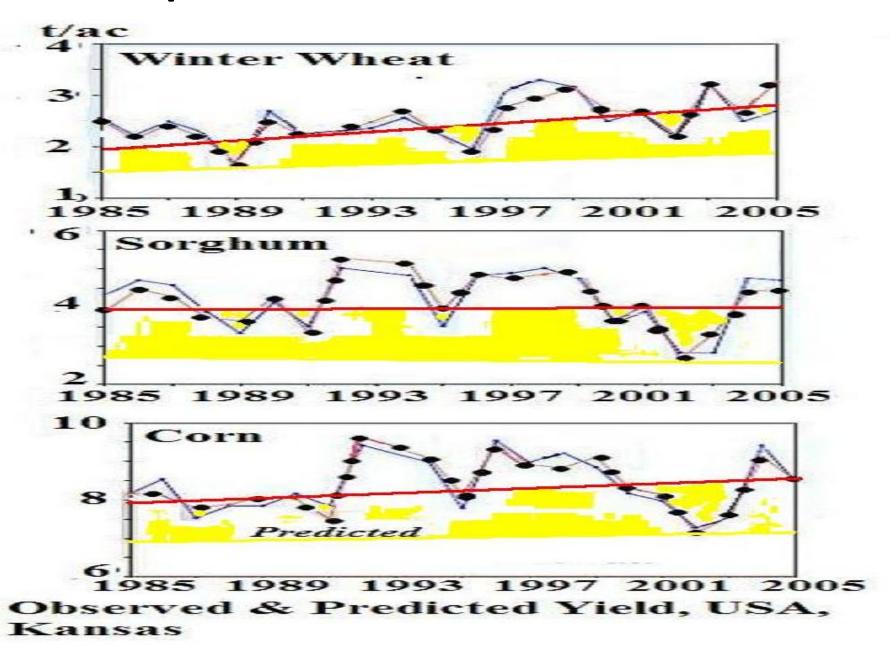
VH estimates vegetation health (condition) from extreme stress (dark red) to favorable conditions(dark blue)

