



Royal Netherlands
Meteorological Institute
Ministry of Infrastructure and the
Environment



EU SURFACE TEMPERATURE FOR ALL CORNERS OF EARTH (EUSTACE)

Nick Rayner, Met Office Hadley Centre, Science Coordinator EUSTACE project,
and partners

Material provided by Renate Auchmann, Yuri Brugnara, Colin Morice, Darren
Ghent, Lizzie Good, John Kennedy, Kristine Madsen, Iestyn Woollway, Finn
Lindgren and Janette Bessembinder

MARCDAT-IV, 18th-22nd July 2016



*EUSTACE has received funding from the European Union's Horizon 2020 Programme for
Research and Innovation, under Grant Agreement no 640171*



EUSTACE AIMS

EUSTACE will give publicly available daily estimates of surface air temperature since 1850 across the globe for the first time by combining surface and satellite data using novel statistical techniques. To do this, we need to:

- Identify non-climatic discontinuities in daily weather station data, *so users can trust the changes our records show*
- Produce consistent uncertainty estimates for satellite skin temperature retrievals over all surfaces (land, ocean, ice and lakes), *so we know how far to trust the estimates everywhere*
- Understand how surface temperature measured *in situ* and by satellite relates, *to estimate air from skin temperature*
- Estimate values in areas where we have no *in situ* or satellite data, *so users can have daily information here*





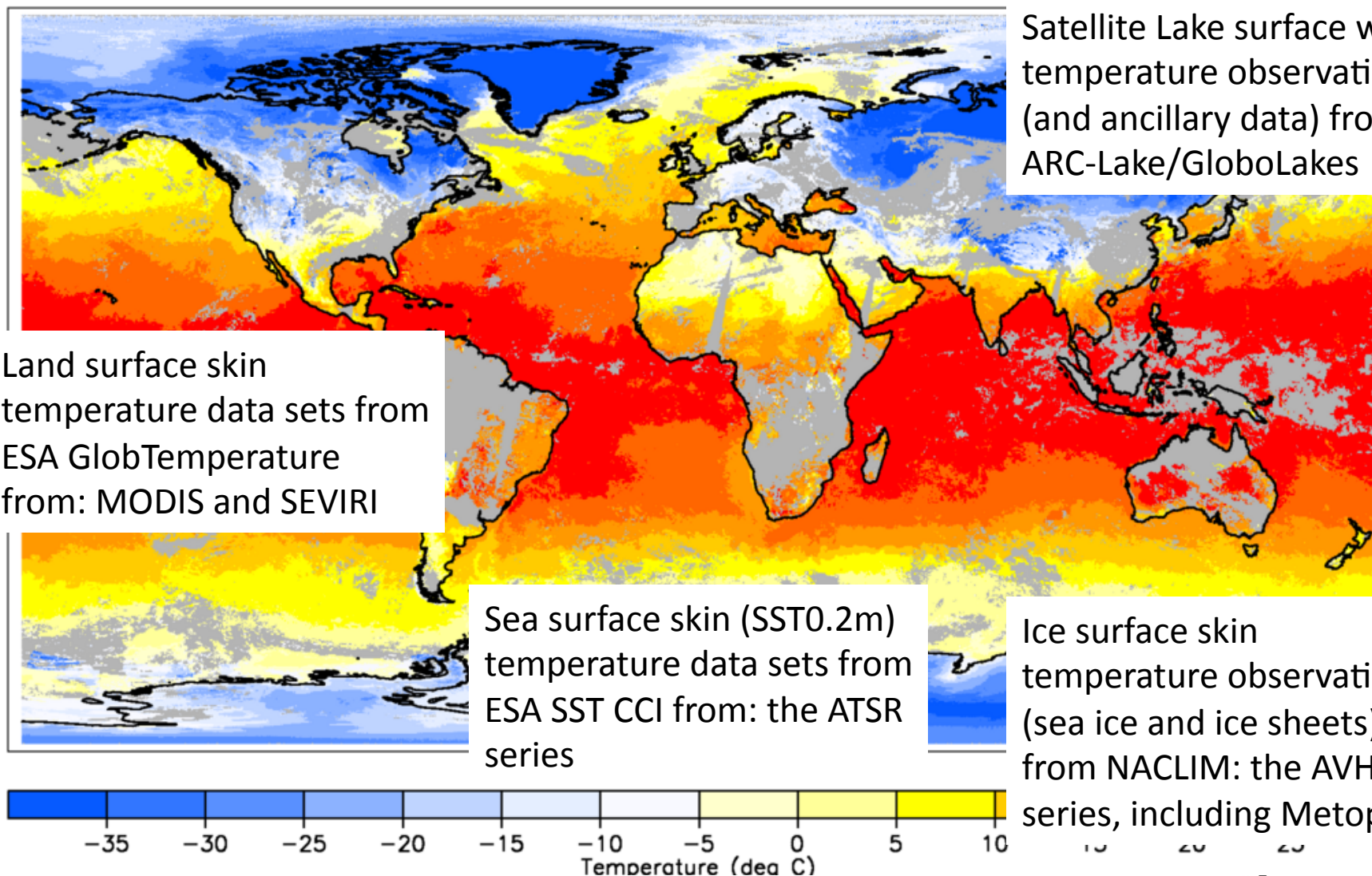
DATA SOURCES



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



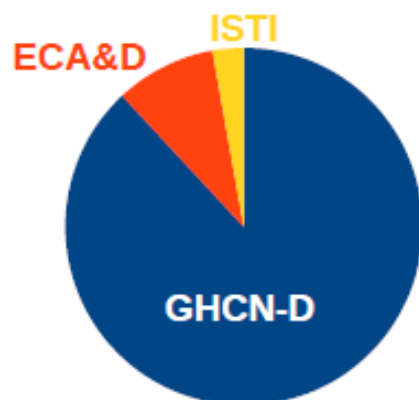
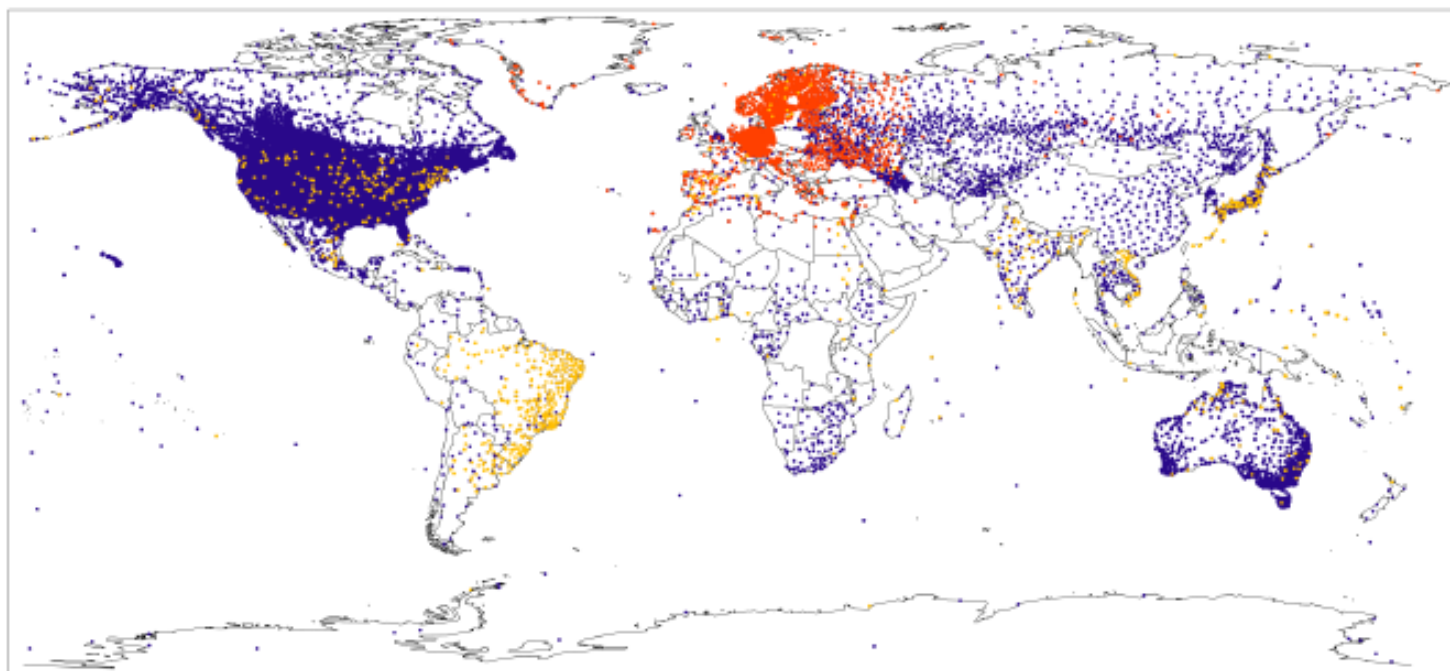
SATELLITE OBSERVATIONS (TSKIN)



Satellite Lake surface water temperature observations (and ancillary data) from ARC-Lake/GloboLakes

Data Sources

TEST DATASET



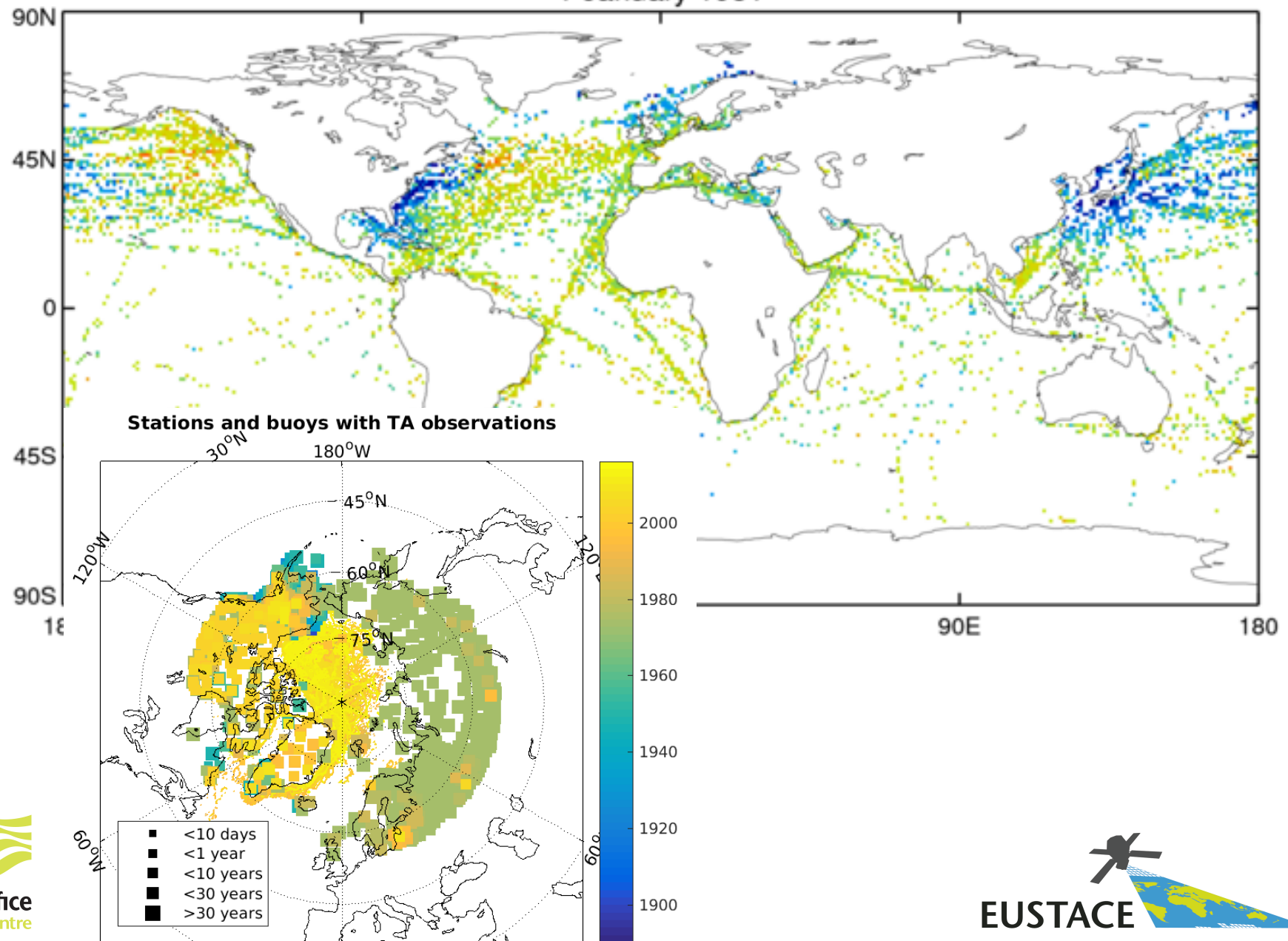
33'511 stations

(32'041 Tmax + 32'265 Tmin)

