

Weighted median filters for meteorological data analysis

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EWGLAM/SRNWP – Athens 2009

Introduction – need of adequate scale separation

- *Meteorological systems work in various characteristic time and space scales*
 - *To compare or merge data of various origin we always meet question of scales*
 - *It is matter of prime importance to have adequate scale separation methods*
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Introduction – from current methods to CWM filters I

- *Current methods suffer because of sensitivity to outliers, depend on assumption of error distribution, depend on lack of data*
 - *To process effectively huge amount of data of different origin meteorological systems need unsupervised scale separation methods*
 - *Robustness become a key feature*
 - *Proper analysis of non-smooth meteorological fields – e.g. precipitation – is a challenge especially in high resolution NWP*
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Introduction – from current methods to CWM filters II

- *Simple median filters are widely applied in image processing for noise removal*
 - *Median filter is effective and robust tool*
 - *Median filter keeps contrast untouched*
 - *Mentioned above features suggest that median filter is good basis for search for new scale separation method applicable to non-smooth meteorological fields e.g. precipitation, cloudiness*
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Introduction – from current methods to CWM filters III

- *Proposed new tool for scale separation is a set of Composite Weighted Median Filters i.e. complex filters which are superposition of weighted median filters with different sizes and weights*
 - *1-D and 2-D CWM filters for gridded data are proposed in the presentation*
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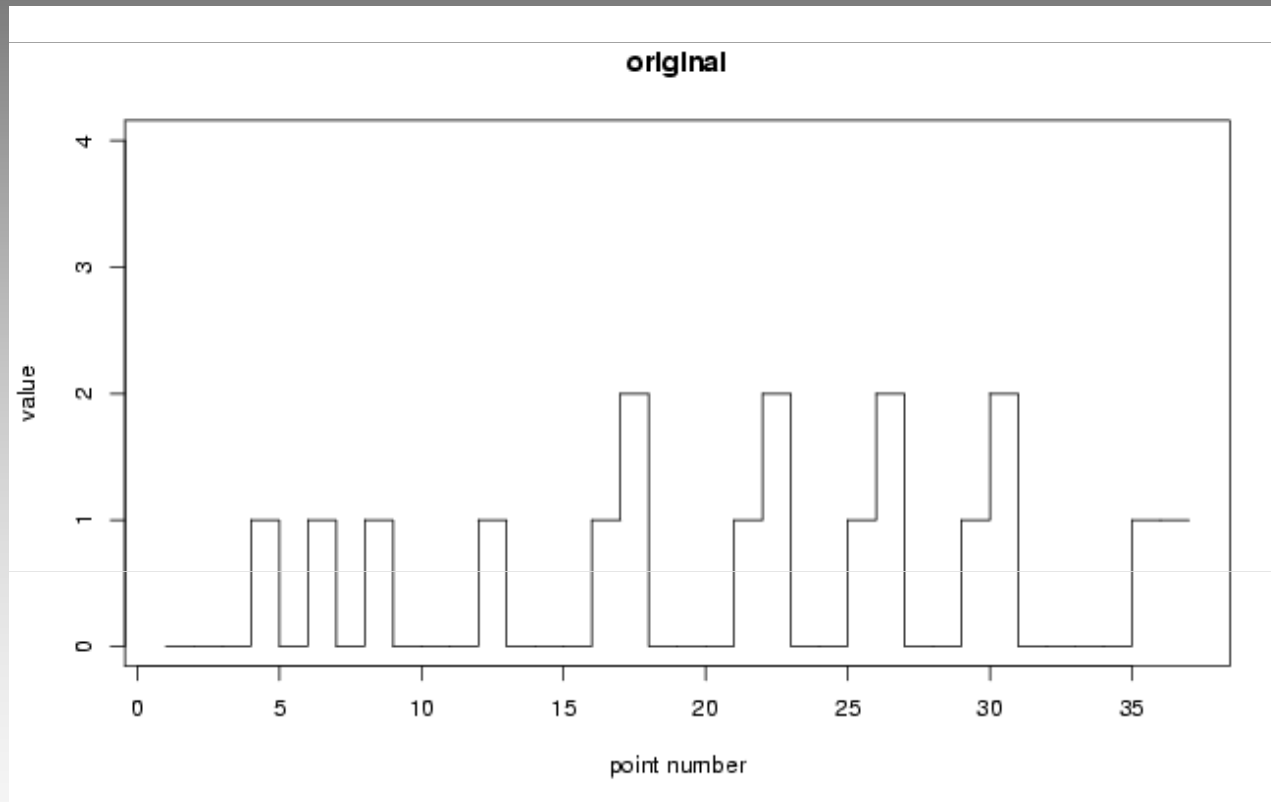
1-D and 2-D CWM Filters – recipes

