ECOSTRESS METRIC
EVAPOTRANSPIRATION
ALGORITHM

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HyspIRI TQ3 Science Support

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Justin Huntington, Desert Research Institute
Baburao Kamble, Univ. Nebraska-Lincoln
Definitions

• ECOSTRESS - ECOsystem Spaceborne Thermal Radiometer Experiment on Space Station

• METRIC – Mapping EvapoTranspiration at high Resolution using Internalized Calibration

Applications

• METRIC is used for water resources management and planning and water rights management by the States of:
  • Idaho, Montana, Oregon, Nevada, New Mexico, Wyoming, Nebraska, North Dakota and Washington

• by entities including
  • Metropolitan Water District of Southern California
  • Gallo Wines

• in litigation including
  • Montana vs. Wyoming (US Supreme Court)
  • Texas vs. New Mexico (US Supreme Court)
  • Antelope Valley vs. City of Los Angeles (impacts of ground-water pumping)
  • Upper Klamath River basin of Oregon and California (endangered species v. irrigation)
ET is needed at the field scale and for historical and present...
Science Questions on ECOSTRESS relative to ET and METRIC

• How should thermal/energy balance-driven/end-member calibration algorithms such as METRIC be evolved to work accurately and consistently in the precessing west-to-east orbit of the International Space Station (ISS)?

• What sizes of automated moving-windows best provide the statistical description for end member-based calibration given:
  • continually varying sun angle and radiation energy balance
  • continually varying impacts of wind speed on convective fluxes of sensible heat
  • continually varying residual evaporation rates from precipitation wetting events
  • high seasonal variation in vegetation dynamics ranging from heat and drought to partial snow cover
  • complex terrain

• What are best time-integration strategies to produce ET maps showing total fluxes over periods of weeks and months that are meaningful and useful to water managers and studies of water behavior of ecosystems?
Two Primary Families of Energy Balance:

- Forcing / boundary layer growth models
  - Federal GDAS / LDAS / LIS systems
  - NOAH – NOAA
  - MOSAIC – NASA
  - VIC – Letenmeier
  - HIRLAM (KNMI, The Netherlands, Europe)
  - Alexi two source – Norman, Anderson, Kustus

- Continuous in time
- Driven by gridded weather data
- Informed by Remote Sensing
- Large grid sizes (1 km – 32 km)

- Snapshot models
  - SEBAL
  - SEBI
  - SEBS
  - REEM
  - disAlexi
  - METRIC
  - etc (derivatives)

- Discrete in time
- Driven by Remote Sensing
- Smaller grid sizes (30 m – 1 km)

\[ \text{ET} = R_n - G - H \]

"Snapshots" will, in effect, become ‘movies’ with ECOSTRESS
Three major components of the Surface Energy balance are driven by LST (sensible heat flux, net long-wave radiation, ground heat flux).
Example Portion of METRIC “Mountain” model coded in ERDAS ModelMaker

METRIC model is coded in ERDAS ModelMaker, Python, and Google Earth Engine JavaScript
Carson Valley NV - Comparison of METRIC ET in Wet and Dry Years 2006 and 2009 to Support USGS Groundwater Modeling Efforts

 Analyses by Dr. Justin Huntington Res. Group, DRI

• GW Model average annual ET = \textbf{3.0 ft} and is being used for all years no matter if Carson River flow is high or low
• Assisting USGS to refine this assumption in ongoing transient modeling efforts

• METRIC Average ET 2006 = \textbf{3.5 ft}
• METRIC Average ET 2009 = \textbf{2.6 ft}

• 2006 = 150\% of normal flow
• 2009 = 80\% of normal flow
• Yearly ET a function of supply \textbf{AND} Demand – 2009 was a cool spring…
2011 Blind Comparison of Automated Daily ET to Measured Daily ET in Carson and Mason Valley, NV / CA

Analyses by Dr. Justin Huntington Res. Group, DRI
2011 Blind Comparison of METRIC ET to Measured ET – Nevada

- Daily error is not bad when considering the error in measured daily ET is ~ +/- 20%

- Whiskers on X = +/- 12% USGS estimated uncertainty in measured ET, Y = +/- 95% confidence interval of 100 automated estimates of ET using different input parameters

- Over a season the error in daily estimates largely cancels out

Analyses by Dr. Justin Huntington Res. Group, DRI
2014 Blind Intercomparison of RS of ET Models and Measurements – Palo Verde, CA

Accuracy Requirements

Five Remote Sensing of ET models

Five different results

Example of Results

Estimates of actual ET for May 10th, 2008 (DOY 131)

Comparison Analyses by Dr. Christopher Neale and Hatim Geli, Utah State University
2014 Blind Intercomparison of RS of ET Models and Measurements – Palo Verde, CA