(Data are stored on EUSTACE workspace
in a common ASCII and NetCDF format)

Ship observations

1 January 1981





IDENTIFYING NON-CLIMATIC BREAKS IN DAILY LAND STATION SERIES



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



DETECTING DISCONTINUITIES IN DAILY LAND STATION DATA

- s
- r

Para
Steve

Diffe
obse
varia
in c

3 DETECTION METHODS

Caussinus-Mestre

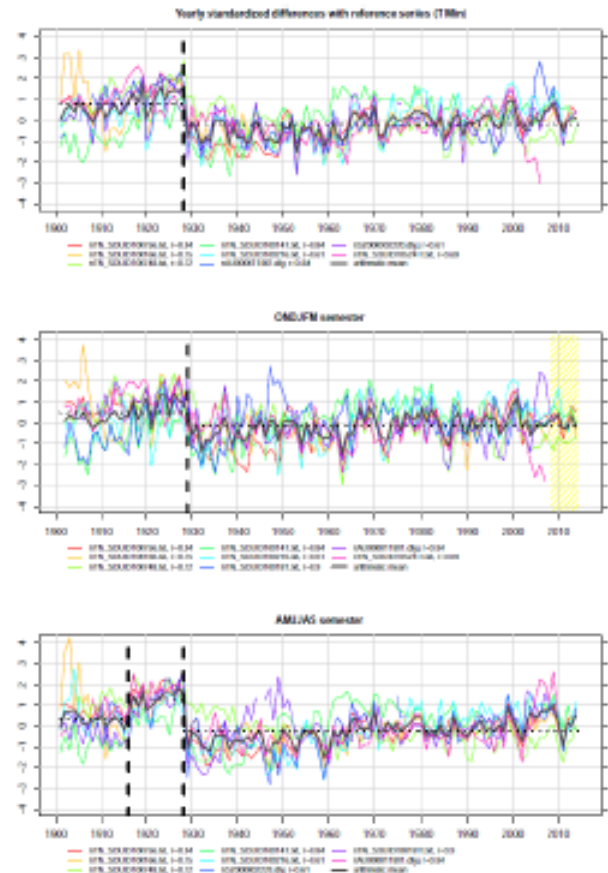
RHtests

GAHMDI

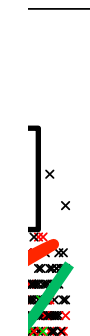
combined:

a discontinuity is
significant if confirmed
by at least

3 reference stations
using at least
2 of the methods



the mean



partly cloudy

clear

Delivered breakpoints Basel:
1916 and 1928

day of year






DEVELOPING CONSISTENT SATELLITE UNCERTAINTY ESTIMATES



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



ESTIMATING CONSISTENT UNCERTAINTIES IN SATELLITE RETRIEVALS

- Uncertainties categorised by effects whose errors have distinct correlation properties:
 - random
 - ~~locally systematic~~ aka “structured random”
 - (large-scale) systematic
- This three-component model applies to all processing levels and LST products
- Propagation of uncertainties:
 - L1  L2  L3  L4 (Merged Product)



UNIVERSITY OF
LEICESTER



**National Centre for
Earth Observation**
NATURAL ENVIRONMENT RESEARCH COUNCIL



LST Uncertainty Components

VARIABLE	METHOD	COMMENTS
LST_UNC_RAN	L2 Random 1 / Radiance noise Propagation	$u_{\downarrow \text{ran}, y}(x) = \sqrt{\sum_{c=1}^n \left(\frac{\partial R}{\partial y_{\downarrow c}} u_{\downarrow \text{ran}}(y_{\downarrow c}) \right)^2}$ <p>Random component of L1 channel uncertainties propagated through the retrieval</p>
	L2 Random 2 / Emissivity noise Propagation	$u_{\downarrow \text{ran}, \varepsilon}(x) = \sqrt{\sum_{c=1}^n \left(\frac{\partial R}{\partial \varepsilon_{\downarrow c}} u_{\downarrow \text{ran}}(\varepsilon_{\downarrow c}) \right)^2}$ <p>Estimate of the magnitude of pixel-to-pixel scale emissivity variability within areas based on land cover class</p>
LST_UNC_LOC	L2 Local 2 / Uncertainty from atmosphere/fit for regression-based retrieval	$u_{\downarrow \text{loc}, \text{fit}}(x) = \sqrt{\text{Var}(x - x_{\downarrow \text{in}})}$ <p>Atmospheric fields correlated on timescales >1 day and length scales >100 km. For coefficient based retrieval methods the retrieval ambiguity is a contributor of residuals in the fit</p>
	L2 Local 2 / Uncertainty from Emissivity	$u_{\downarrow \text{loc}, \varepsilon}(x) = \sqrt{\sum_{c=1}^n \left(\frac{\partial R}{\partial \varepsilon_{\downarrow c}} u_{\downarrow \text{loc}}(\varepsilon_{\downarrow c}) \right)^2}$ <p>Across a particular land class area, there may be a mean difference between the assumed and true mean emissivity</p>



UNIVERSITY OF
LEICESTER



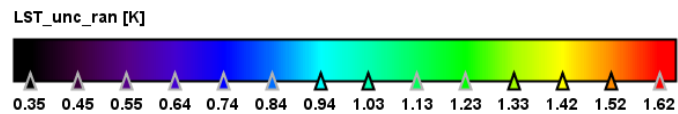
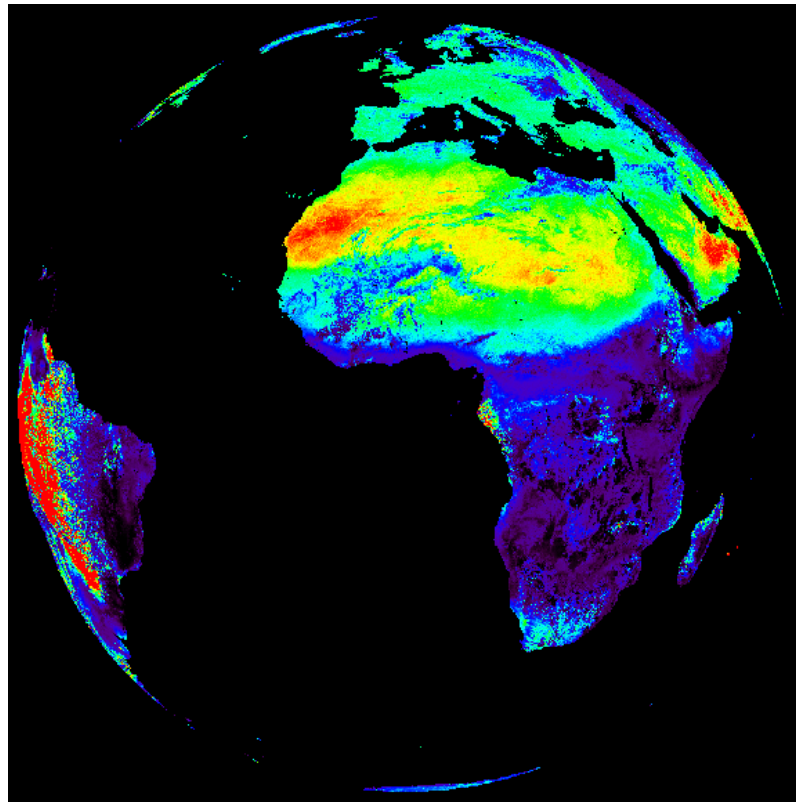
National Centre for
Earth Observation
NATURAL ENVIRONMENT RESEARCH COUNCIL

Corrections have been applied by data priority in the bias of the satellite surface temperature (ie from validation)

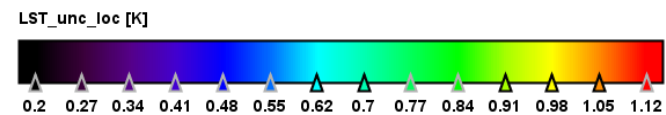
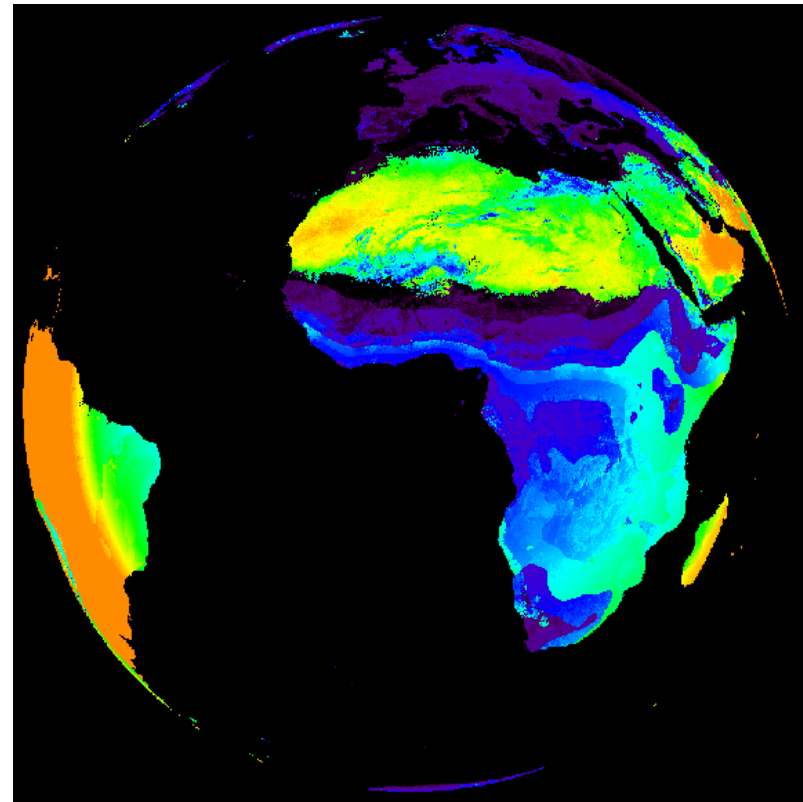


SEVIRI Uncertainties

Random



Locally correlated/structured random



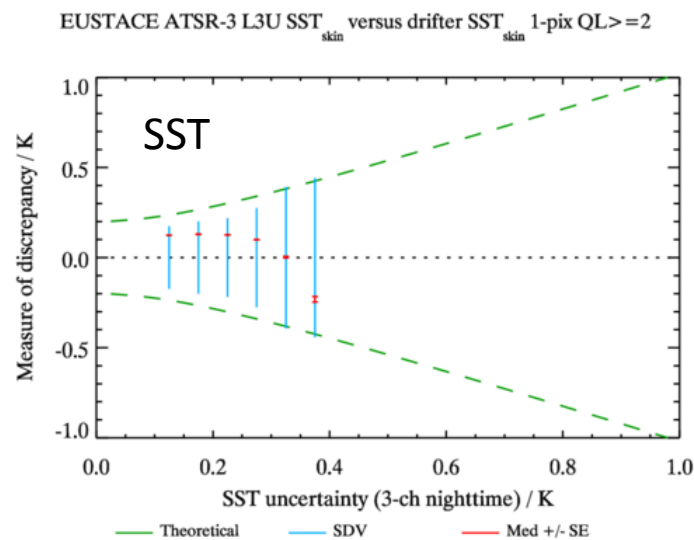
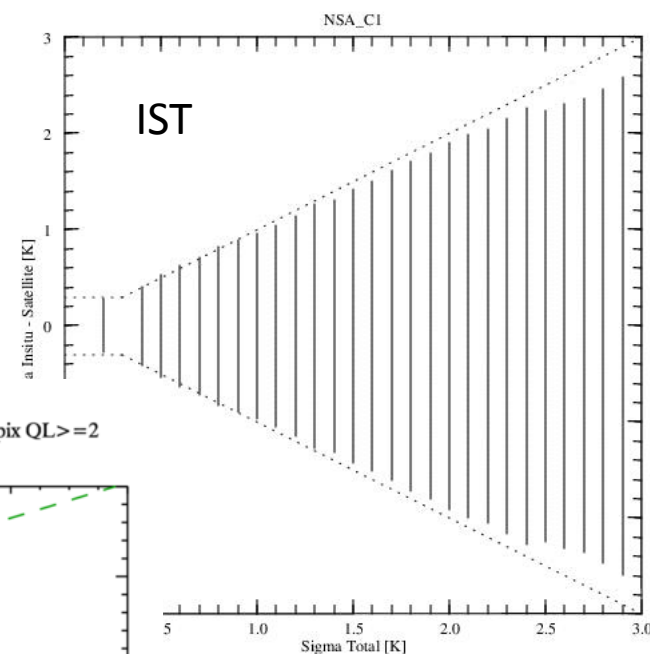
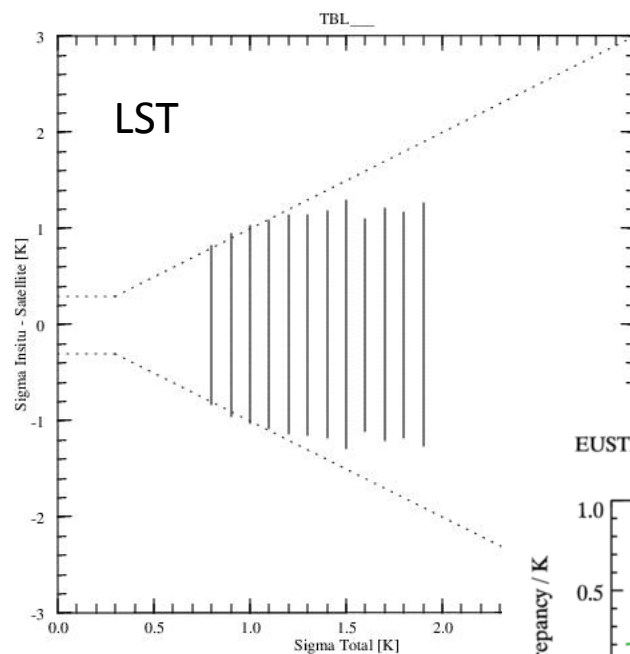
UNIVERSITY OF
LEICESTER