- *1-D CWM filter is superposition of 2 kinds of elementary bricks: A-brick and B-brick*
 - *A-brick of level l is WM filter of size $2*l+1$ and weights $(1,2,\dots,2,1)$*
 - *B-brick of level l is WM filter of size $2*l+1$ and weights $(1,1,\dots,1,1)$*
 - *CWM filter F of level l is defined as follows:
 $F(l) = B(l)A(l)B(l-1)A(l-1)\dots B(1)A(1)$*
 - *2-D CWM filter is built similarly but elementary brick is not just weighted median filter but weighted median of weighted medians filter*
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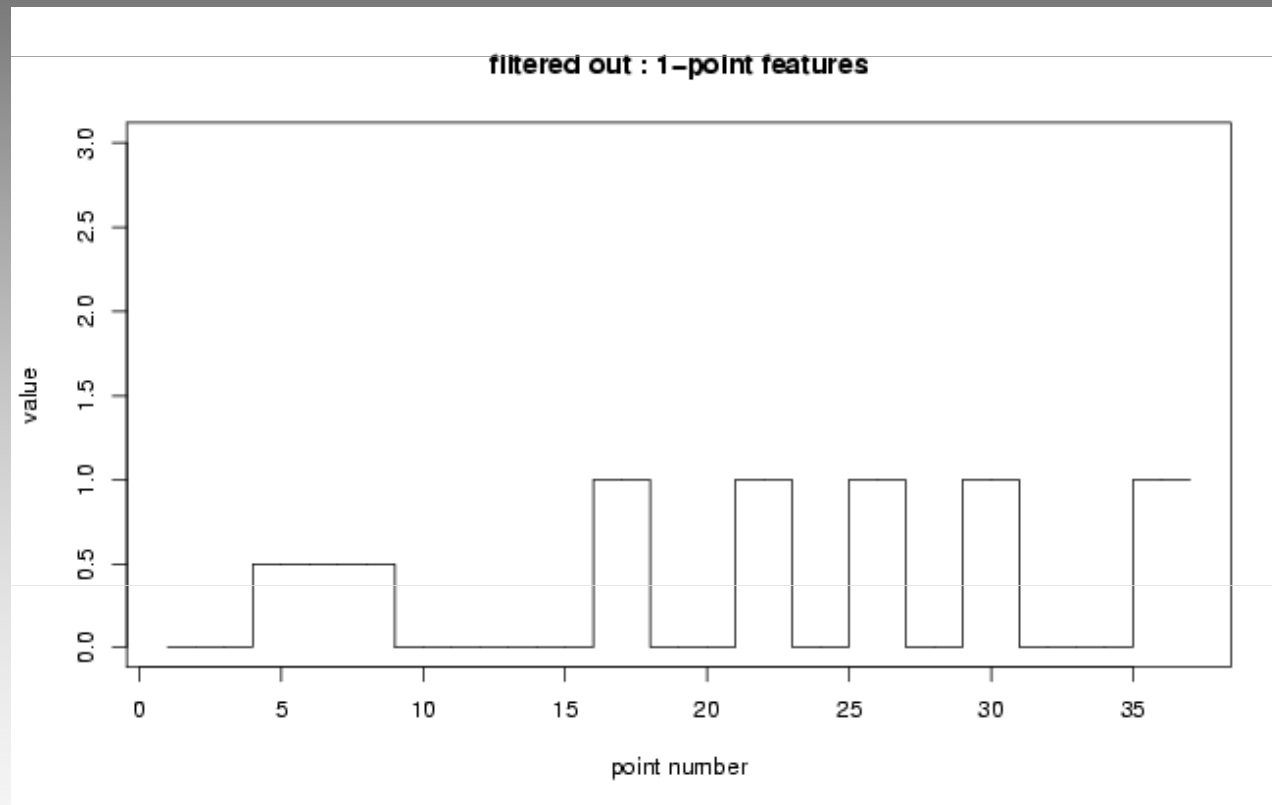
1-D and 2-D CWM Filters – how they work

- *Shortly speaking: 1-D CWM filter of level l leaves peaks of size $l+1$ points or greater and averages $2l$ -point waves*
 - *Shortly speaking: 2-D CWM filter of level l leaves features of size $(l+1) \times (l+1)$ or greater and averages $2l$ -point waves*
 - *Let's take a look for examples !*
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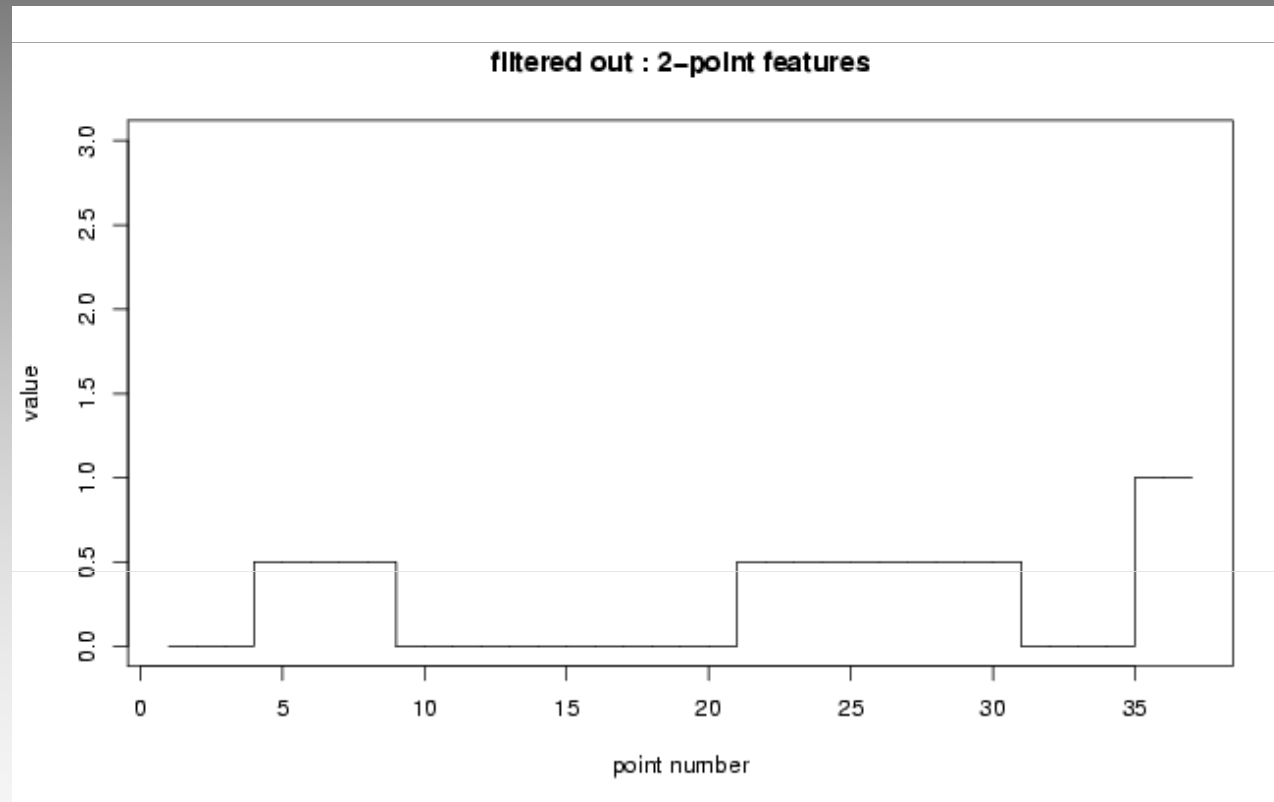
1-D filtering example – original