Drought area for three intensities, 1982-2014 Western USA

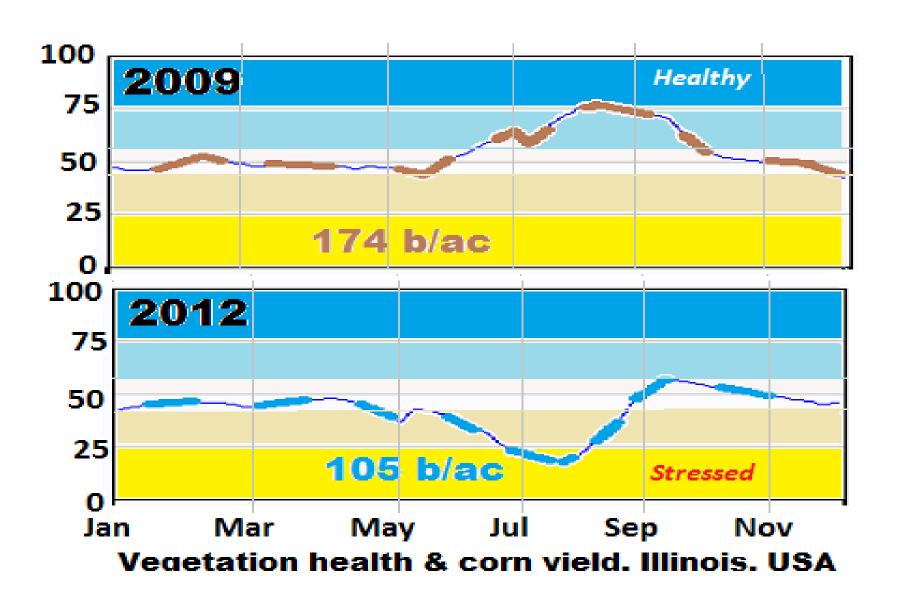


Drought Area & Intensity by weeks: Western United States, 1982-2014

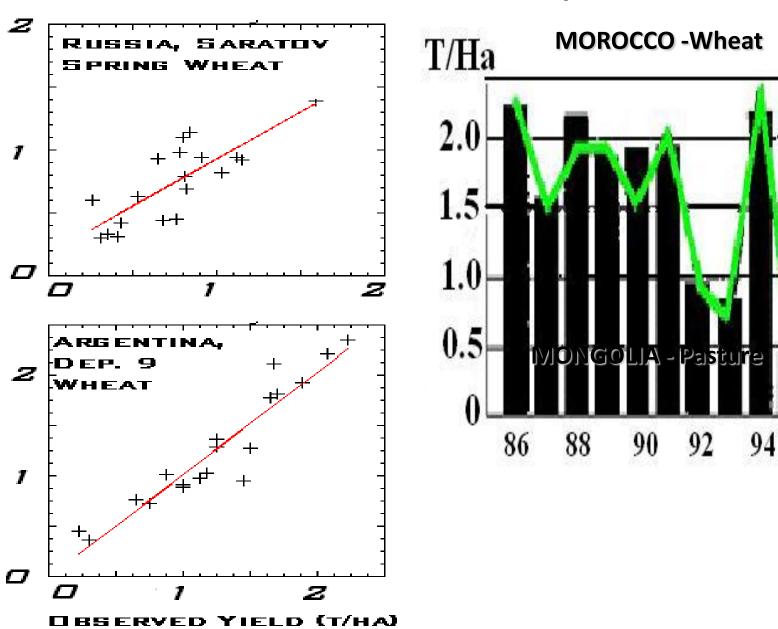
VH-Crop Losses Prediction: USA, Kansas