**National Centre for
Earth Observation**
NATURAL ENVIRONMENT RESEARCH COUNCIL



VALIDATION OF LST, IST AND SST TOTAL UNCERTAINTIES AT REFERENCE STATIONS



UNIVERSITY OF
LEICESTER



**National Centre for
Earth Observation**
NATURAL ENVIRONMENT RESEARCH COUNCIL





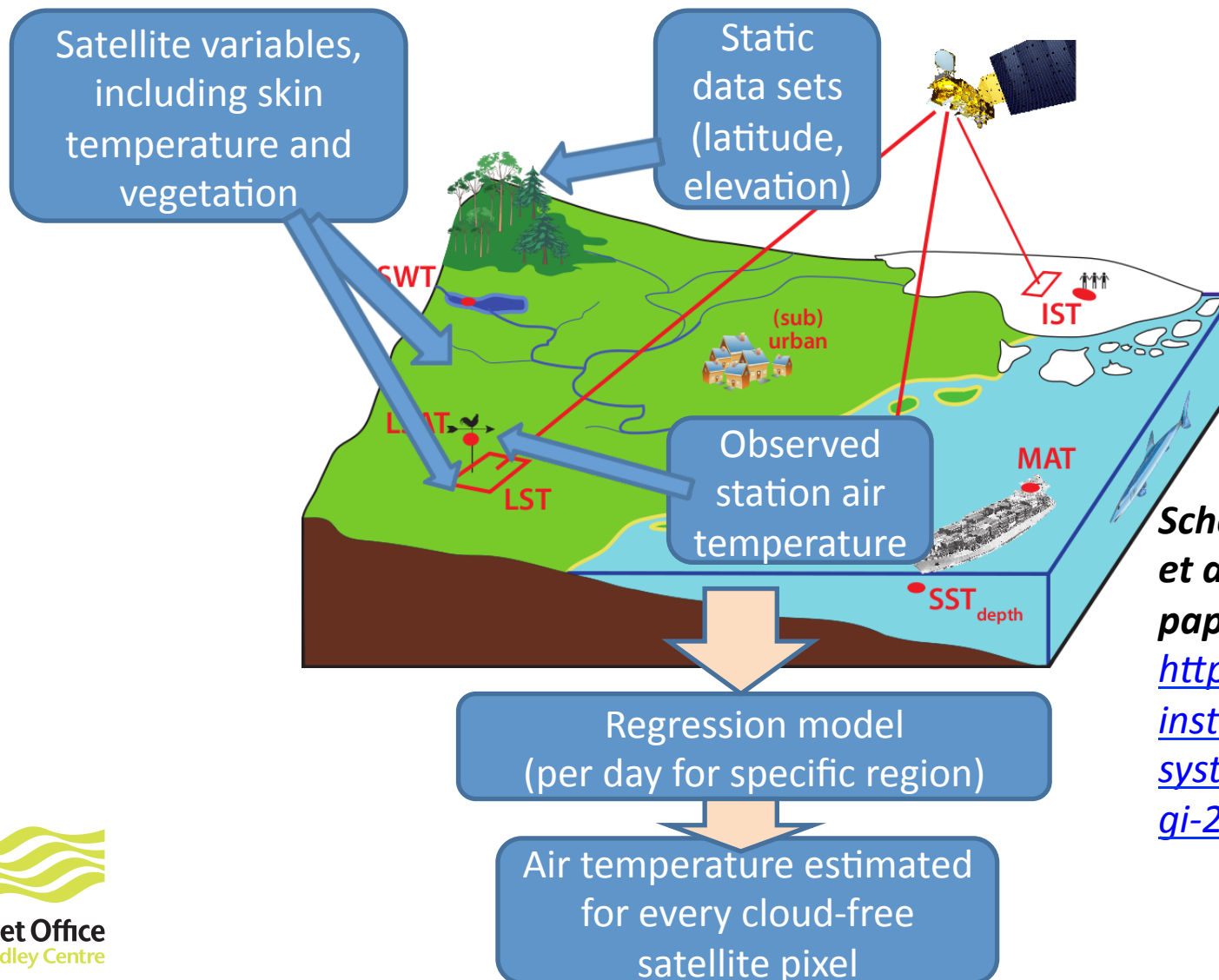
ESTIMATING AIR TEMPERATURE FROM SATELLITE DATA: (1) LAND



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



ESTIMATING AIR TEMPERATURE FROM SKIN TEMPERATURE

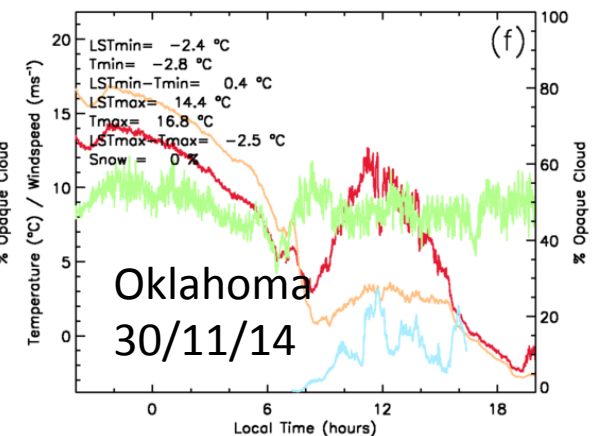
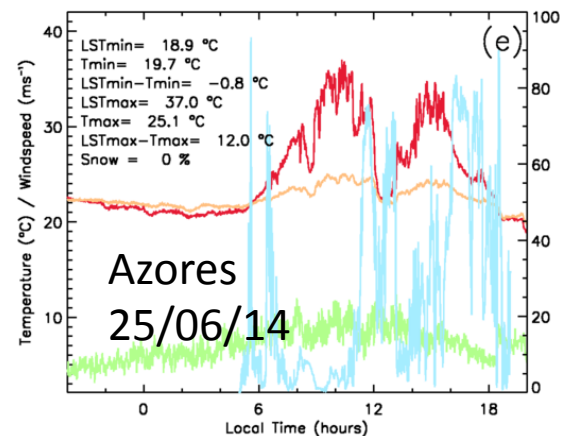
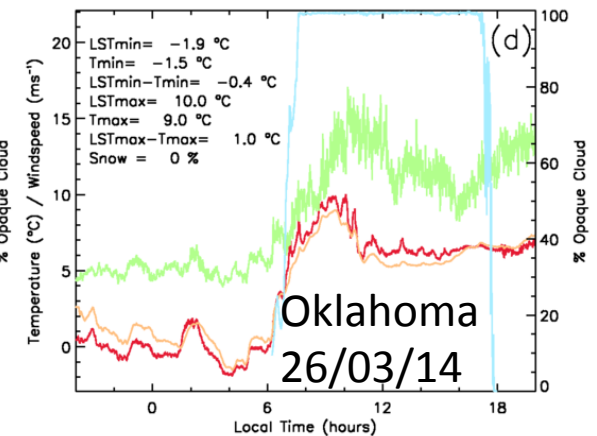
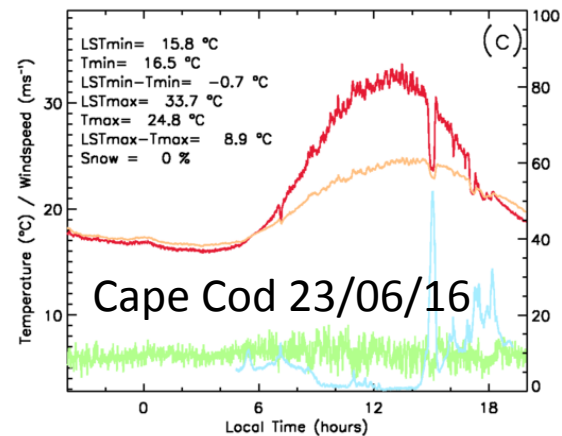
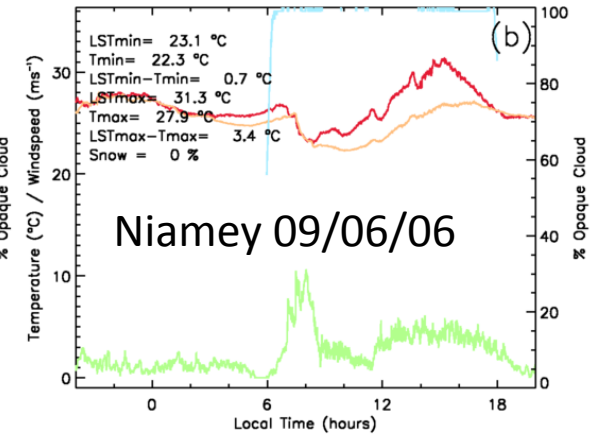
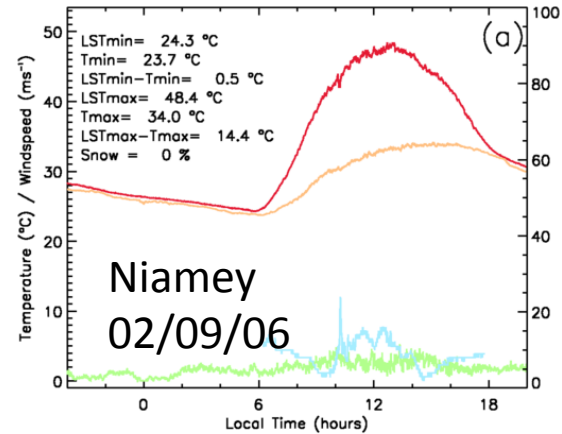


Schematic from Merchant et al., 2013 community paper and roadmap:

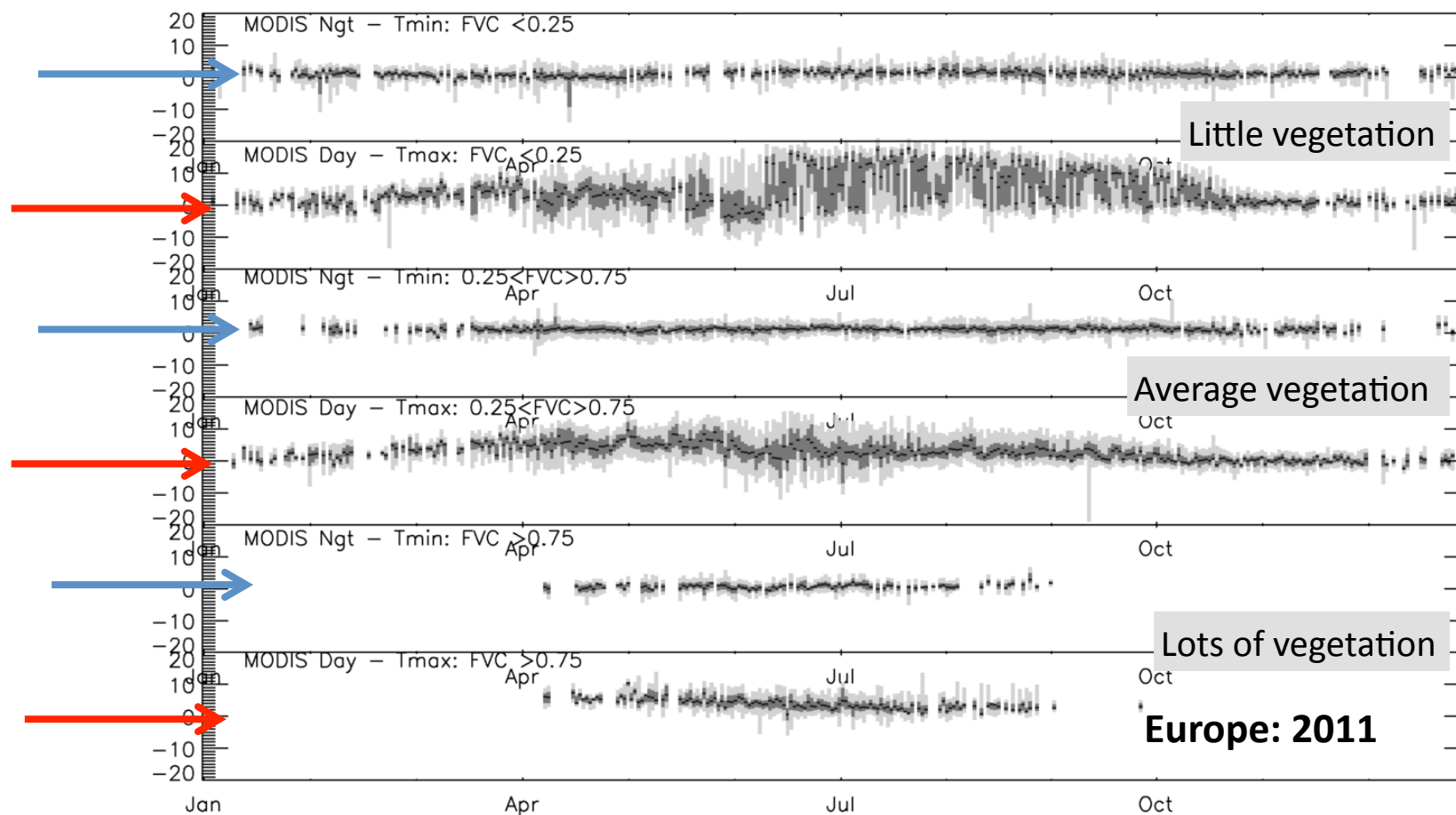
<http://www.geosci-instrum-method-data-syst.net/2/305/2013/qi-2-305-2013.html>