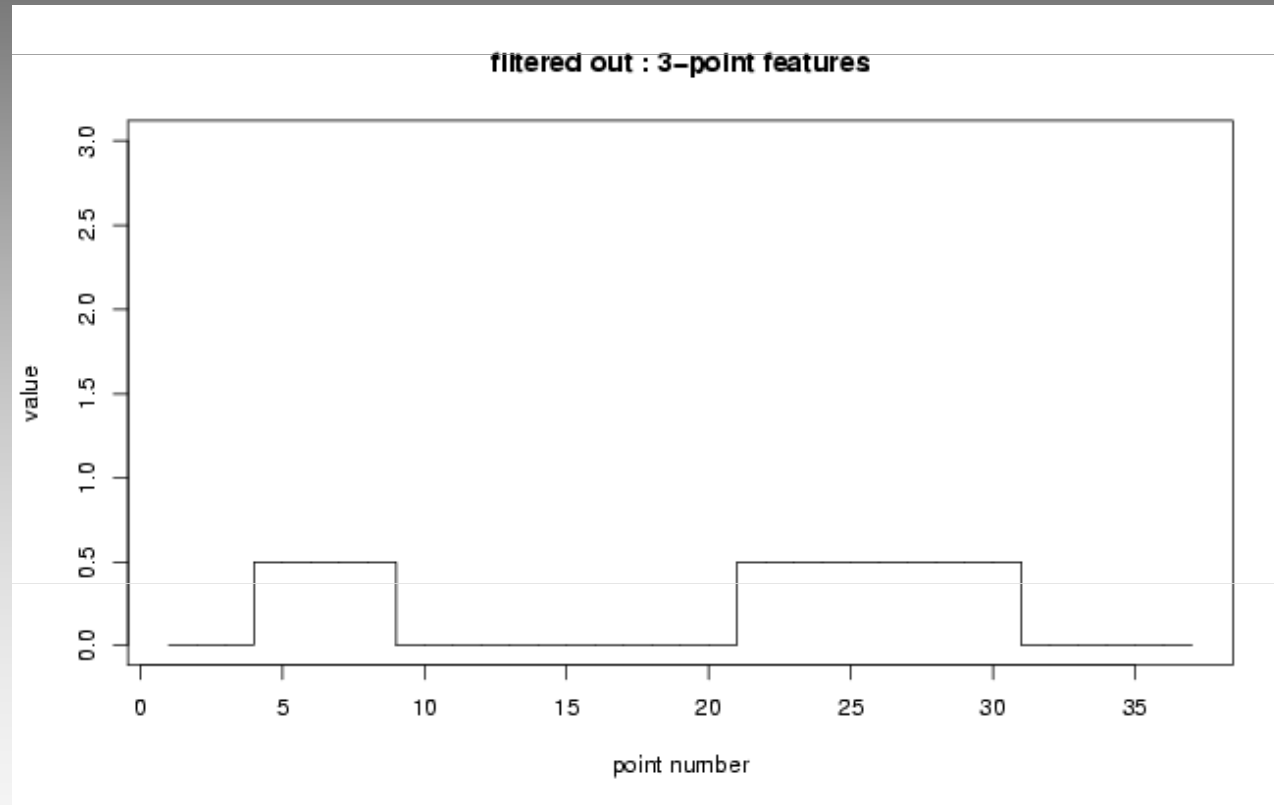
1-D filtering example – 1-point features filtered out



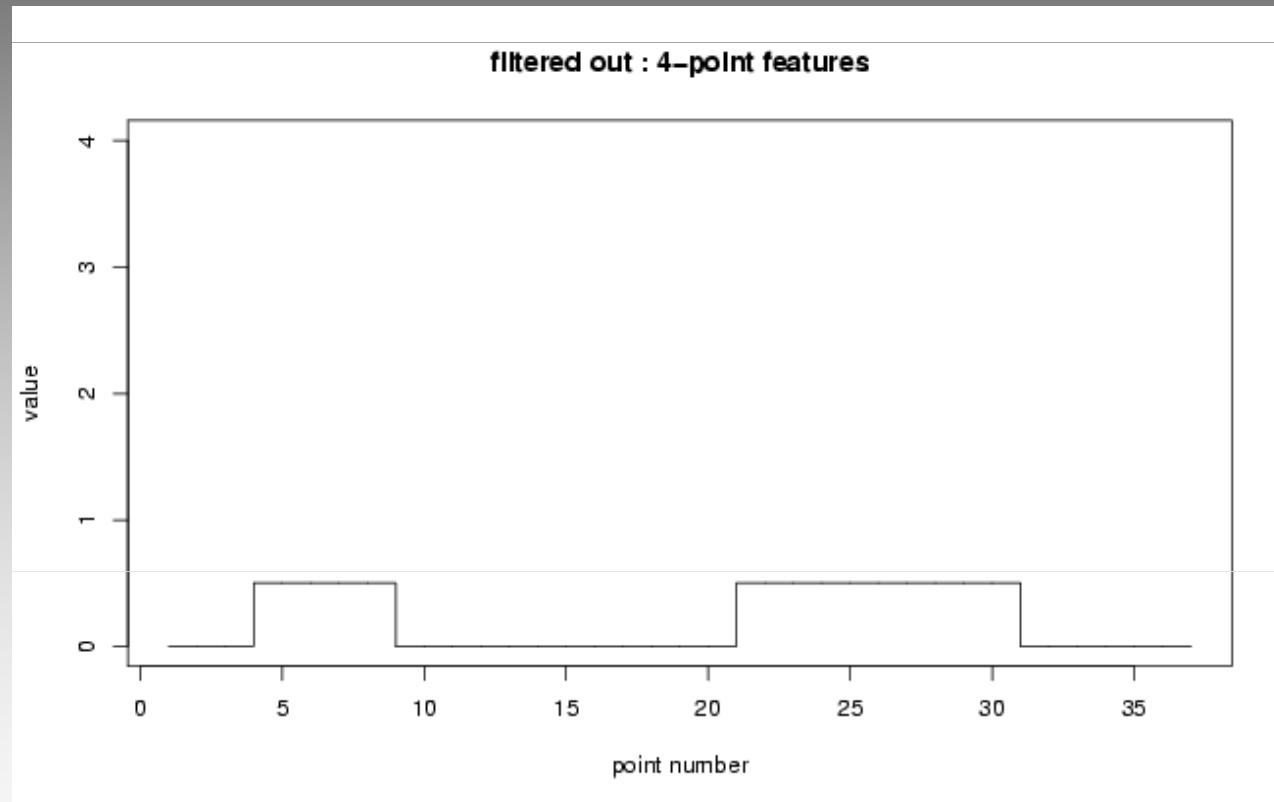
1-D filtering example – 2-point features filtered out