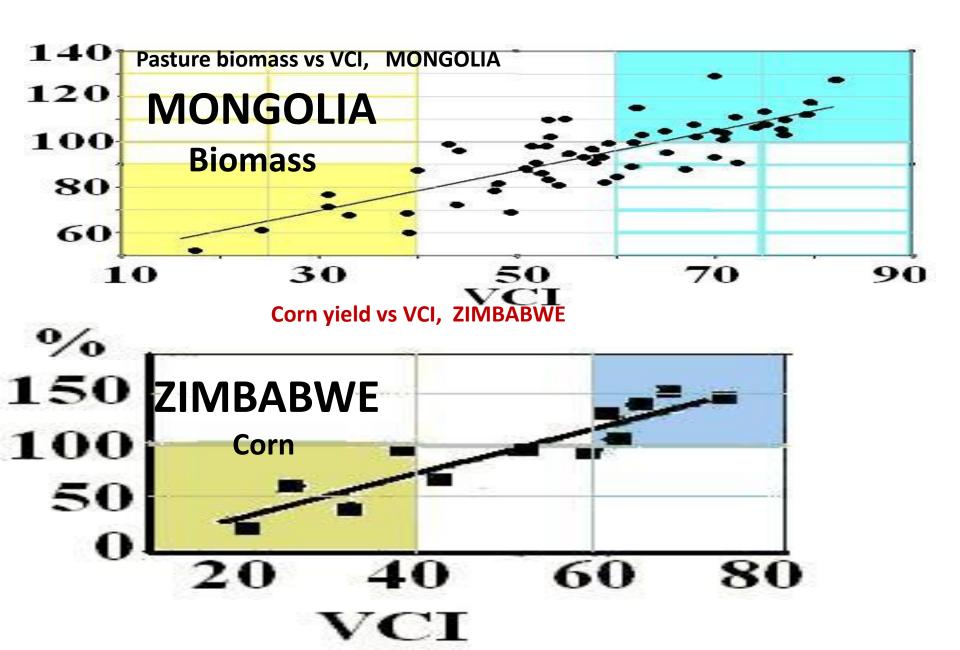
VH & Corn Yield, Illinois, USA



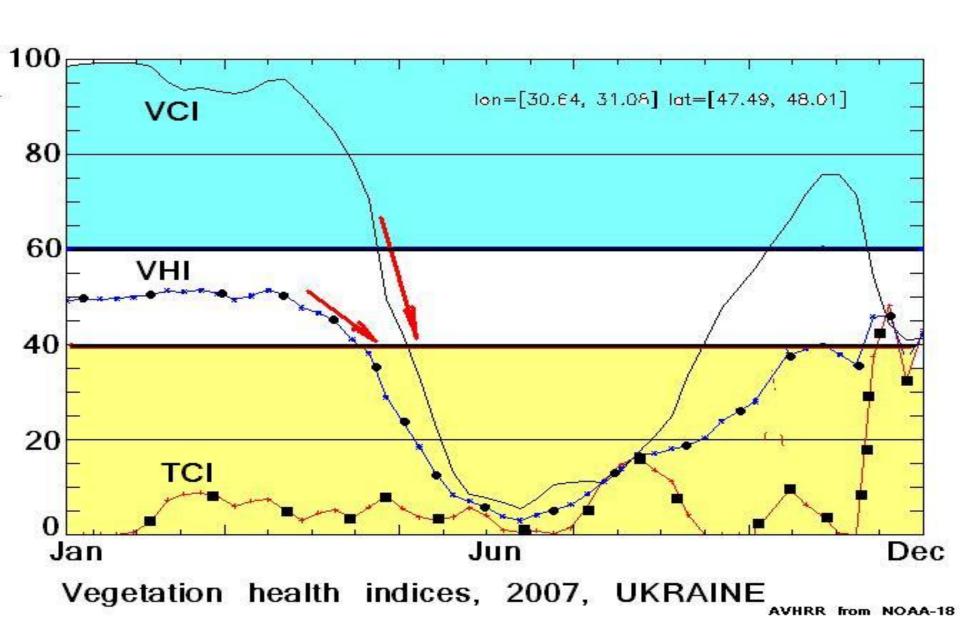
VH and Crop Yield



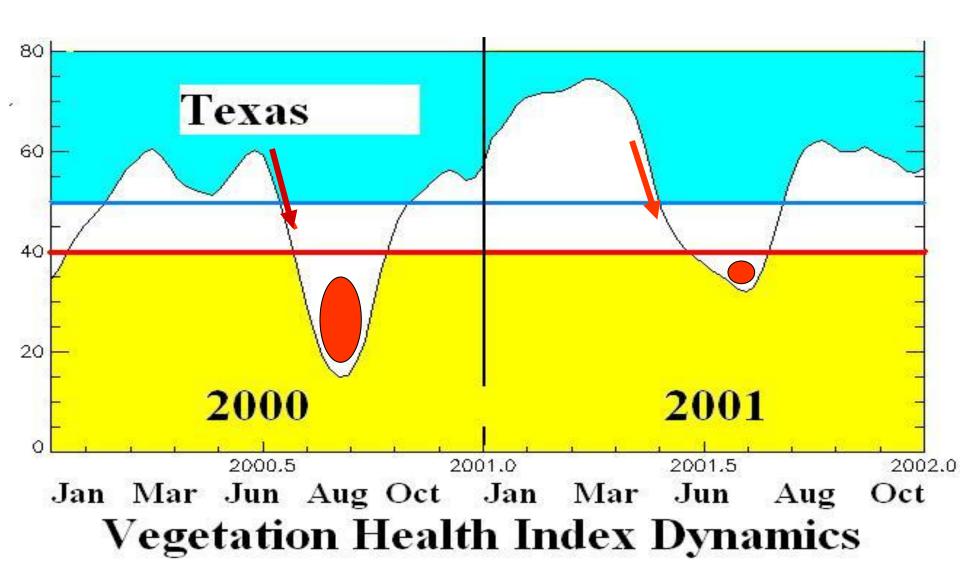
VH-Biomass & Corn Yield Modeling & Prediction



