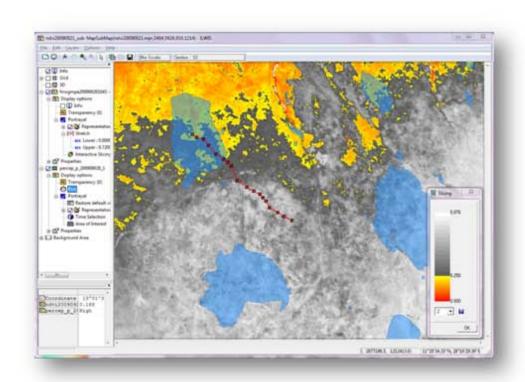


CONTENTS

- ✓ Problem
- ✓ How to reduce ...
- ✓ Use of the toolbox
- ✓ Conclusions
- √ Future





PROBLEM

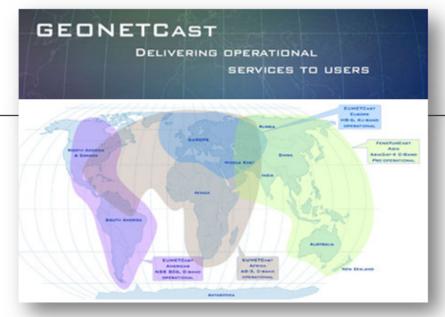
Long time series of RS data available to study behaviour of dynamic phenomena

Problems if animated with few controls

- √ extraction of subsets
- √ comparisons
- ✓ change blindness → mixture of bottom-up (driven by visual info)

 and top-down (attention- or task-driven)

Attention: crucial role in change perception, but limited!



http://www.earthobservations.org/geonetcast.shtml



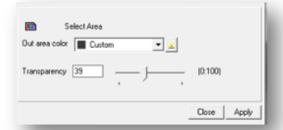
UNIVERSITY OF TWENTE.

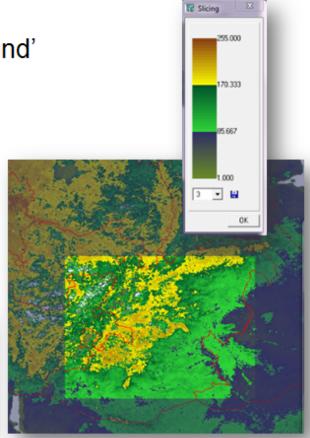
HOW TO REDUCE ...



Our aim is to facilitate top-down processing through effective (bottom-up) visualization:

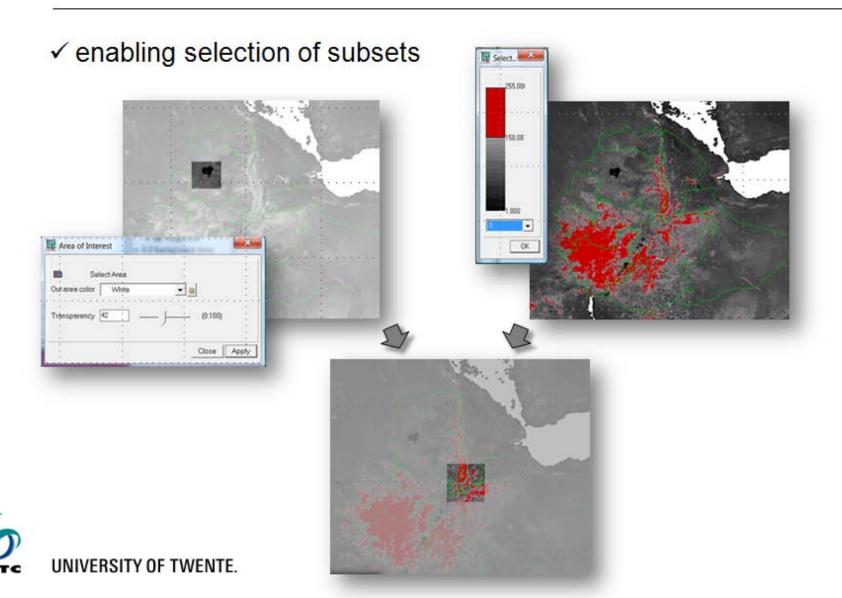
- by letting 'figures' stand out against a 'ground'
 - √ tracking of objects → polygons
 - √ classification
 - √ enabling selection of subsets
 - ✓ option to keep the context ('figure') around a selection visible, but subdued ('ground')







HOW TO REDUCE ...



HOW TO REDUCE ...

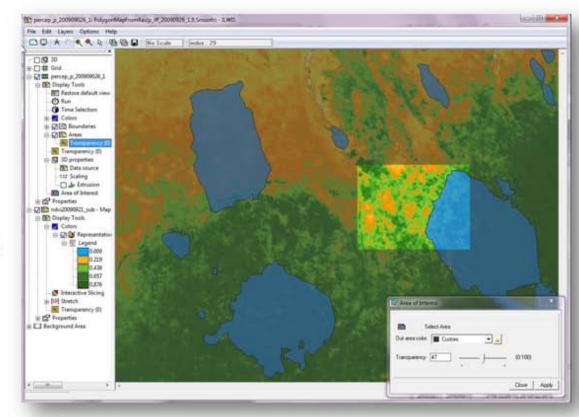
- by adding other functions
 - ✓ multiple layers
 - √ control of display speed



http://52north.org/

. . . .