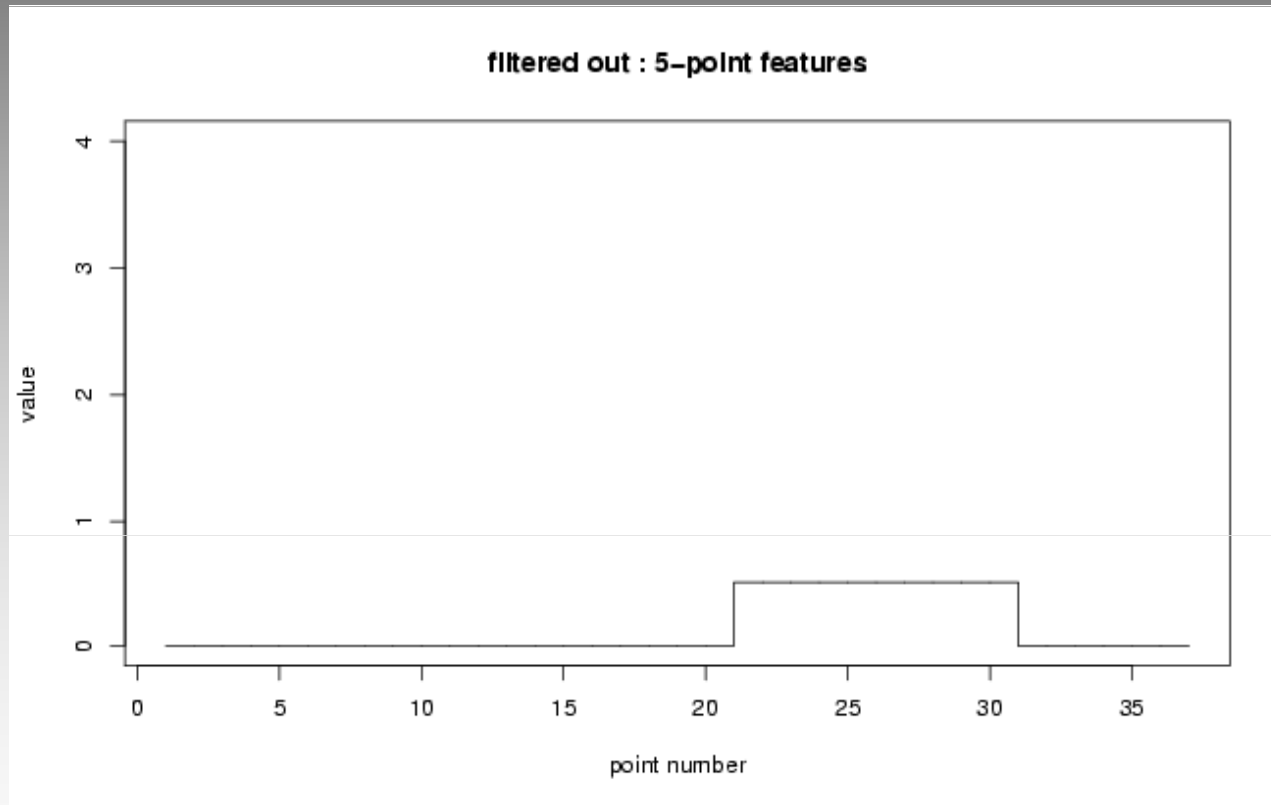
1-D filtering example – 3-point features filtered out



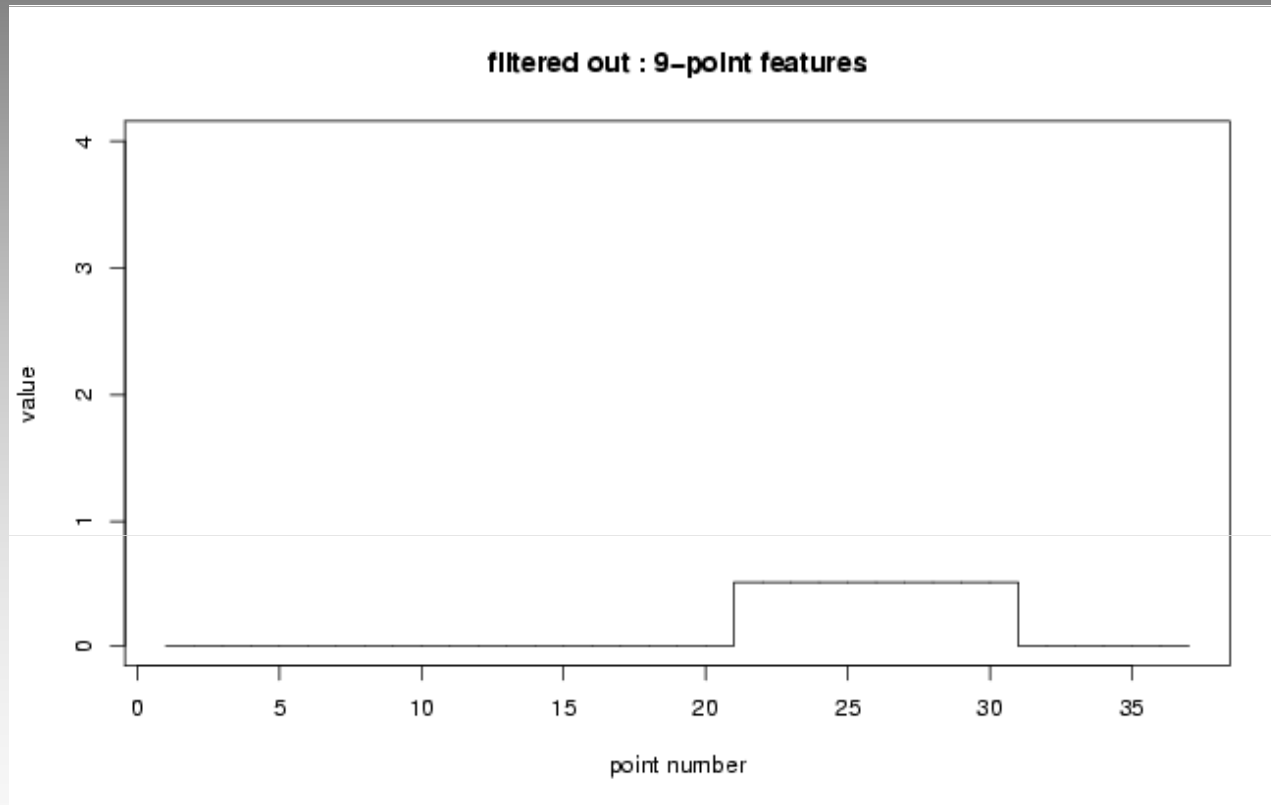
1-D filtering example – 4-point features filtered out