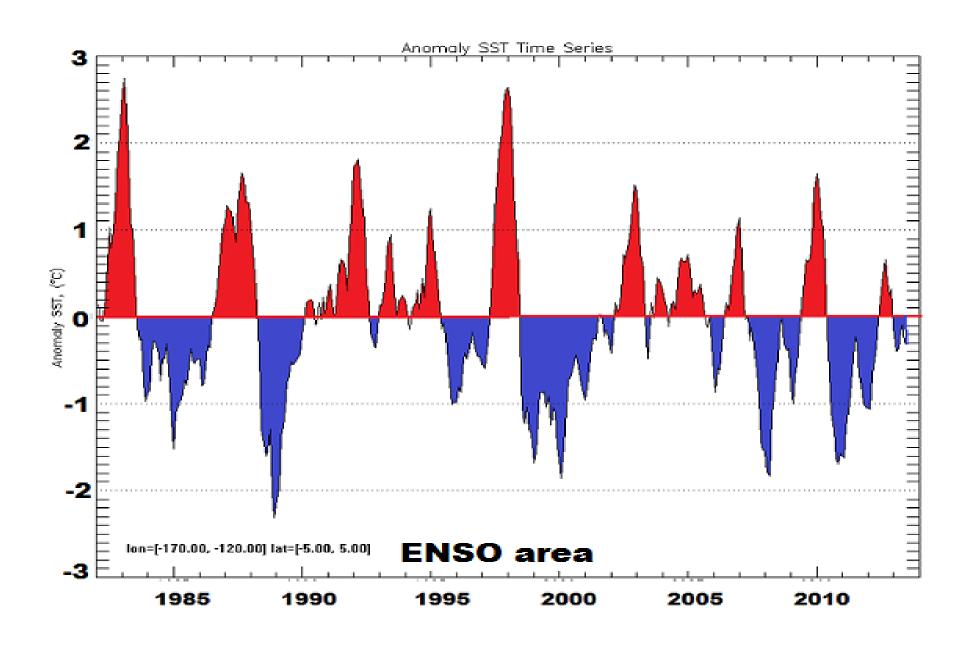
2007 Vegetation Health, UKRAINE



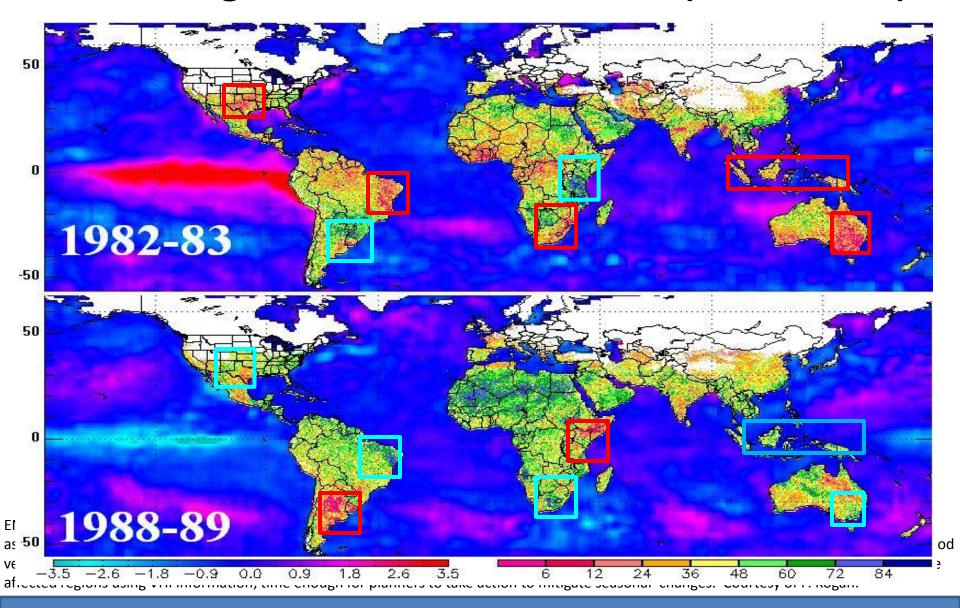
VH-Drought Prediction (1-2 months)



SST Anomaly in 3.4 ENSO

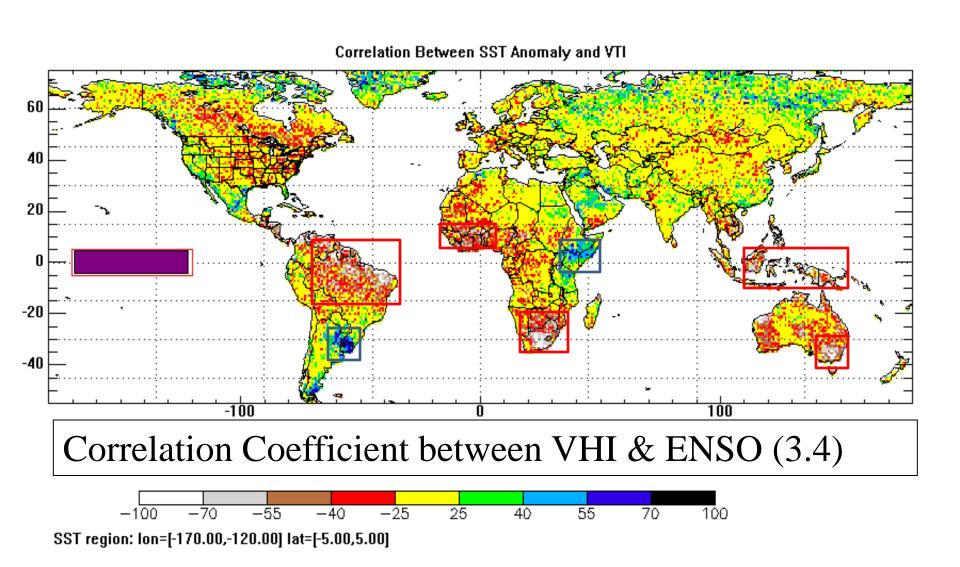


VH-Drought Prediction from ENSO (3-6 months)