VARIABILITY ANALYSIS

- Paper in press in JGR-Atm
 - *'An in situ-based analysis of the relationship between skin and screen-level temperatures'*
- 19 ARM deployments, analysing up to 1 year of data from each.
- Wind speed and cloud are important factors governing instantaneous T_{skin}-T_{air} differences. Influence less clear on T_{skin}-T_{air} extremes.



RELATIONSHIP WITH VEGETATION

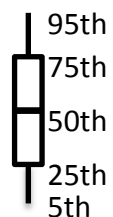


Europe: 2011

Minimum temperature
difference: little impact by
changing vegetation filter

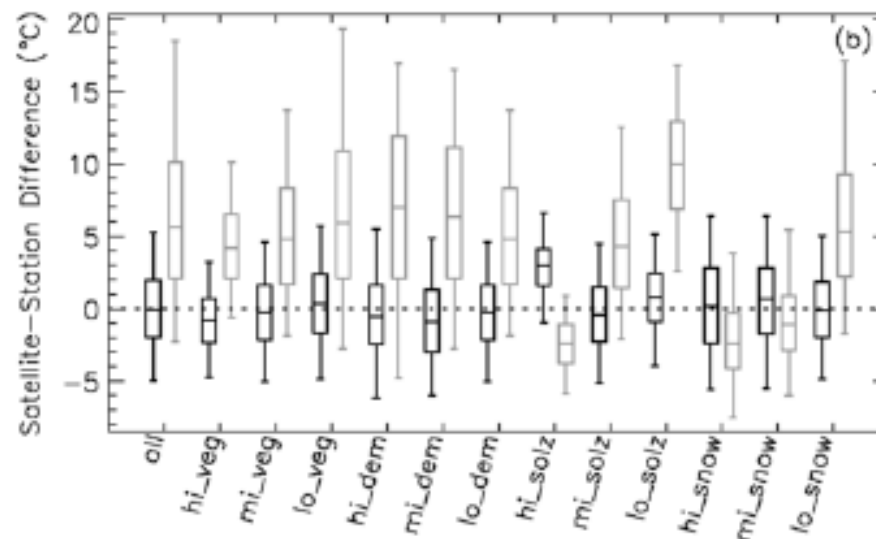
Maximum temperature
difference: more vegetation
= less extreme differences

SURFACE AIR TEMPERATURE ESTIMATES

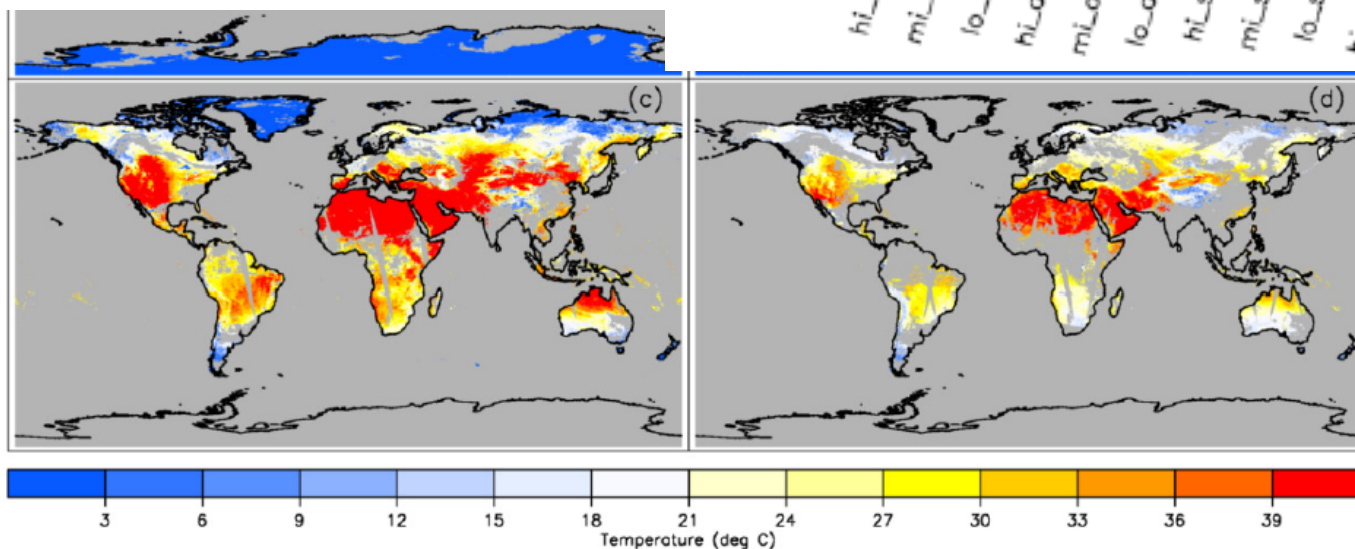


Independent evaluation
2002-2012 with 5000 GHCN-D
stations not used for training

LST Nigl Grey indicates results for LST
Black indicates results for satellite
T2m



LST Day



Tmax



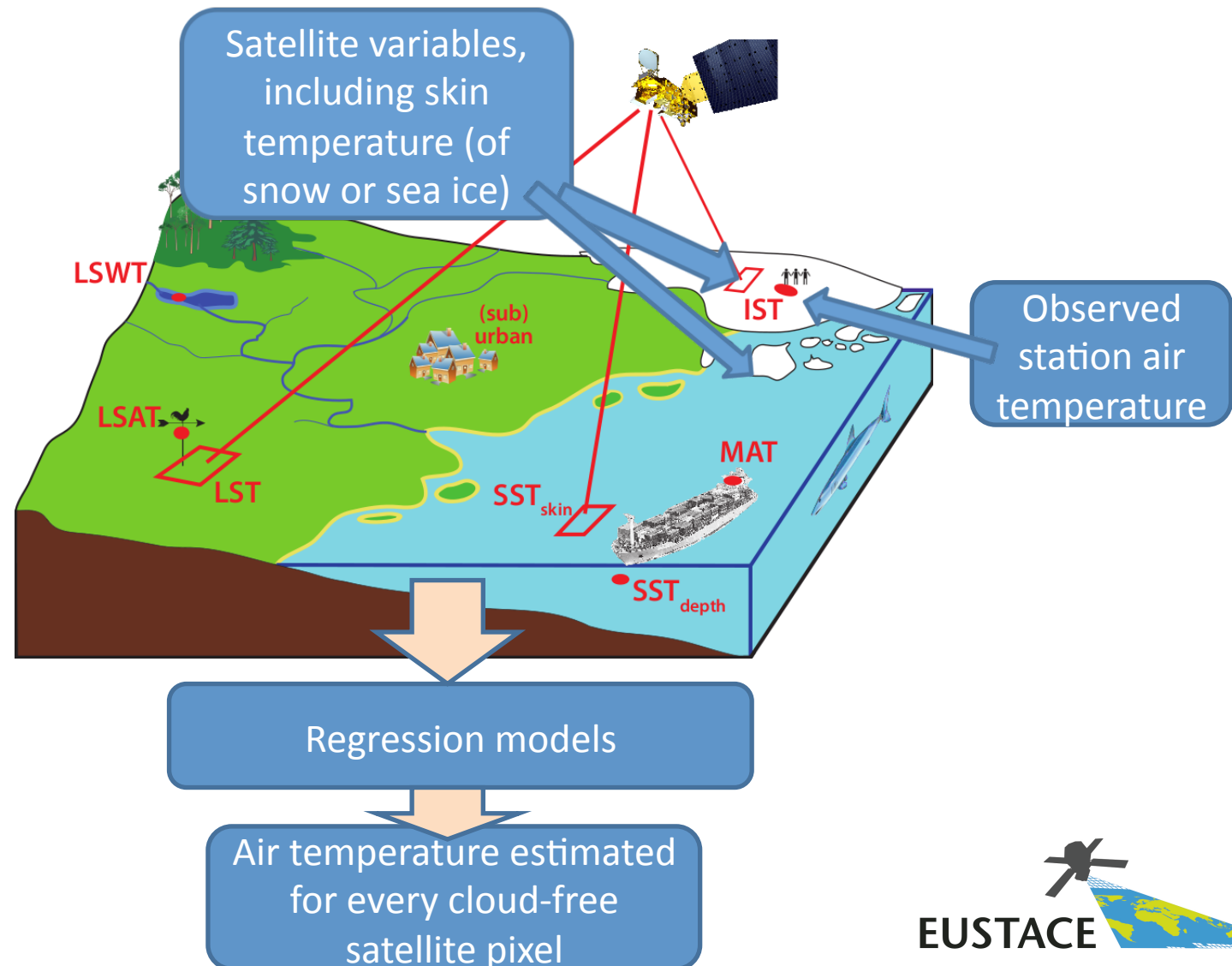
ESTIMATING AIR TEMPERATURE FROM SATELLITE DATA: (2) ICE