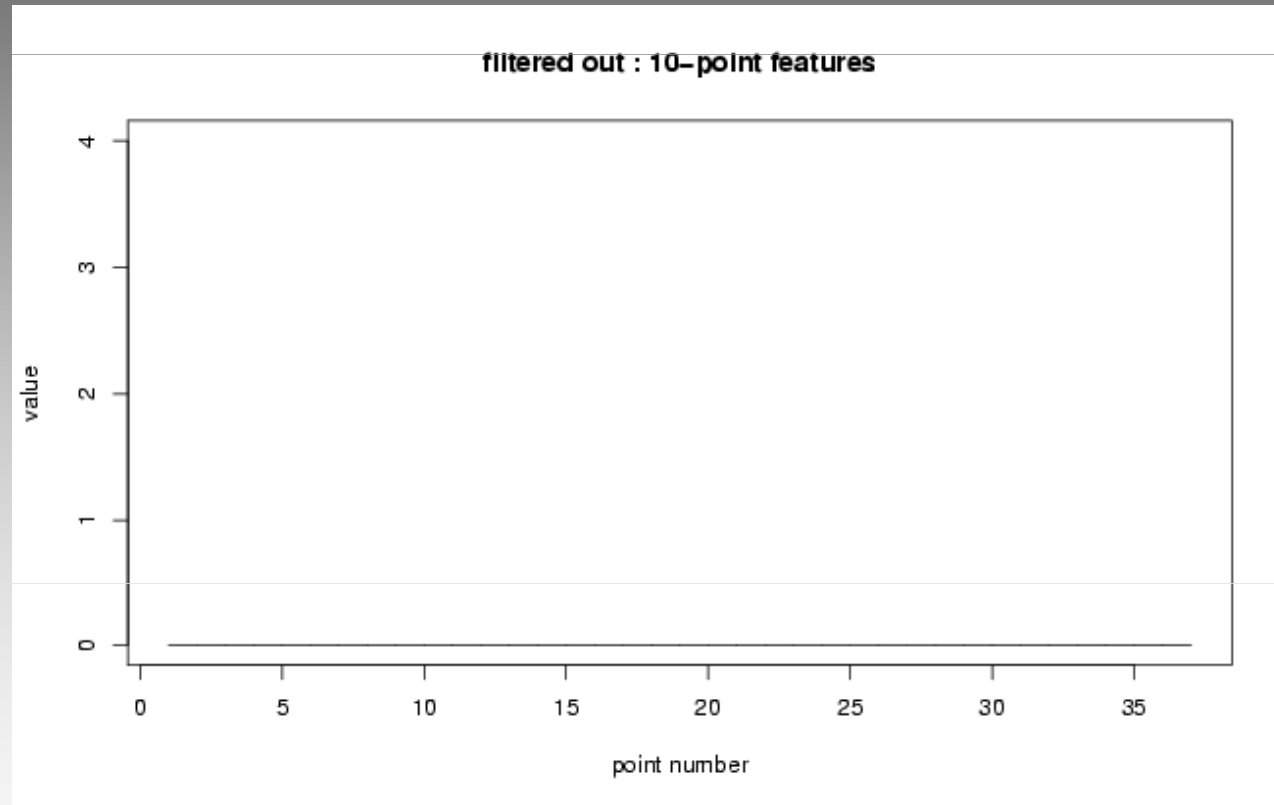
1-D filtering example – 5-point features filtered out