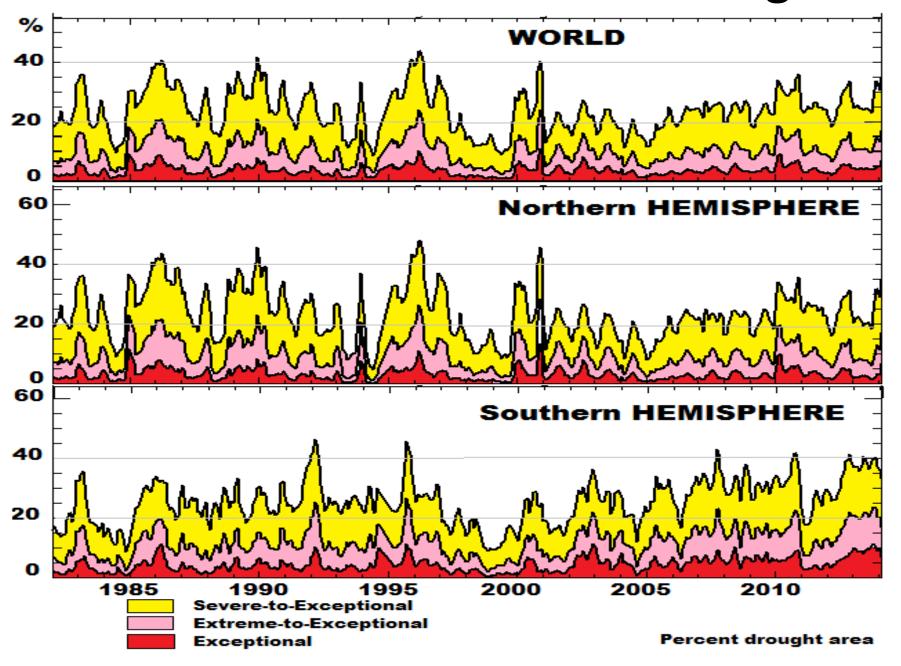


Sensitivity of Ecosystems to ENSO

Correlation of VH with SST anomaly in Tropical Pacific



Climate: Percent Land under Drought



Drought portal on NOAA-Web

http://www.orbit.nesdis.noaa.gov/smcd/emb/vci

Every week on Tuesday

VH Products

DISASTERS (droughts, soil saturation)

FORESTRY (fire risk)

AGRICULTURE (production, losses)

LAND CONDITION (moisture, thermal)

CLIMATE FORCING (ENSO)

HUMAN HEALTH (epidemics)

WATER (irrigation)

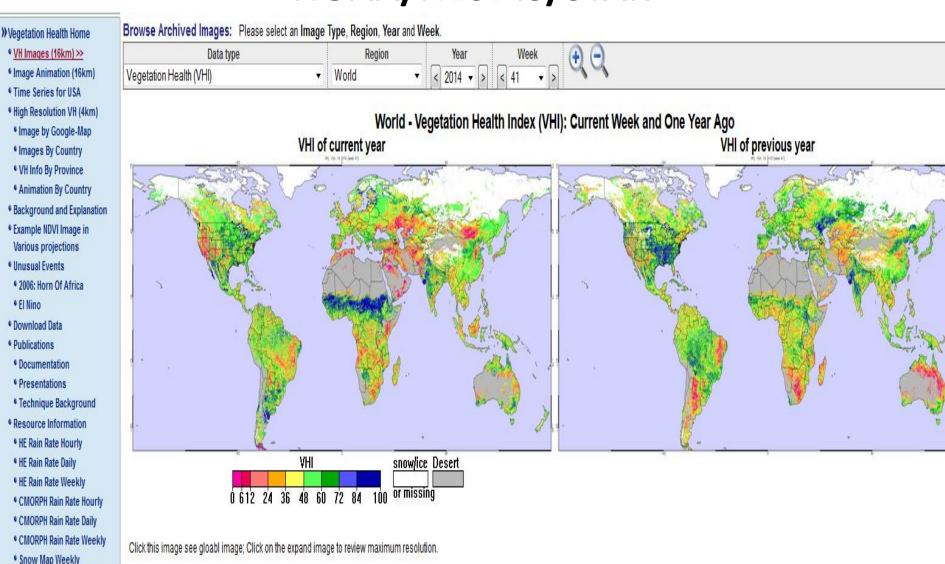
NWS MODELS (vegetation fraction)

ECOSYSTEMS (distribution & change)