Implementation:
extension of ILWIS
for visual exploration
of animated time series





Ethiopean lowlands:

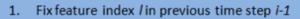
- √ serious droughts
- √ flash floods

Explore relations vegetation \iff precipitation:

- ✓ Spot-5 VEGETATION: NDVI
- ✓ MSG: Multi-Sensor Precipitation Estimates

Algorithm detects precipitating features and binds them into objects

- √ polygons
- ✓ paths



2. For a given feature index *l*, determine feature index *k*_{MQ} in current time step *j* with maximum spatial overlap:

$$k_{MO}(l) = argmax [Size(F_l^{i-1} \cap F_k^i)]$$

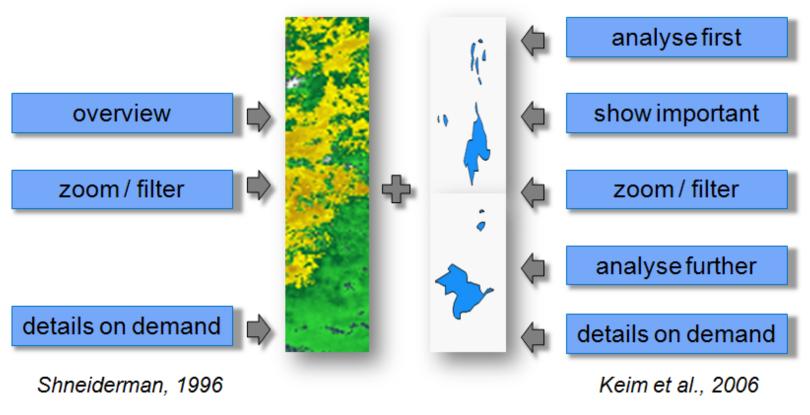
For a given feature index kmo, determine feature index lmo previous time step i-1 with maximum spatial overlap:

$$I_{MO}(k_{MO}) = argmax [Size(F_l^{i-1} \cap F_{k_{MO}}^i)]$$

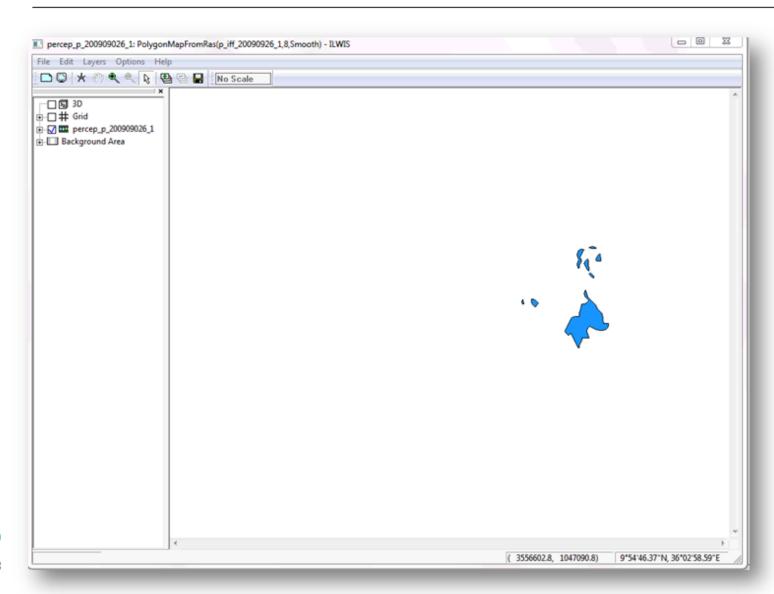
4. Check if feature indices are the same: I_{MO}=I



Use strategies









Early evaluation of selected visualization functions



Monitoring tasks – NDVI time series of Ethiopia

✓ 8 participants ('with') : toolbox with the new functions

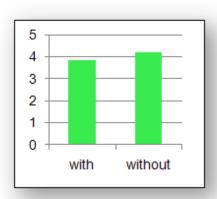
√ 8 participants ('without') : toolbox without the new functions

Methods used: think aloud, observation, questionnaires, interview

Effectiveness



Efficiency (av. time / task, minutes)



Satisfaction:

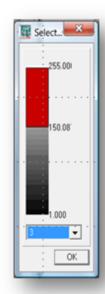


comparable (high) rates, but 'with' had more confidence in the tasks

Less difference than expected due to bugs / deficiencies

Improvements made in:

- √ selection of attribute values
- √ display of variations within selected values
- √ dynamic link between legend and display area
- √ speed control and interface



Main rendering: by the graphics hardware (OpenGL)

Potential bottlenecks:

- ✓ raster: amount of available graphics RAM
- ✓ vector: number of triangles needed to render the polygons.

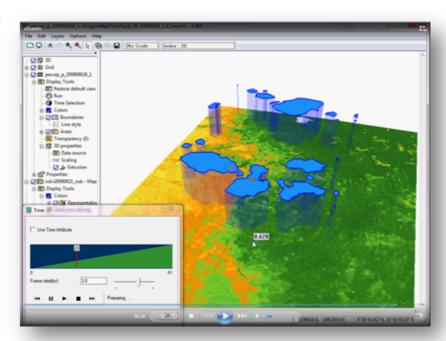


CONCLUSIONS

Single Open Source environment:

- √ access to GEONETCast data
- ✓ pre-processing
- ✓ analytical and animation functions

α - version readybut further work to be done ...





FUTURE

Preprocessing of time series:

- √ projection / resampling of images
- √ filling missing pixels
- √ image smoothing

Analytical / visualization functionality:

- √ tracking algorithms
- √ quantification / visualization of attributes
- √ options to compare / synchronize time series
- ✓ optimization of the temporal legend



Evaluation: do the tools reduce change blindness!?