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171

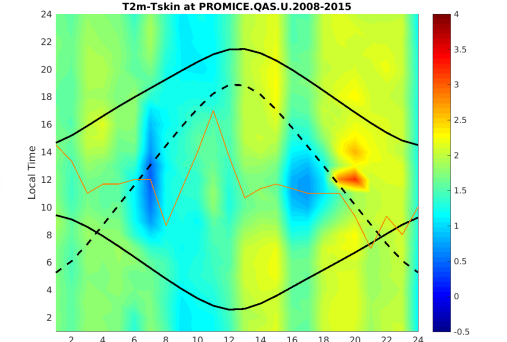
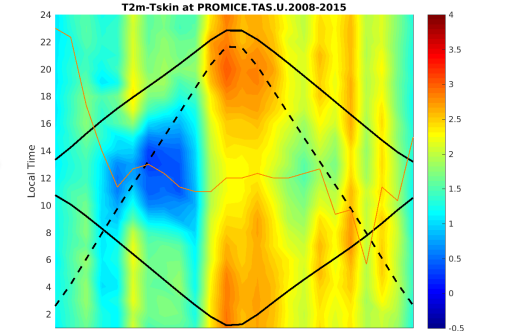
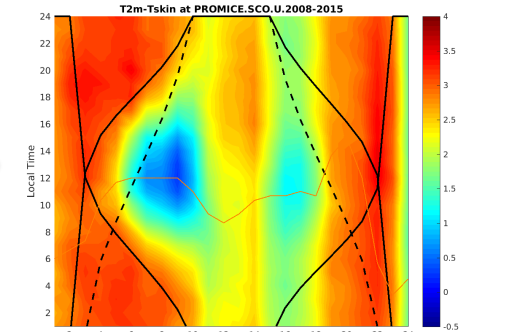
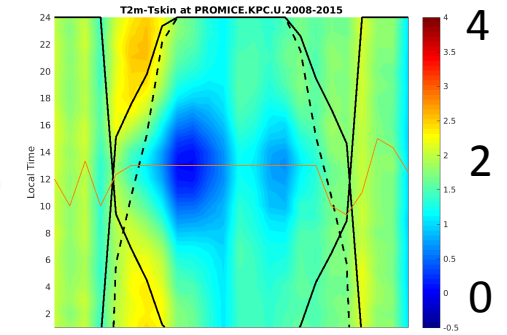
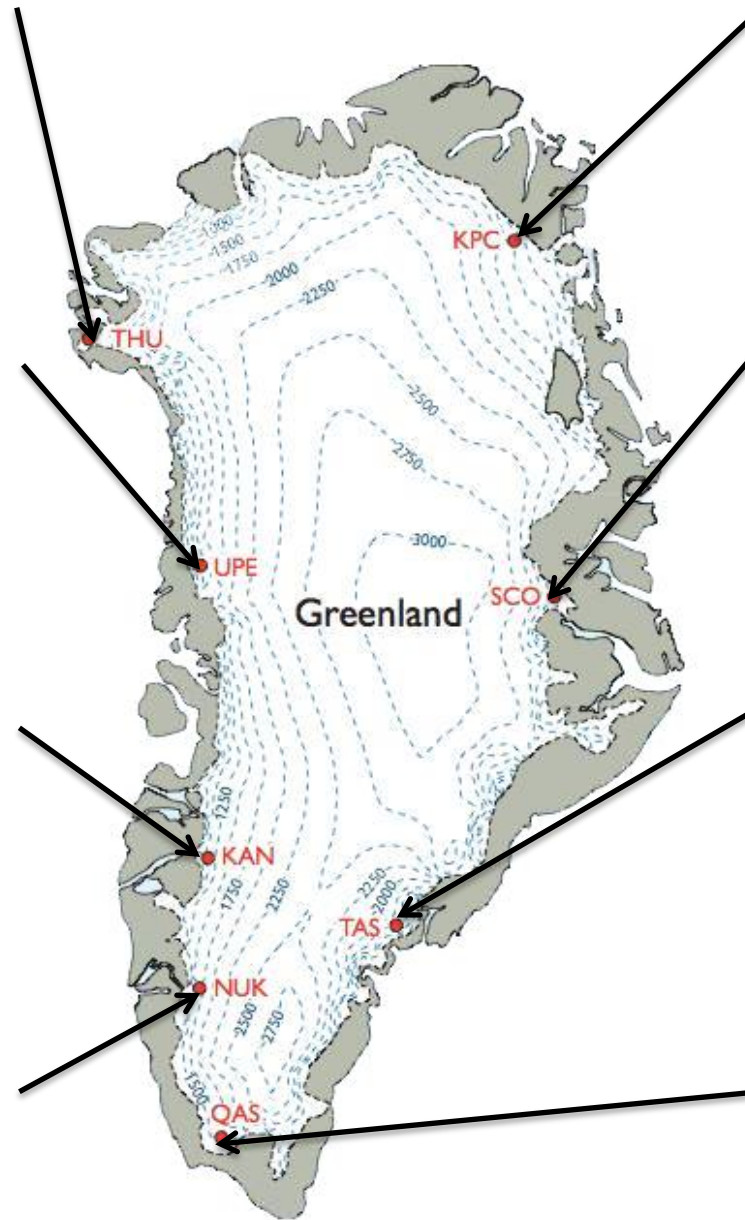
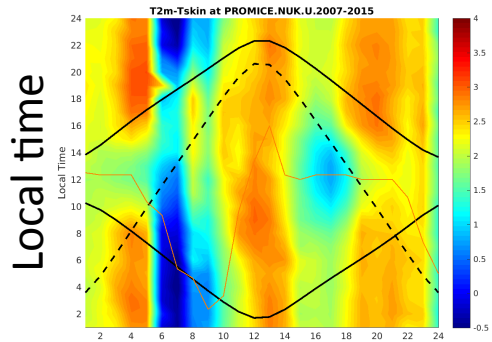
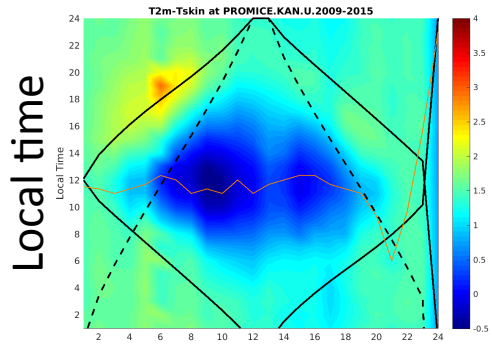
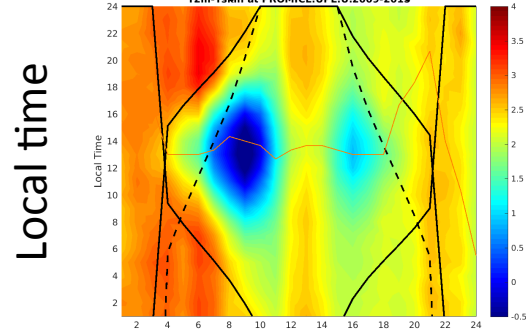
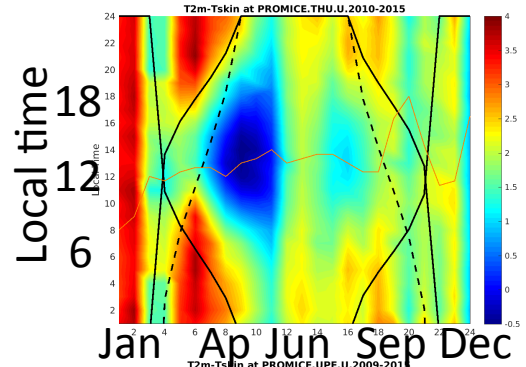


ESTIMATING AIR TEMPERATURE FROM SKIN TEMPERATURE – OTHER SURFACES



PROMICE

Programme for Monitoring of the Greenland Ice Sheet



Tair-Tskin

Tair-Tskin

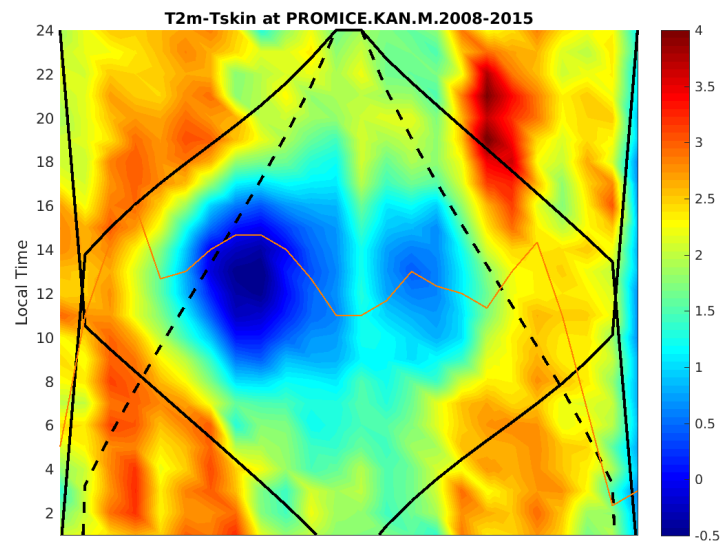
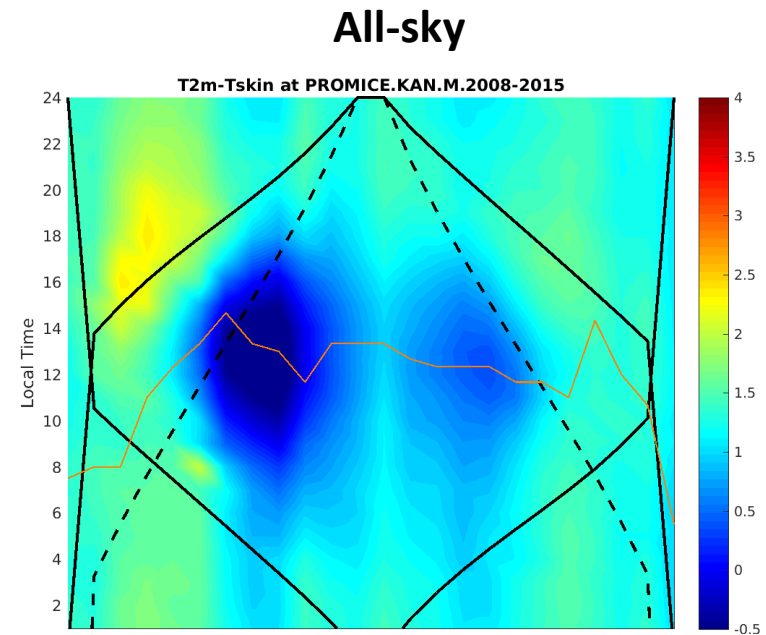
Tair-Tskin

Tair-Tskin

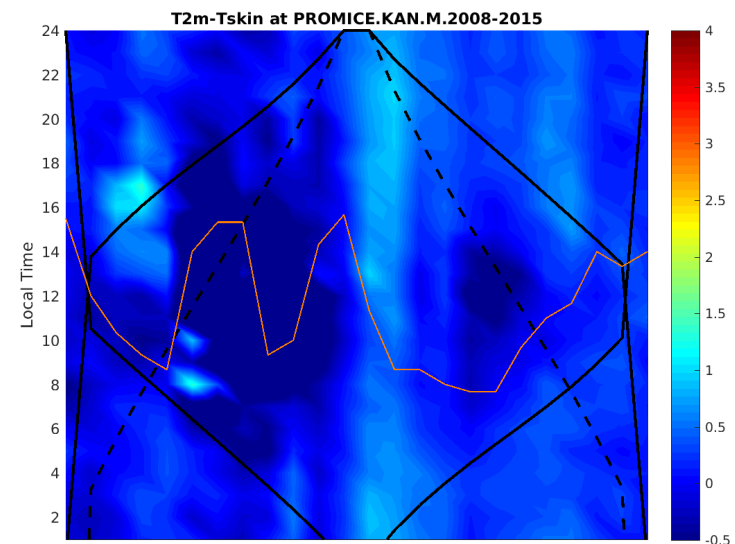
IMPACT OF CLOUDS

PROMICE - KAN M

Surface and air temperature differences at KAN_M in cases of all-sky, clear-sky, and overcast conditions.

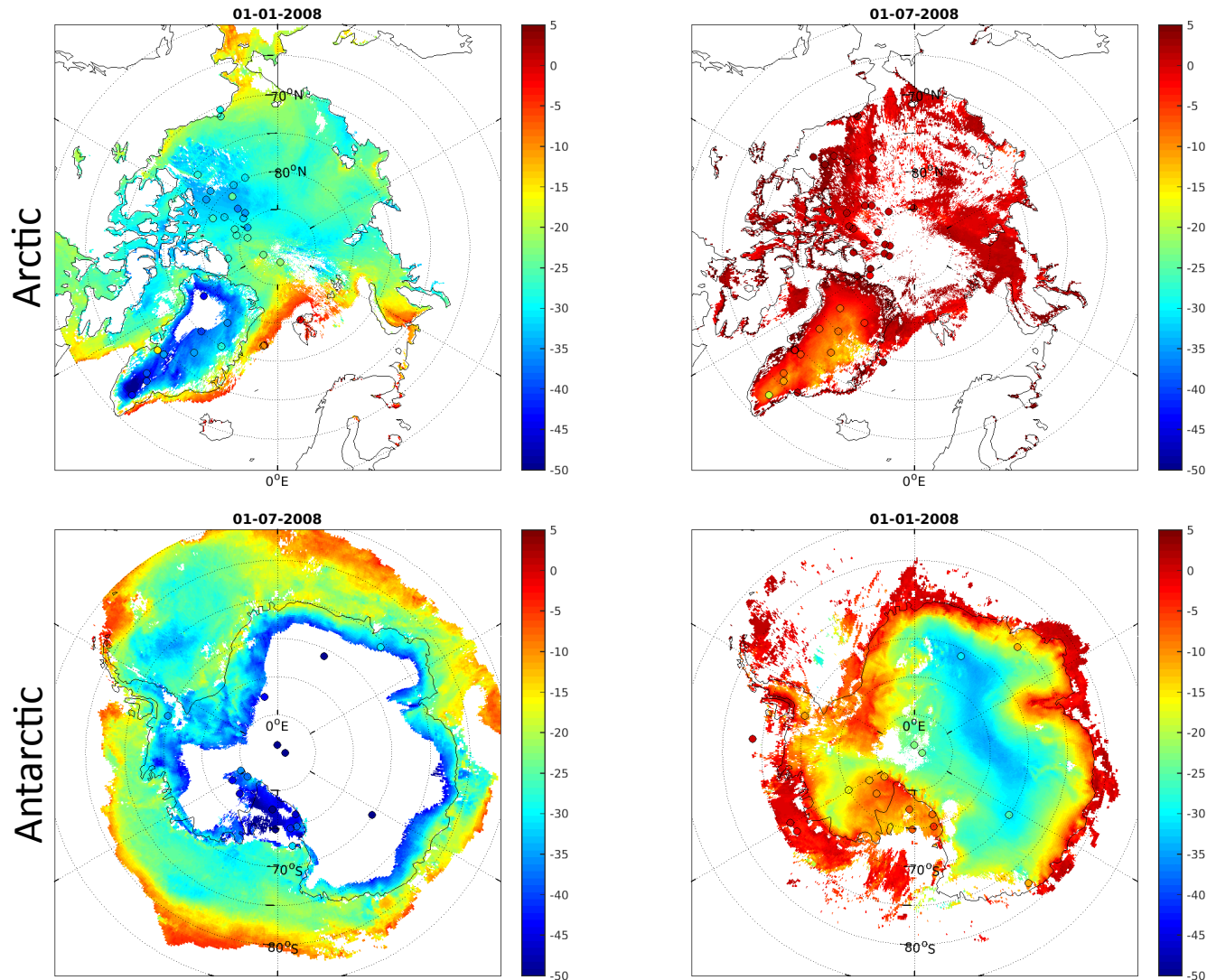


Clear sky (< 0.1)



Overcast (>0.9)

ICE SURFACE AIR TEMPERATURES



Four examples of daily mean surface air temperature over ice (test version), estimated from satellite IST. Circles show in-situ measurements and white areas indicate missing data, often due to clouds.