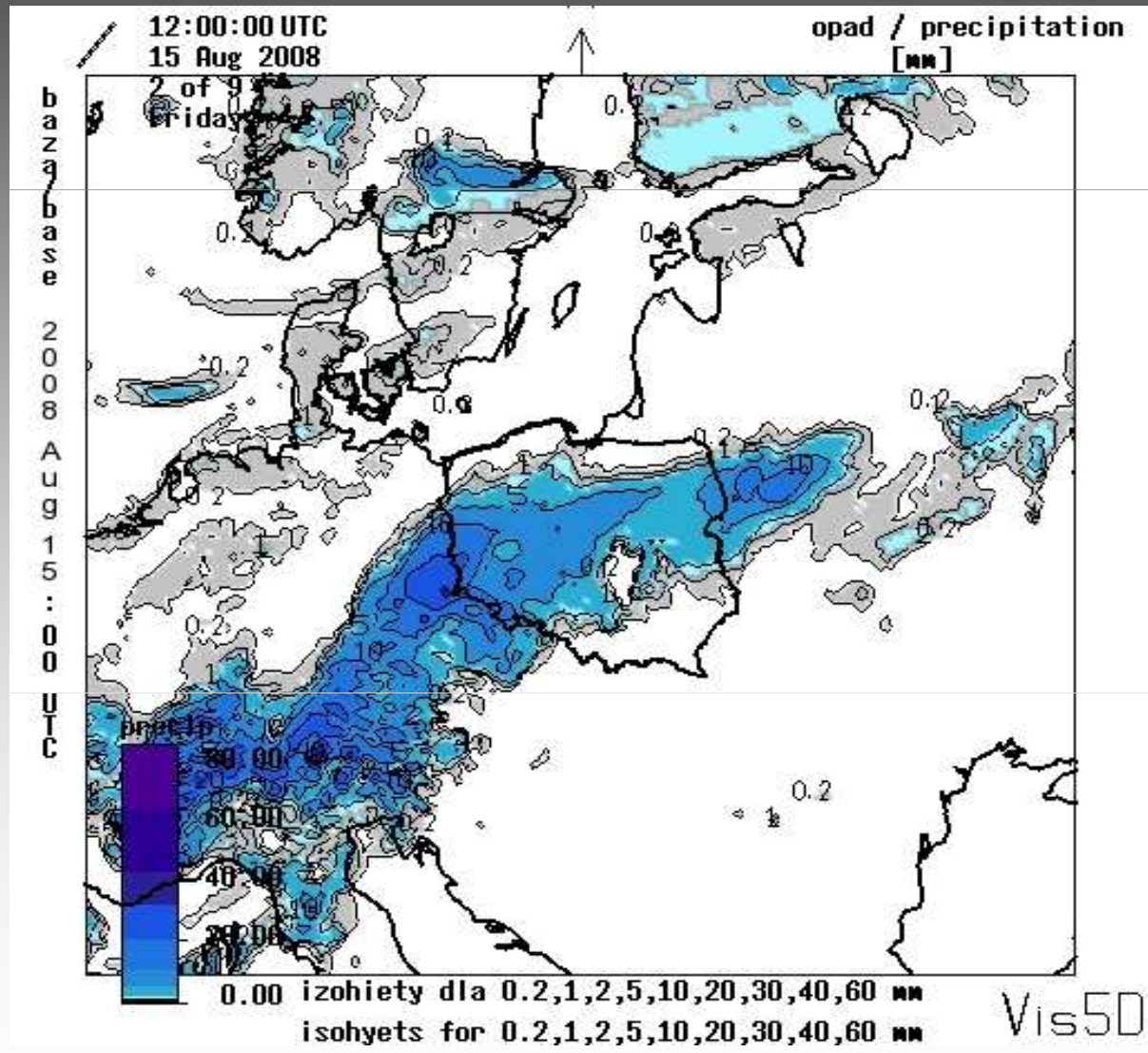
1-D filtering example – 9-point features filtered out



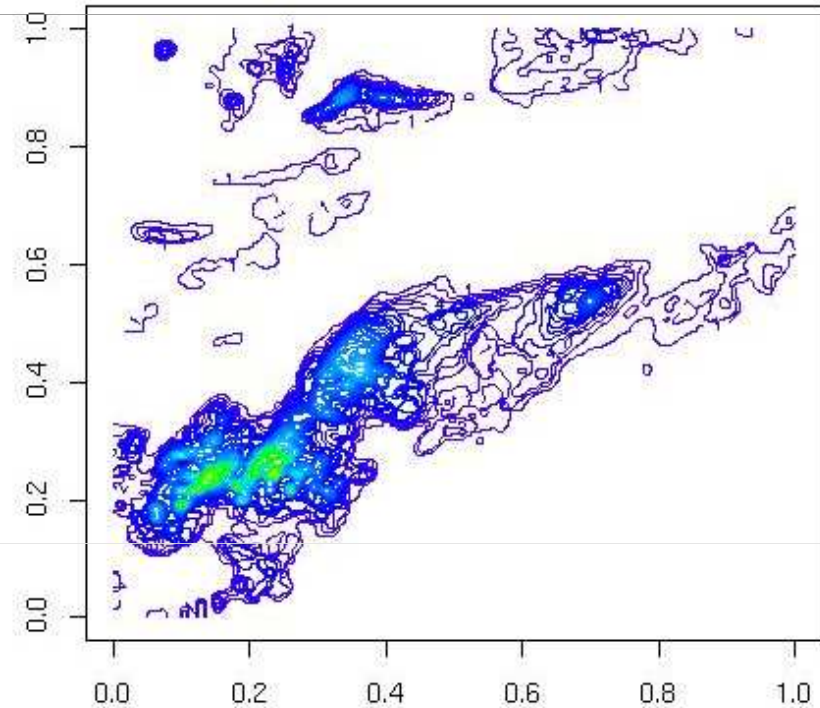
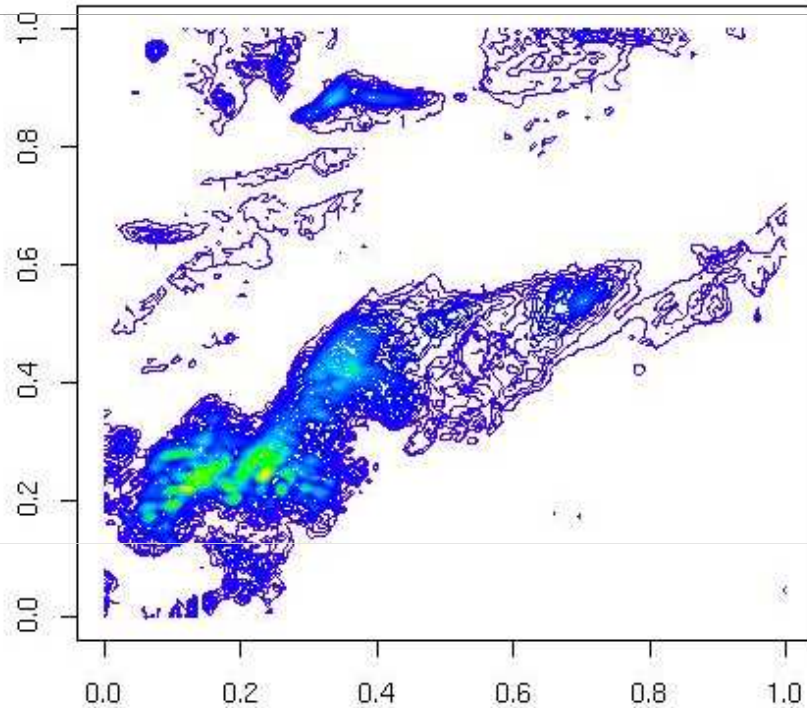
1-D filtering example – 10-point features filtered out