CLIMATE CHANGE

DATA RECORDS

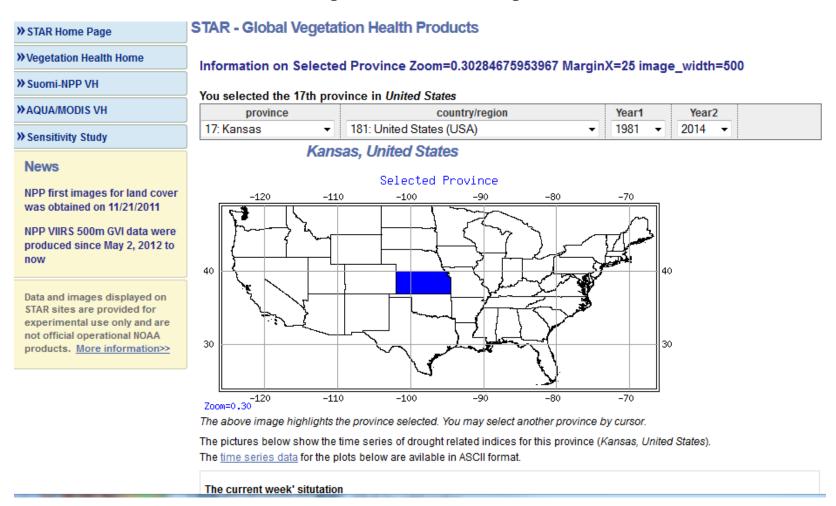
Drought & Vegetation Health Portal-1 NOAA/NESDIS/STAR



star.nesdis.noaa.gov/smcd/emb/vci/gvix_webImages/2014/WorldBig_VHI_201441.png.

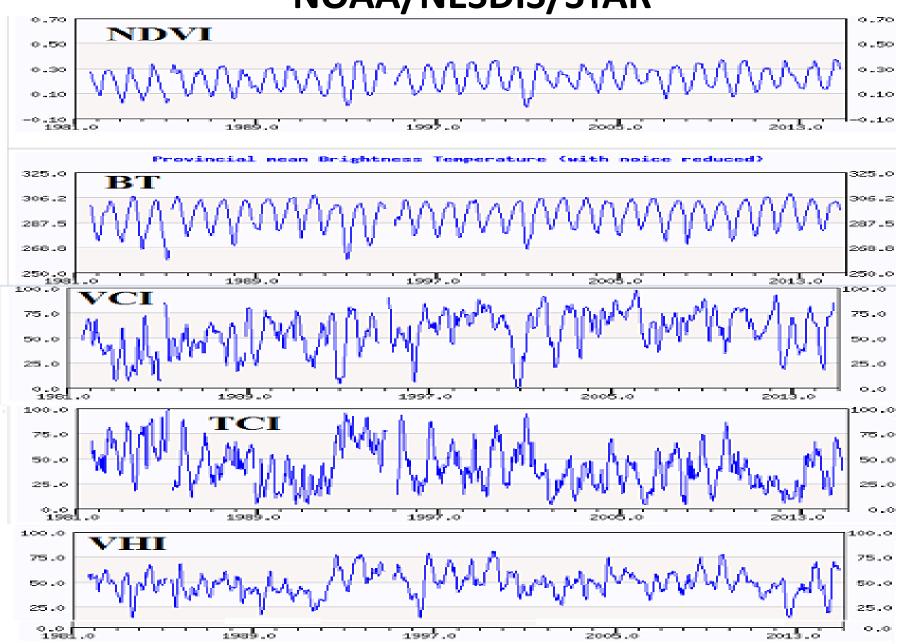
Faustorial Crossina Time

Drought & Vegetation Health Portal-2 NOAA/NESDIS/STAR



http://www.star.nesdis.noaa.gov/smcd/emb/vci/VH/vh_browseByCountry_province.php?country=USA&x=249&y=123