Summary of Intercomparison

Comparison against Ground-Flux Measurements over Alfalfa

<table>
<thead>
<tr>
<th>Method</th>
<th>RMSE</th>
<th>BIAS</th>
<th>BIAS (%)</th>
<th>MAE</th>
<th>Sample</th>
<th>Mean</th>
<th>Std. Dev.</th>
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<tbody>
<tr>
<td>Measured</td>
<td>1.5</td>
<td>-0.2</td>
<td>-7.2%</td>
<td>1.3</td>
<td>16</td>
<td>6.55</td>
<td>2.4</td>
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<tr>
<td>SSEBop</td>
<td>2.7</td>
<td>-2.5</td>
<td>-42.0%</td>
<td>2.5</td>
<td>16</td>
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<tr>
<td>SEBS</td>
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<td>6.45</td>
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<td>METRIC</td>
<td>1.3</td>
<td>-0.8</td>
<td>-9.8%</td>
<td>1.1</td>
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<td>5.70</td>
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<td>ReSET</td>
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<td>-1.7</td>
<td>-22.9%</td>
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<td>16</td>
<td>4.87</td>
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<td></td>
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<tr>
<td>DisALEXI</td>
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<td></td>
<td></td>
<td></td>
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</tr>
</tbody>
</table>

Comparison Analyses by Dr. Christopher Neale and Hatim Geli, Utah State University
Accuracy Requirements

2014 Blind Intercomparison of RS of ET Models and Measurements – Palo Verde, CA

Seasonal Water Balance

<table>
<thead>
<tr>
<th>Water balance Component</th>
<th>Depth (mm/year)</th>
<th>DisALEXI</th>
<th>ReSET</th>
<th>METRIC</th>
<th>SSEBop</th>
<th>SEBS</th>
<th>P-T</th>
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<tbody>
<tr>
<td>Precipitation</td>
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<tr>
<td>Total Inflow</td>
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<td>2550</td>
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<td>2550</td>
<td>2550</td>
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<tr>
<td>Canal Spills</td>
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<td>284</td>
<td>284</td>
<td>284</td>
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<tr>
<td>Drainage</td>
<td>998</td>
<td>998</td>
<td>998</td>
<td>998</td>
<td>998</td>
<td>998</td>
<td>998</td>
</tr>
<tr>
<td>ET</td>
<td>(1000)</td>
<td>956</td>
<td>1223</td>
<td>952</td>
<td>X</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Total Outflow</td>
<td>(2282)</td>
<td>2238</td>
<td>2505</td>
<td>2234</td>
<td>X</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Inflow-Outflow</td>
<td>(268)</td>
<td>312</td>
<td>34</td>
<td>316</td>
<td>(-12.2%)</td>
<td>(-1.8%)</td>
<td>(-12.4%)</td>
</tr>
</tbody>
</table>

Conclusion – The specific endmember calibration approach of METRIC tends to function well and consistently

Comparison Analyses by Dr. Christopher Neale and Hatim Geli, Utah State University
Olives in Chile

Study area is in the center of Chile
Path 233, Row 85, Landsat 7 processing (2011 & 2012)

Collaboration between Samuel Ortega, Univ. Talca and Ayse Kilic, UNL and UI Remote Sensing Research groups
New olive production in central Chile with relatively dense tree spacing.

ET fluxes were measured using an eddy covariance system mounted above the crop.

Photos by Dr. Samuel Ortega, Univ. Talca, Chile, collaboration with A. Kilic.
Olives have low ground-cover and leaf area, making ET estimation challenging. 1000 m blockiness in ET (right) is caused by land use classification error in the Global Land Cover product.

- We have made some enhancements to the model (Perrier function for surface roughness, Leaf Area Index)
- Correction to the global land cover map

Analyses by Samuel Orlando Ortega Salazar, with UNL group
METRIC-ECOSTRESS

Porting Algorithms
Sun angle continually changing
Land / precipitation conditions continually changing

Orbital speed
17000 mph
7660 meters/second

Altitude: 420 km
Topography of the United States and World is Complex – impacting solar radiation balance.

METRIC includes radiation algorithms for slopes and terrain roughness algorithms.
Upper Klamath River Basin of Oregon, California

- 2013 Settlement on water rights
- used ET maps from METRIC as basis for negotiation
- using METRIC for near-real time monitoring of retired irrigated parcels

Evapotranspiration, October 3, 2014
Satellite Energy Balance is ‘Challenged’ by Uncertainty and Bias in components – These are reduced by use of Endmember Calibration Based on Maximum and Minimum ET:

- Surface temperature
  - Aerodynamic vs. Radiative Temperature
  - Bias in Satellite Sensor Calibration
  - Atmospheric Correction
- Air temperature *(varies with H)*
- Albedo calculation
- Net radiation calculation *(incoming long-wave)*
- Soil heat flux
- Aerodynamic resistance calculation
  - Aerodynamic roughness for momentum
  - Aerodynamic roughness for sensible heat flux
  - Buoyancy effects
- Wind speed field
- Extrapolation of instantaneous ET to 24-hour periods
METRIC-ECOSTRESS

Student Collaboration
Student Collaboration

• Participation in development of strategies and coding and testing of algorithms for applying a moving window-based calibration scheme to develop endmember based calibration points for the surface energy balance employed in the METRIC ET process.