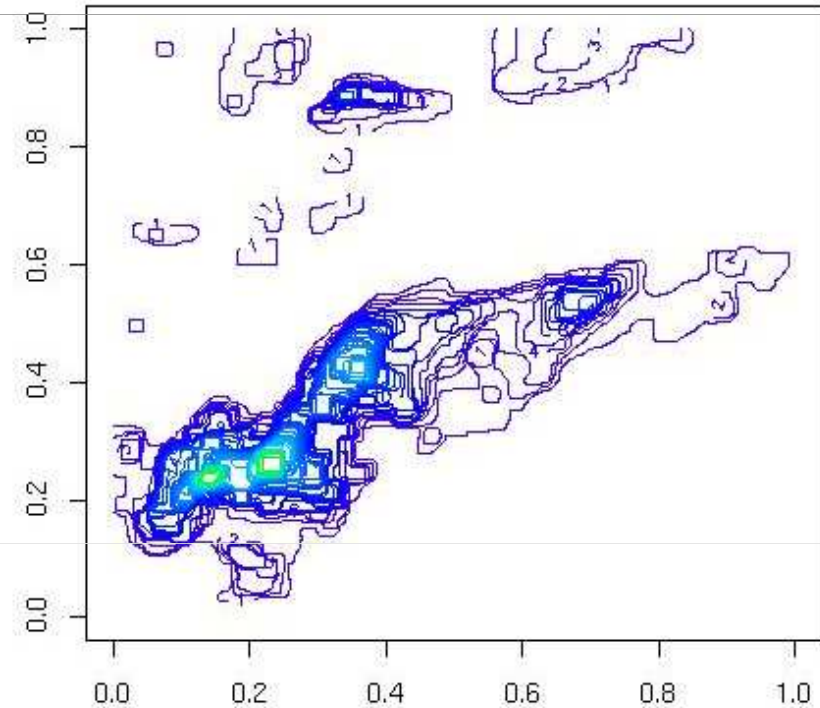
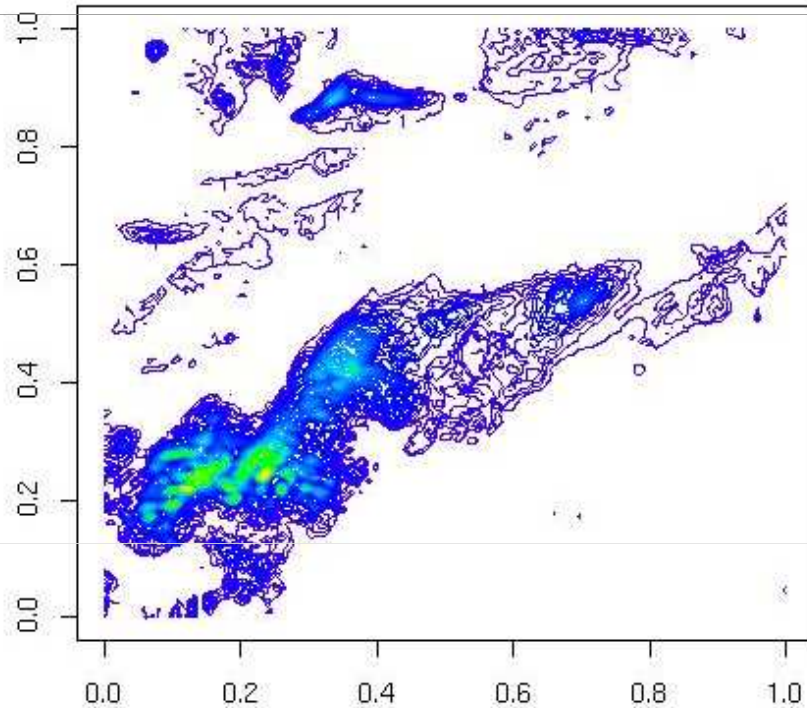
2-D filtering example - 12-hour cumulated precipitation field