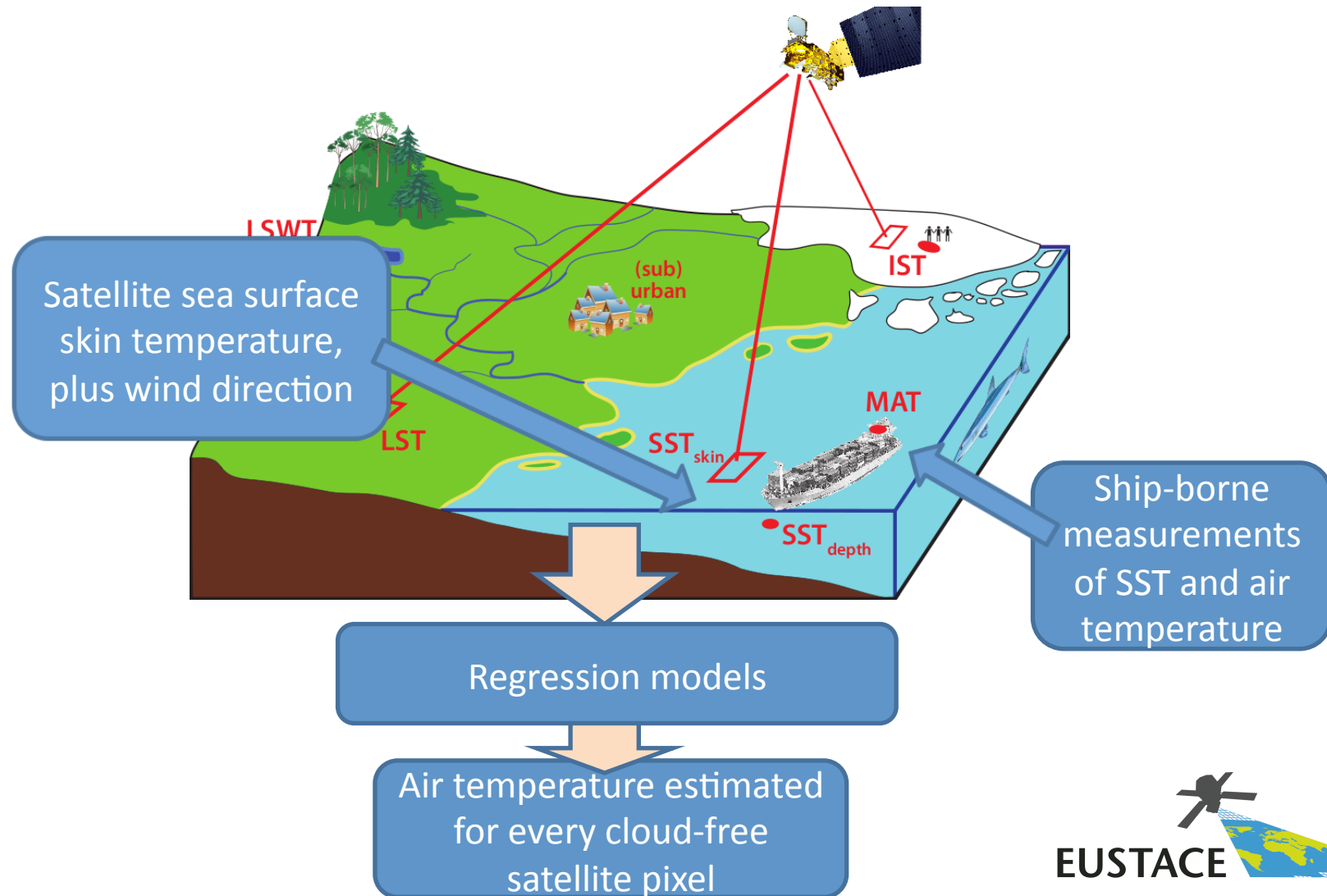
ESTIMATING AIR TEMPERATURE FROM SATELLITE DATA: (3) OCEAN



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



ESTIMATING AIR TEMPERATURE FROM SKIN TEMPERATURE – OTHER SURFACES



UNDERSTANDING RELATIONSHIPS BETWEEN SST AND MAT

- Uses ICOADS 2.5 data, 1961-2000 and UIDs
- New quality control applied to SST and MAT, together with bias corrections
- Extract reports with SST and NMAT
- Night is defined as one hour after sundown to one hour after sunrise
- Estimated uncertainties in individual SST and MAT measurements is about 1°C.

MODEL

$$\text{MAT} = \text{SST} + \Delta + \varepsilon$$

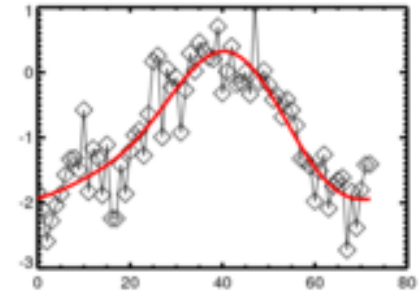
Sea-surface temperature

SST from CCI ATSR
Possibly AVHRR also

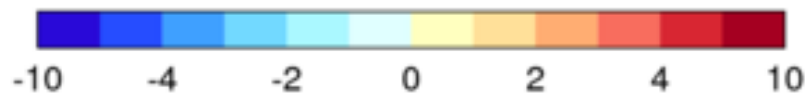
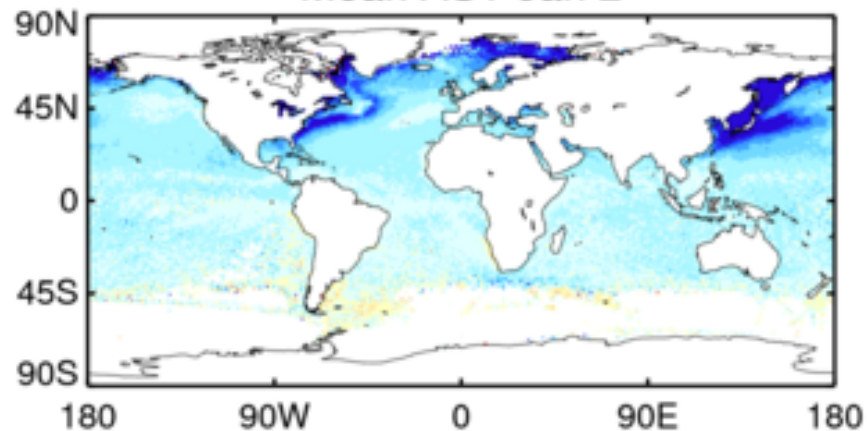
Climatological average air-sea temperature difference

Time-varying locally-correlated "error"

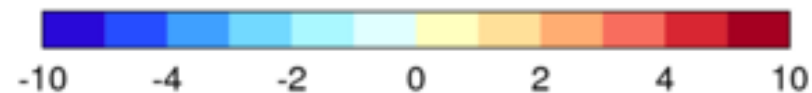
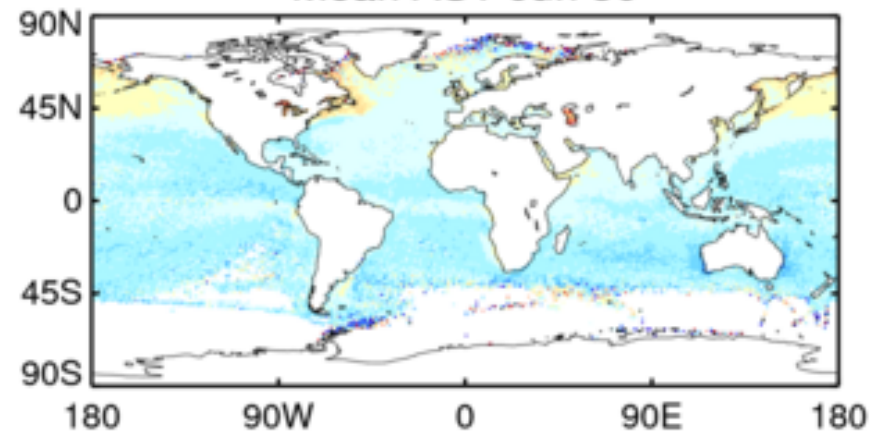
1. Simple average
2. Fixed seasonal cycle
3. Spatially varying length and time scales
4. Incorporating ship observations



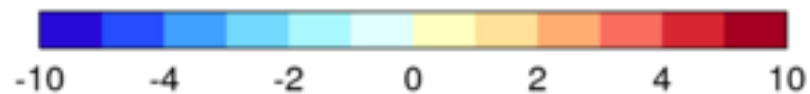
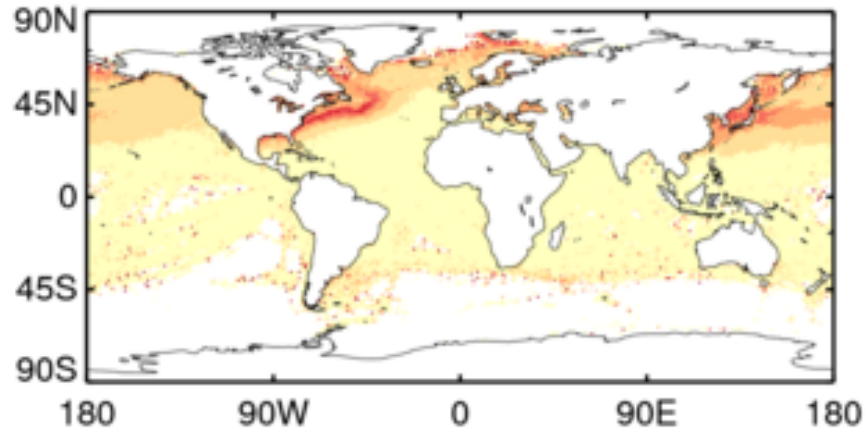
Mean AST Jan 2



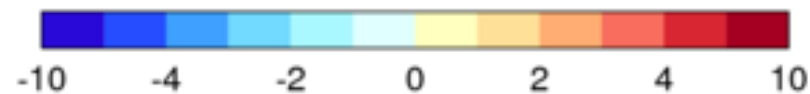
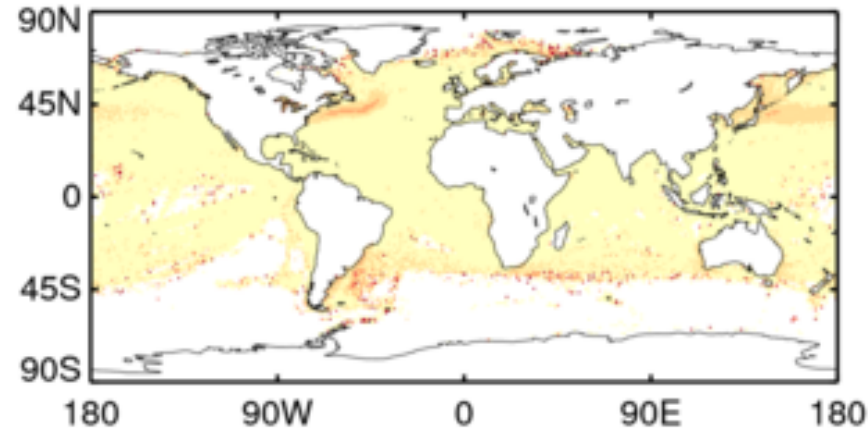
Mean AST Jun 30