• Students will conduct algorithm development and testing pre-launch using an extensive archive of METRIC-based ET imagery retrieved from Landsat-based processing since 2000.

• Three NSF-supported energy and CO₂ flux systems in Idaho will provide ground data to students for testing the moving window-based calibration against challenging targets of energy balance and ET retrieval from sagebrush and lodgepole pine.

• Post launch testing of algorithms and calibration strategies will be shared with students, as will extension of early ET results with real-time water resources processes in Idaho, California and Nevada, where state Departments of Water Resources will be invited to explore ingestion of ET data into their water operations.
Island Park Lodgepole Pine

- Installed 2010

South Tower Looking North towards North Tower
6 km SW of Macks Inn
Hollister Sage Brush site  
– Installed Feb. 2010

Equipment List:

<table>
<thead>
<tr>
<th>Quantity</th>
<th>Equipment Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>3</td>
<td>3-D sonic anemometers</td>
</tr>
<tr>
<td>1</td>
<td>LiCor 7500 H2O/CO2 Infrared A.</td>
</tr>
<tr>
<td>3</td>
<td>Net radiometers</td>
</tr>
<tr>
<td>1</td>
<td>Scintec BLS900 Scintillometer</td>
</tr>
<tr>
<td>16</td>
<td>Soil Heat Flux Sensors</td>
</tr>
<tr>
<td>32</td>
<td>Soil Temperature Sensors</td>
</tr>
<tr>
<td>20</td>
<td>Soil Water Content Sensors</td>
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<tr>
<td>7</td>
<td>Soil Water Potential Sensors</td>
</tr>
<tr>
<td>2</td>
<td>Rain Gages</td>
</tr>
<tr>
<td>2</td>
<td>Infrared Temperature Sensors</td>
</tr>
</tbody>
</table>

“Sensor Redundancy”
Nighttime Spikes are numerical artifacts

LE from LAS as residual agrees well with LE from Eddy Covariance
Student-based accuracy assessments

Integrated METRIC Estimations VS. Ground RMY Data

- Soil Heat Flux (G)
- Sensible Heat Flux (H)
- Latent Evaporation (LE)
- Net Radiation (Rn)

Hollister Sagebrush site, 2010
Thank you
Extra Slides
Why use Inverse Modeling?

To reduce the impacts of biases in the energy balance components: $R_n$ and $G$
Sensible Heat Flux (H) – by “Classical” methods (non-CIMEC)

\[ H = \rho \ c_p \ \frac{(T_{aero} - T_{air})}{r_{ah}} \]

Challenge (BIAS):

*Up to 2 K different from \( T_{rad} \) (satellite)*

\( T_{aero} \) = aerodynamic temperature

\( r_{ah} \) = the aerodynamic resistance

\[ r_{ah} = \frac{\ln \left( \frac{z_2 - d}{z_{oh}} \right) - \Psi_h(z_2) + \Psi_h(z_1)}{u_* k} \]

\( u_* \) = friction velocity

\( k \) = von karman constant (0.41)
Sensible Heat Flux (H) – CIMEC models

\[
H = \frac{\rho \times c_p \times dT}{r_{ah}}
\]

Advantage: 
\(dT\) is inverse calibrated (simulated) (free of \(T_{rad}\) vs. \(T_{aero}\) and free of \(T_{air}\))

\(dT\) = “floating” near surface temperature difference (K)

\(r_{ah}\) = the aerodynamic resistance from \(z_1\) to \(z_2\)

\[
r_{ah} = \frac{\ln\left(\frac{z_2}{z_1}\right) - \Psi_h(z_2) + \Psi_h(z_1)}{u_\times k}
\]

\(u_\times\) = friction velocity

\(k\) = von karman constant (0.41)
Calibration of *SEBAL* and *METRIC*

CIMEC models:

\[
dT_{\text{cold}} = \frac{H_{\text{cold}} \cdot r_{\text{ah cold}}}{\rho_{\text{air cold}} \cdot c_p}
\]

\[
dT_{\text{hot}} = \frac{H_{\text{hot}} \cdot r_{\text{ah hot}}}{\rho_{\text{air hot}} \cdot c_p}
\]

\[
R_n - G - 1.05 ET_{\text{ref alfalfa}} \quad \text{(METRIC)}
\]

or 0 \quad \text{(SEBAL – classical)}
Calibration of SEBAL and METRIC CIMEC's:

Derivation of linear $dT$ vs. $T_s$ function

\[
a = \frac{dT_{\text{hot}} - dT_{\text{cold}}}{T_s \text{ hot} - T_s \text{ cold}}
\]

\[
b = dT_{\text{hot}} - a T_s \text{ hot}
\]

and \[dT = a + b T_s\] at all pixels

Regardless of ‘1-source’ or ‘2-source’ model:
‘the dry condition’ (bare, dry field) is a ‘1-source’ condition.

Regardless of ‘1-source’ or ‘2-source’ model:
‘the wet condition’ (fully veg. field) is a ‘1-source’ condition.
Daily water balance using gridded weather and precipitation