Drought & Vegetation Health Portal-3 NOAA/NESDIS/STAR

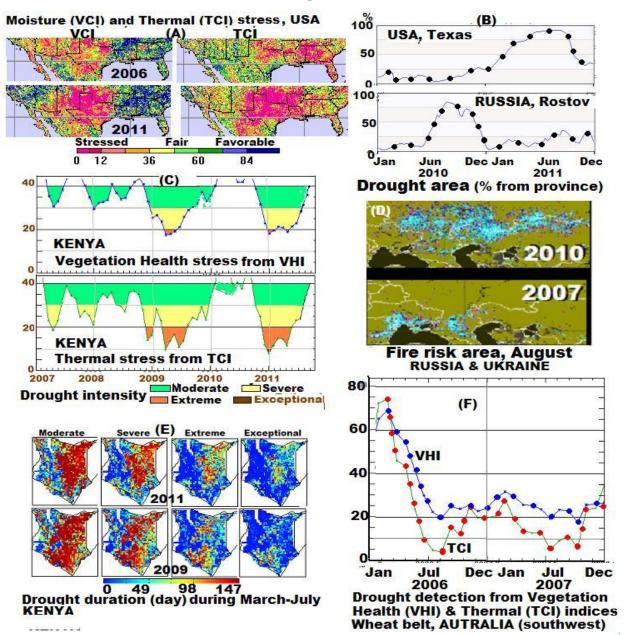


VH Applications

http://www.orbit.nesdis.noaa.gov/smcd/emb/vci

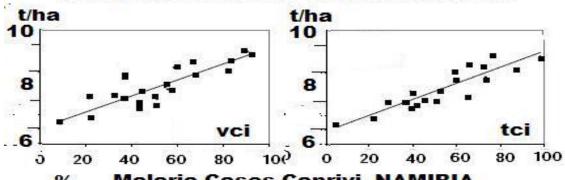


- (A) Moisture & Thermal stress
- B) Drought area
- (C) Intensity of vegetation stress
- (D) Fire risk
- (E) Drought duration
- (F) Drought detection/prediction



VH Applications

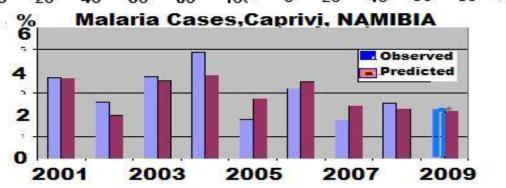
Corn yield versus VCI & TCI, Kasas, USA

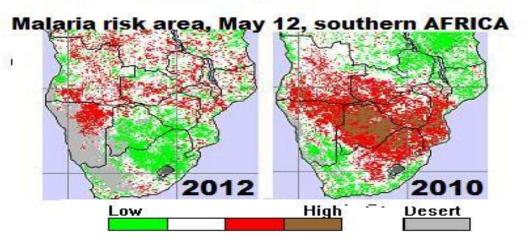


APPLICATIONS

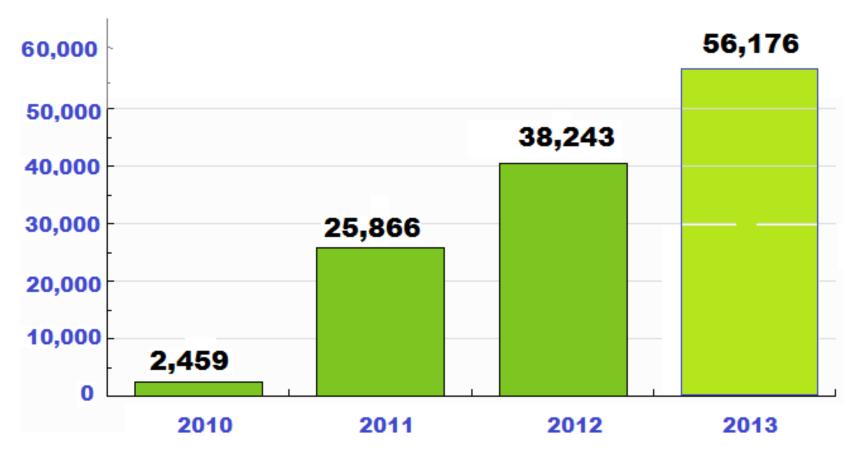
Crop/Pasture Production Malaria:

Number of affected people
Affected area
Intensity





VH WEB Users' number



WEB page total users from a day of calculation started in Nov 2010

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

Conclusions

- 2014: World Population (7.3 bil.) Increases with Accelerating Rate; World Grain Production Increases with Decelerating Rate
- **Grain supply drops below demands (in the 21st century 8 years out of 13)**
- <u>Severe Droughts</u> Reduces Global Grain Production 4-7% every 4-6 years Moderate Drought – Reduces Grain Yield up to 3% every 2-3 years
- <u>Satellite-based Vegetation Health (VH)</u> Technology Provide Tools for Drought Monitoring & 1-2 Month Advanced Prediction of its Start
- VH Provide Prediction of Drought-related Crop & Pasture Losses: (a) 3-6 Months in Advance of Harvest, (b) During ENSO years 8-12 months
- VH can be used for predictions & assessments
 - **Prediction of Drought start**
 - **Assessment of Drought Area & Intensity**
 - **Crop and pasture production prediction**
 - Land cover change
 - **Global & Regional climate warming dynamics**
 - **Provide assessment of Regional Food Security**