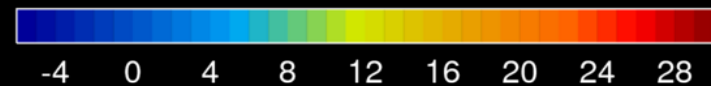
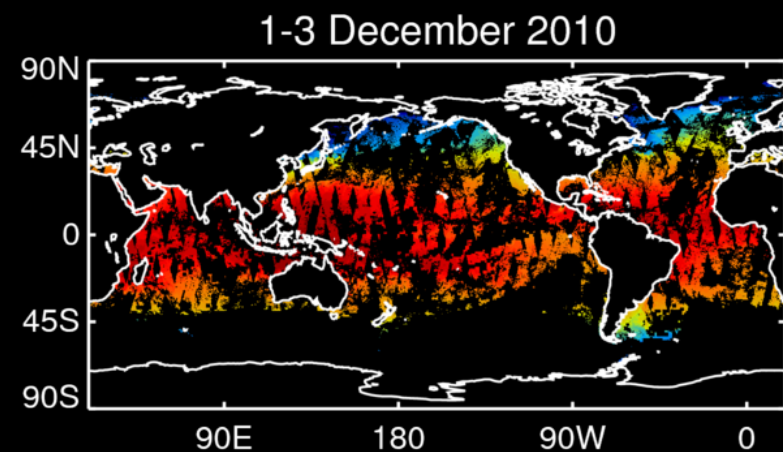
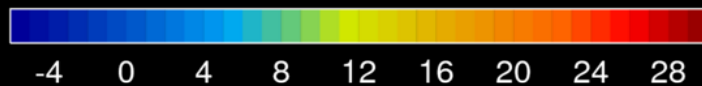
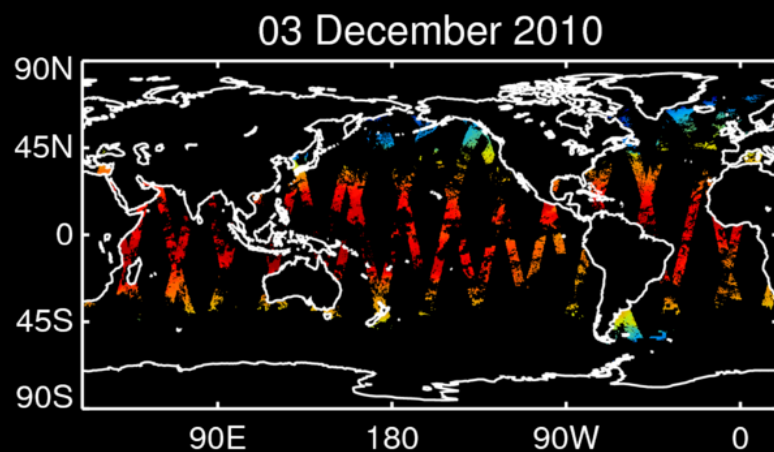
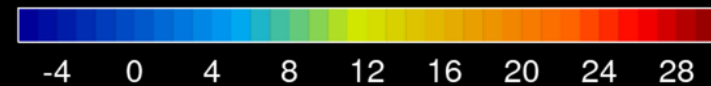
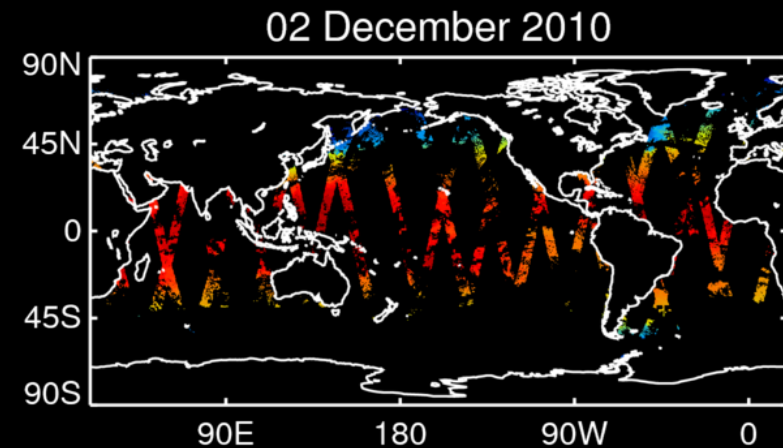
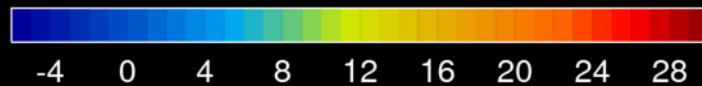
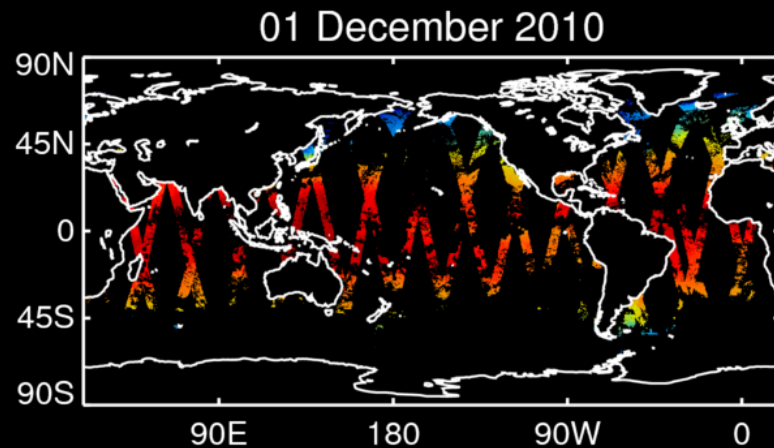
Standard Deviation AST Jan 2



Standard Deviation AST Jun 30



Estimated air temperatures over oceans using ATSR series SST retrievals



ESTIMATING UNCERTAINTY IN MAT DERIVED FROM SST

- Several components
- Uncertainty component arising from deviations from the mean air-sea temperature difference, which is represented by the standard deviation. Locally correlated with length scale of around 2000 km
- Uncertainty in the fourier components, which is correlated in time, but uncorrelated in space.
- Uncertainty in the SST, which depends on the SST source. For ATSR a combination of random, locally correlated, large scale correlated.



ESTIMATING AIR TEMPERATURE FROM SATELLITE DATA: (4) LAKES



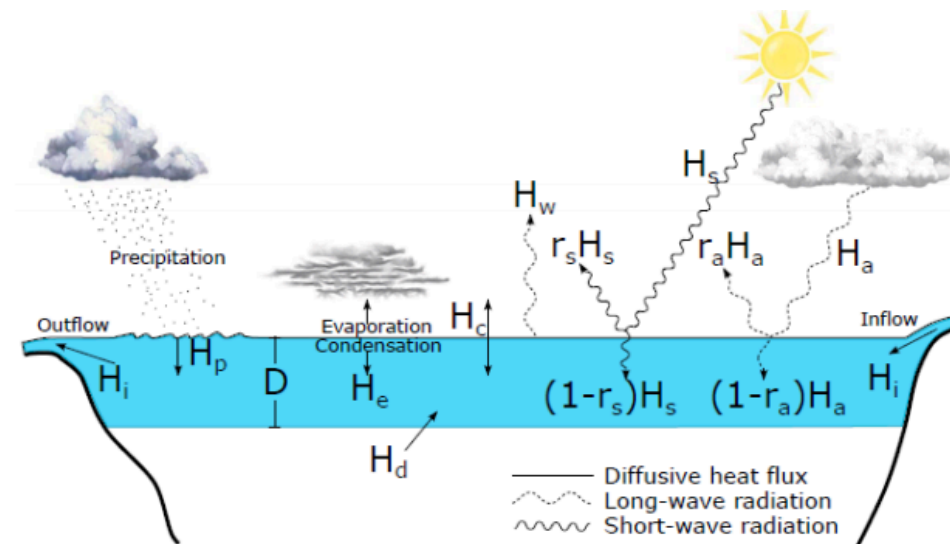
EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



RELATIONSHIP BETWEEN SAT AND LSWT

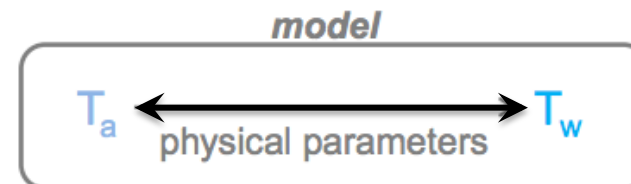
Heat budget in the well-mixed surface layer

$$\rho c_p V \frac{dT_w}{dt} = H_{\text{net}} A$$



Main forcing factor: air temperature T_a

Main result: surface water temperature T_w



SUMMARY OF APPROACH

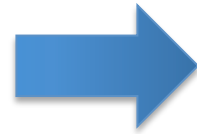
Reanalysis data



Flake model



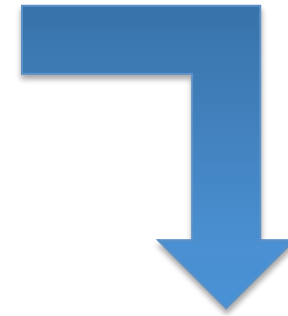
Simulated mixed
depth and
temperature
series



Validate series
with observed
LSWT



Relationship building –
determined for times
where both SAT and LSWT
are available



Compute
smoothed air
temperature
series



ESTIMATE VALUES IN AREAS WHERE WE HAVE NO *IN SITU* OR SATELLITE DATA ON A PARTICULAR DAY



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171

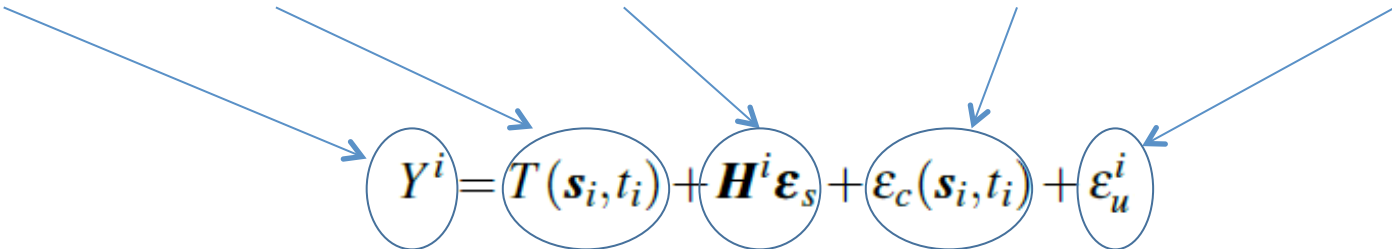


THE EUSTACE ANALYSES

- Should merge data sources, accounting for observational uncertainties to produce spatially- and temporally-complete temperature estimates, with quantified uncertainties.
- Two methods, *Advanced Standard* and *Ambitious*, are being developed in parallel:
 - *Advanced Standard*: links to existing methods for observational climate data sets.
 - *Ambitious*: more experimental.
- Both approaches require novel techniques to work with daily data at EUSTACE resolutions.

MEASUREMENT MODEL (TMEAN)

Observation_i = Temperature + Systematic error + Locally correlated error + Uncorrelated error


$$Y^i = T(s_i, t_i) + H^i \epsilon_s + \epsilon_c(s_i, t_i) + \epsilon_u^i$$

Temperature – at observation location. See next slide.

Systematic error – error terms with spatial or temporal structure described by H^i (e.g. station homogenisation, covariate in calibration model).

Locally correlated – Error terms with correlation structure in space or time (e.g. atmospheric correction errors for satellite observations).

Uncorrelated error – Error associated only with observation i .

THE ADVANCED STANDARD MODEL (TMEAN)

Temperature = Climatology + Low rank decomposition + Local process

$$T(s,t) = C(s,t) + w^T(s)x(t) + Z(s,t)$$

Standard approach is to estimate each component in turn, ignoring components to the right.

Climatology – Covariates (e.g. altitude, latitude) + Seasonal component.

Low rank – Principal patterns of variability.

Local process – “weather”

LOW RANK COMPONENT

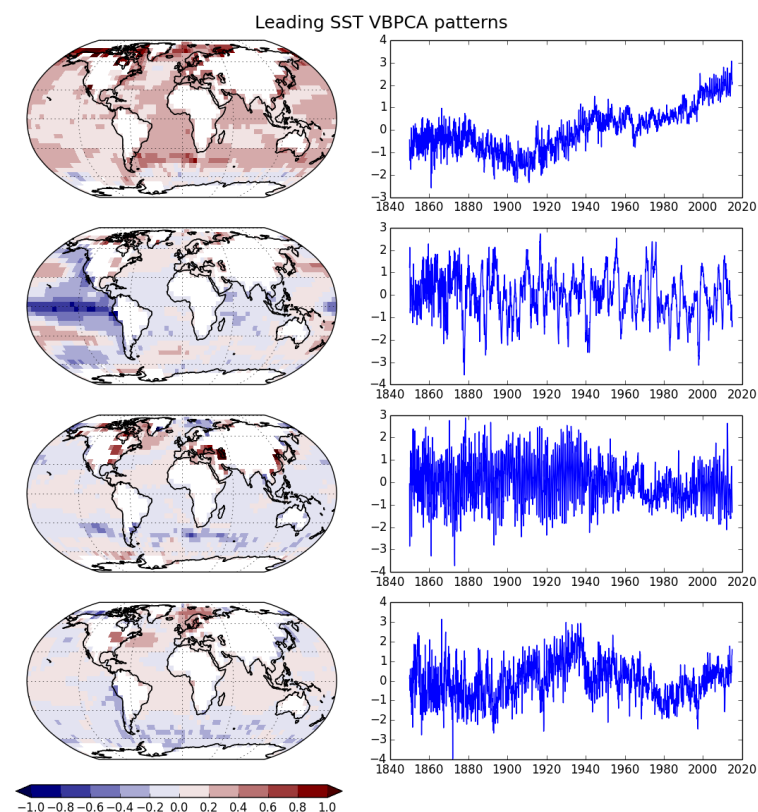
Decomposition into a small set of spatial patterns and a corresponding time series of pattern weights.