Principal Publications since 2010

- 2015 Felix Kogan and Wei Guo, 2015: Agricultural Drought monitoring using space-derived biophisical products: a global perspective. Advances in Land Remote Sensing: Last 50 Year (chapter 13 of the Book)
- 2014 Kogan, F. and W.Guo 2014, Early twenty-first-century droughts during the warmest climate. Geomatics, Natural Hazards and Risk. http://dx.doi.org/10.1080/19475705.2013.878399
- **2013 Nizamuddin, M., F. Kogan, R. Dihman, W. Guo, L. Roytman, 2013:** Modeling and Forecasting Malaria in Tripura, India using NOAA/AVHRR-Based Vegetation Health Indices. *Int. J. Rem. Sens. Applications*, Vol. 3, Issue 3, 108-116.
- 2013 Kogan, F., N. Kussul, T. Adamenko, S. Skakun, O. Kravchenko, O. Kryvobok, A. Shelestov, A. Kolotii, O. Kussul, A. Lavrenyuk, 2013: Winter wheat yield forecasting in Ukraine based on Earth observation, meteorological data and biophysical models. *International Journal of Applied Earth Observation and Geoinformation* 23 (2013) 192–203
- **2013 Kogan, F., T. Adamenko and W. Guo, 2013.** Global and Regional Drought Dynamics in the Climate Warming Era. *Rem. Sens.Letters.* Vol 4, Issue 4, 364-372
- **2012 Kogan, F., L. Salazar and L. Roytman, 2012**: Forecasting crop production using satellite-based vegetation health indices in Kansas, USA. *J. Rem. Sens.* **33(9)**, 2798-2819.
- **2012 Rahman, A., K. Khan, N. Y Krakauer, L. Roytman, F. Kogan, 2012:** Use of Remote Sensing Data for Estimation of Aman Rice Yield. *International Journal of Agriculture and Forestry*, 2(1): 101-107 (DOI10.5923/j.iaf.20120201.16).
- **2011 Rahman, A., F. Kogan, L. Roytman, M. Goldberg and W. Guo, 2011.** Modeling and prediction of malaria vector distribution in Bangladesh from remote sensing data. *Int. J. Rem. Sens.*. Vol 32, No. 5, 1233-1251.
- **2011 Atiqur Rahman, Leonid Roytman, Mitch Goldberg, and Felix Kogan, 2011**: Comparative Analysis on Applicability of Satellite and Meteorological Data for Prediction of Malaria in Endemic Area in Bangladesh. *Am. J. Trop. Med. Hyg.,* 82(6), pp. 1004–1009.
- 2011 Felix Kogan, Alfred Powell and Oleg Fedorov (Editors), 2011: *Use of Satellite and In-Situ Data to Improve Sustainability.* Springer, 314 pp.
- **2010 Bhuyan C. and F. Kogan, 2010**: Monsoon Variations and Vegetative Drought Patterns in the Luni Basin under Rain-shadow Zone. *Int. J. Rem. Sens.*. **Vol 31, No 12,** pp. 3223-3242
- 2010 Atiqur Rahman, Nir Krakauer, Leonid Roytman, Mitch Goldberg and Felix Kogan 2010: Application of Advanced Very High
 Resolution Radiometer (AVHRR)-based Vegetation Health Indices for Estimation of Malaria Cases. Am. J. Trop. Med. Hyg., 82(6), pp. 1004–
 1009
- 2010 Xiaoyang Zhang, Mitchell Goldberg, Dan Tarpley, Mark Friedl, Jeffrey Morisette, Felix Kogan, Yunyue Yu 2010: Drought-induced Vegetation Reduction in Southwestern North America. Env Res. Let. Vol. 5, 11 p.
- 2010 Felix Kogan 2010: ENSO impact on vegetation. *Use of Satellite and In-Situ Data to Improve Sustainability.* Springer, pp. 165-174.
- **2010 Kushnir, V., F. Kogan G. Korotaev, and A. Powel 2010**: Predicting natural disasters in the coastal zone based on links between land and marine ecosystems. *Geomatics, Natural Hazards and risk.* Vol. 1, No. 3, September. pp 185-198, (ISSN: 1947-5705)
- **2010 Felix Kogan, Wei Guo, and Aleksandar Jelenak 2010:** Global Vegetation Health: Long-Term Data Records. *Use of Satellite and In-Situ Datato Improve Sustainability.* pp. 247-256.