Reconstruct large scale variability as time t through a weighted sum of spatial patterns.

Long-term changes described by this component.

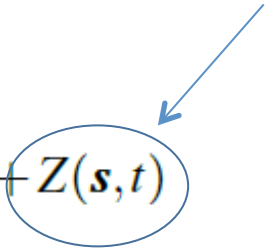
e.g. Variational Bayesian Principal Component Analysis (VBPCA):

- An EOF analysis method that accommodates incomplete data and (basic) measurement uncertainty.
- Other methods are also being explored



LOCAL MODEL

Temperature = Climatology + Low rank decomposition + Local process

$$T(\mathbf{s}, t) = C(\mathbf{s}, t) + \mathbf{w}^T(\mathbf{s})\mathbf{x}(t) + Z(\mathbf{s}, t)$$


Traditional kriging likely infeasible at EUSTACE scales.

Z is Big! So let's map to a lower resolution...

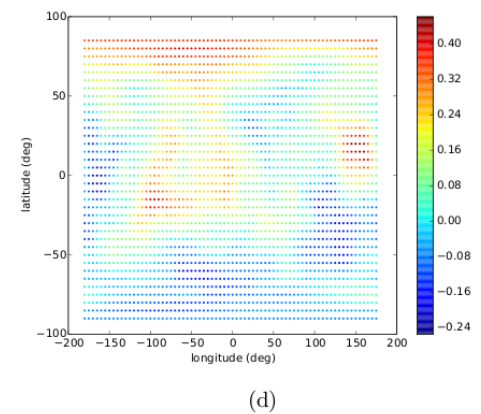
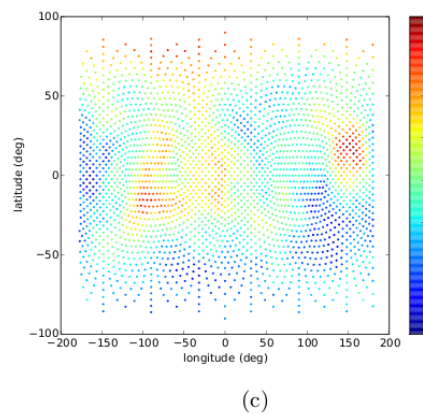
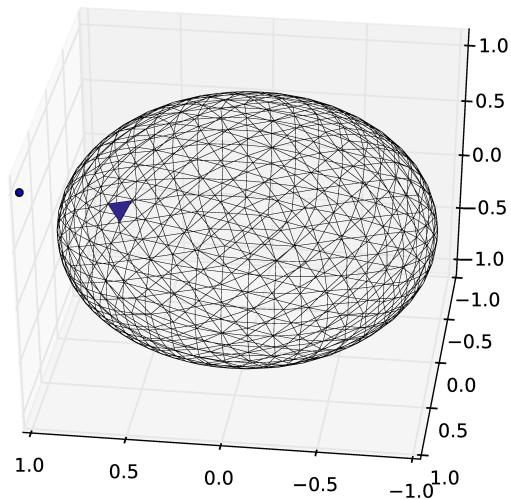
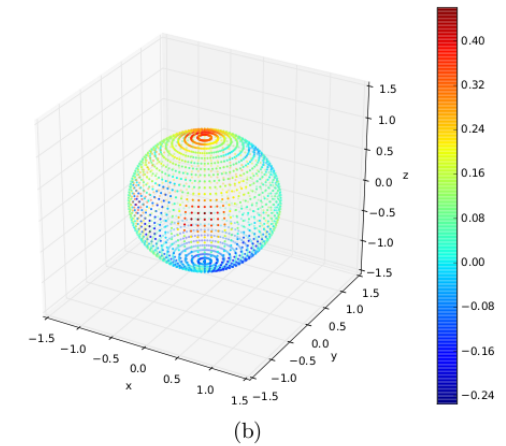
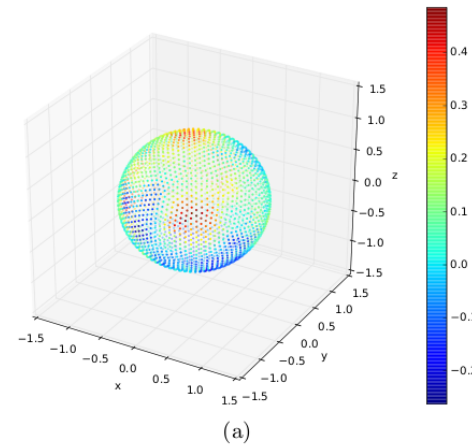
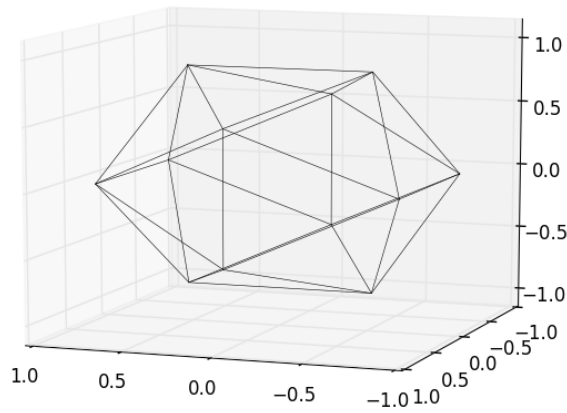
$$Z(\mathbf{s}, t) = \mathbf{A}_z(\mathbf{s}, t)\mathbf{z}$$

Assumption is that covariates describe variation at finest scale.

Testing:

- Kriging with non-stationary covariance function (Paciorek & Shervish, 2004), on moving window;
- Compactly supported basis functions with weights from SPDE (Nychka et al., 2015);
- SPDE on triangulated sphere (Lindgren et al., 2011).

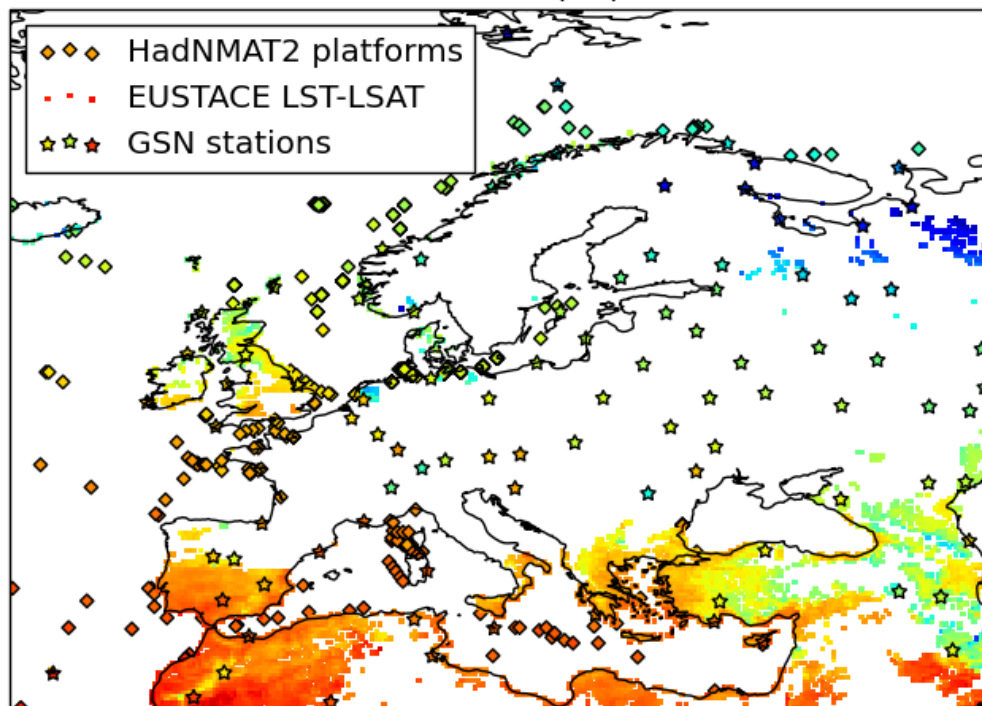
ALL CORNERS OF EARTH



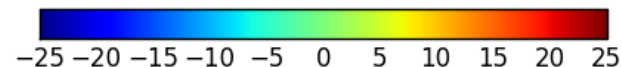
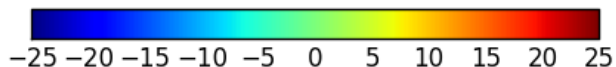
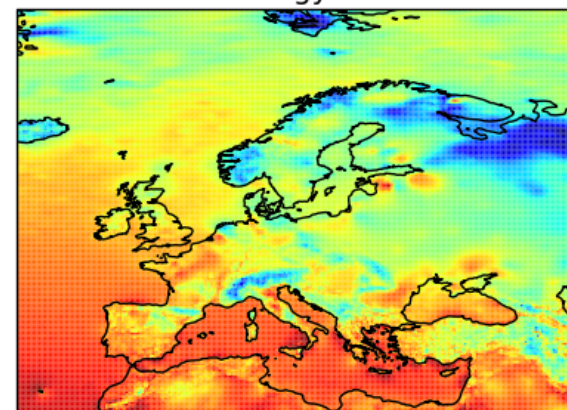
Grid does not converge at poles.

CREATE COMPLETE FIELDS

Observations 01/03/2008



Climatology + Local





Royal Netherlands
Meteorological Institute
Ministry of Infrastructure and the
Environment



VALIDATION



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



HOW WILL WE KNOW IF EUSTACE PRODUCTS ARE ANY GOOD?

We will:

- Adhere to standardised validation protocols for each surface type, with the aim to fit the profile of Quality Assurance for Essential Climate Variables Data (QA4ECV)
- Extend existing principles from SST and LST to match up collocated surface air temperature against *in situ* measurements across all surfaces:
 - Withhold sufficient fully-SI-traceable measurements for the highest quality validation, ensuring full independence from project output datasets
 - Validate the uncertainty information to increase confidence in the quality of the data products
- Inter-compare on common spatial and temporal grids the output surface temperature datasets with other surface air temperature datasets and reanalysis data



**University of
Leicester**



USER ENGAGEMENT WITH EUSTACE

EUSTACE products will be made for wide range of users: researchers, policy makers, health, agriculture, etc.

User engagement needed **from the start** to assure usability/salience

Activities in which users have been engaged and/or informed:

- User workshops (Virtual user consultations, and at EGU 2015 and 2016)
- As trail blazers
- Dissemination on the progress and products through web site
- Guidance on use of products

EUSTACE will have been successful when the products are used.

If you are interested, please contact bessembi@knmi.nl



Royal Netherlands
Meteorological Institute
Ministry of Infrastructure and the
Environment





EUSTACE PRODUCTS



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171



EUSTACE PRODUCTS

Product	Description	Date
Station series and E-OBS update	Global data set of daily weather station air temperature measurements (Tmax and Tmin) with non-climatic breaks identified – Station time series and gridded for Europe	Dec 2016
Satellite skin temperature retrievals	Daily satellite skin temperature estimates for all surfaces of Earth with consistent uncertainty estimates – Gridded or along satellite's track	mid 2016
Skin/air temperature relationships	Understanding of the relationship between surface skin and surface air temperature over all surfaces of Earth and in different seasons – A report	Dec 2016
Air temperature estimates from satellites	Daily estimates of surface air temperature from skin temperature retrievals - Gridded or along satellite's track (TBD)	Jun 2017
Globally complete air temperature fields	Globally-complete daily fields of surface air temperature over all corners of Earth since 1850 – Gridded (0.25° lat/lon) perhaps an ensemble. (Tmax and Tmin over land, Tmean elsewhere.)	May 2018
Derived products	For example, global means and climatologies	May 2018

SUMMARY

EUSTACE will give publicly available daily estimates of surface air temperature since 1850 across the globe for the first time by combining surface and satellite data using novel statistical techniques. To do this, we need to:

- Identify non-climatic discontinuities in daily weather station data, *so users can trust the changes our records show*
- Understand how surface temperature measured *in situ* and by satellite relates, *to estimate air from skin temperature*
- Estimate values in areas where we have no *in situ* or satellite data, *so users can have daily information here*
- Validate our estimates using independent data





MANY THANKS FOR LISTENING

TIME FOR QUESTIONS



EUSTACE has received funding from the European Union's Horizon 2020 Programme for Research and Innovation, under Grant Agreement no 640171