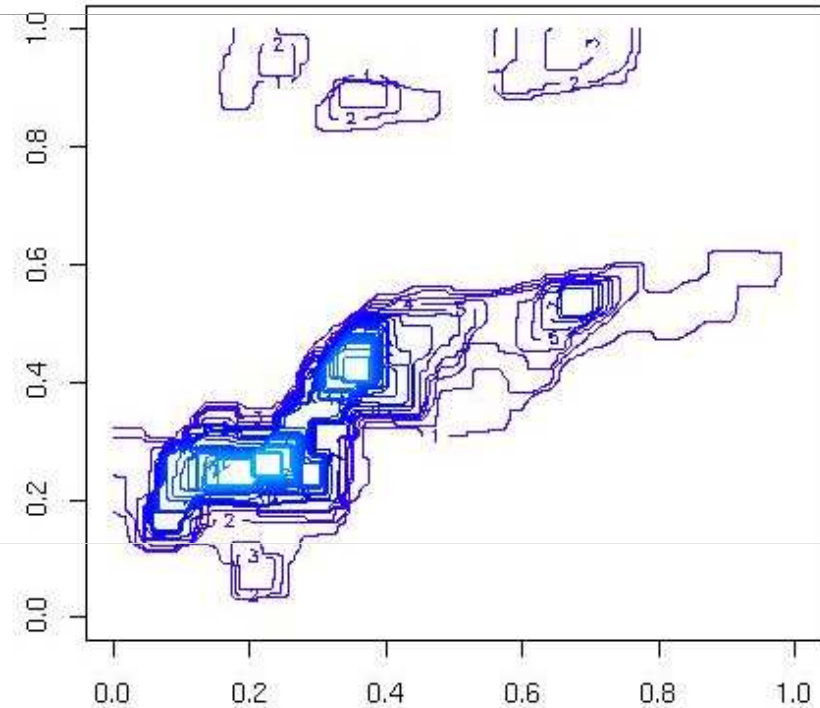
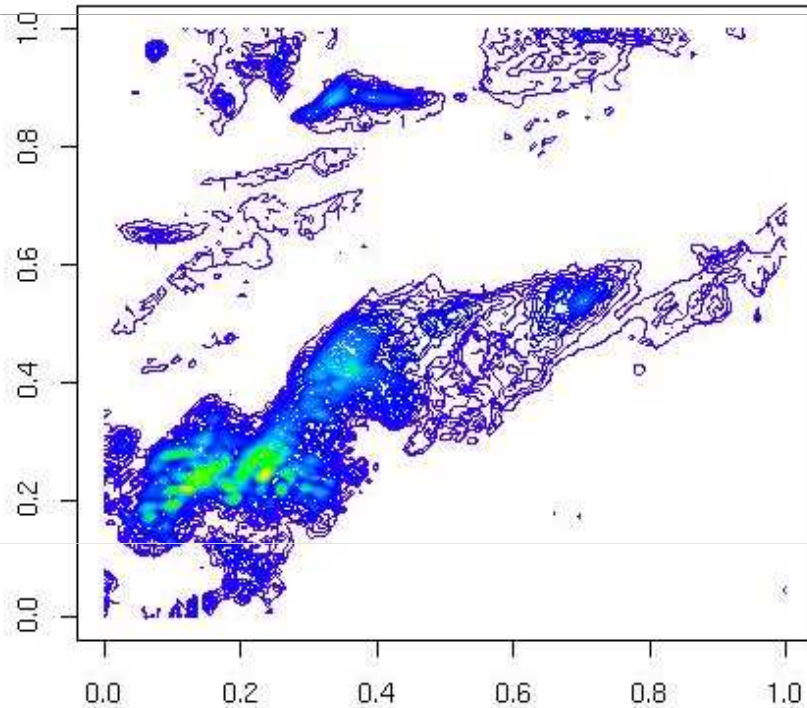
2-D filtering example – original vs 1-point features filtered out



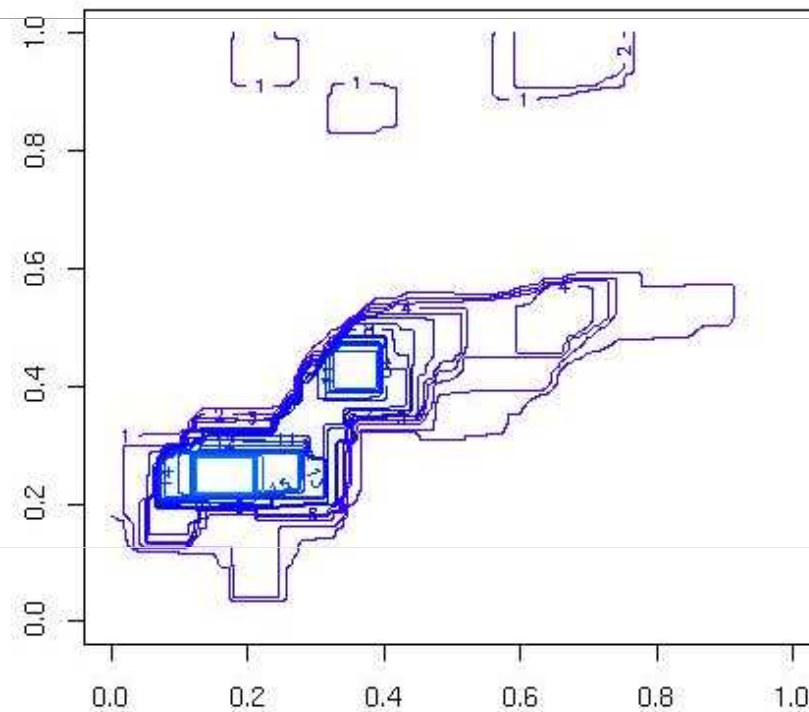
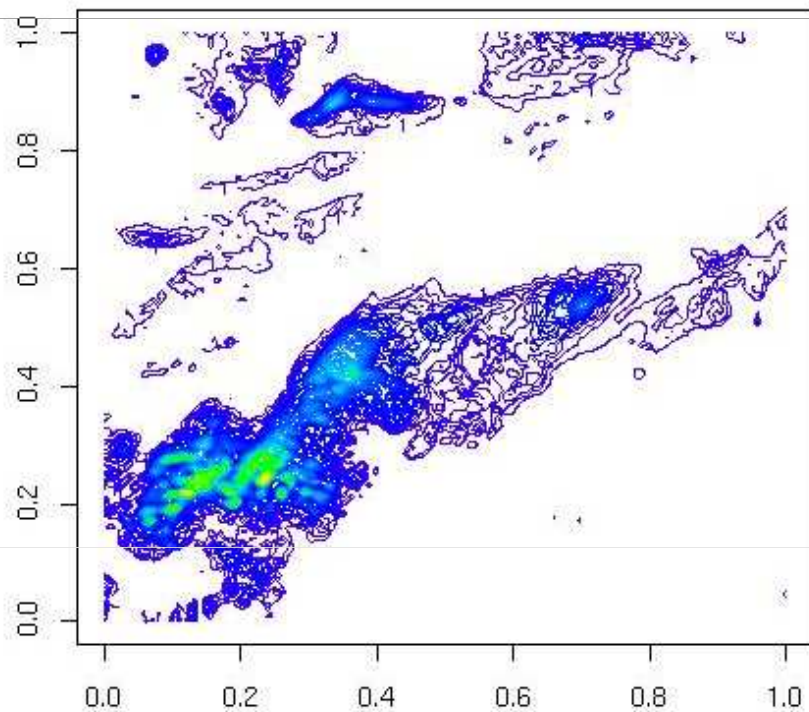
2-D filtering example – original vs 3-point features filtered out



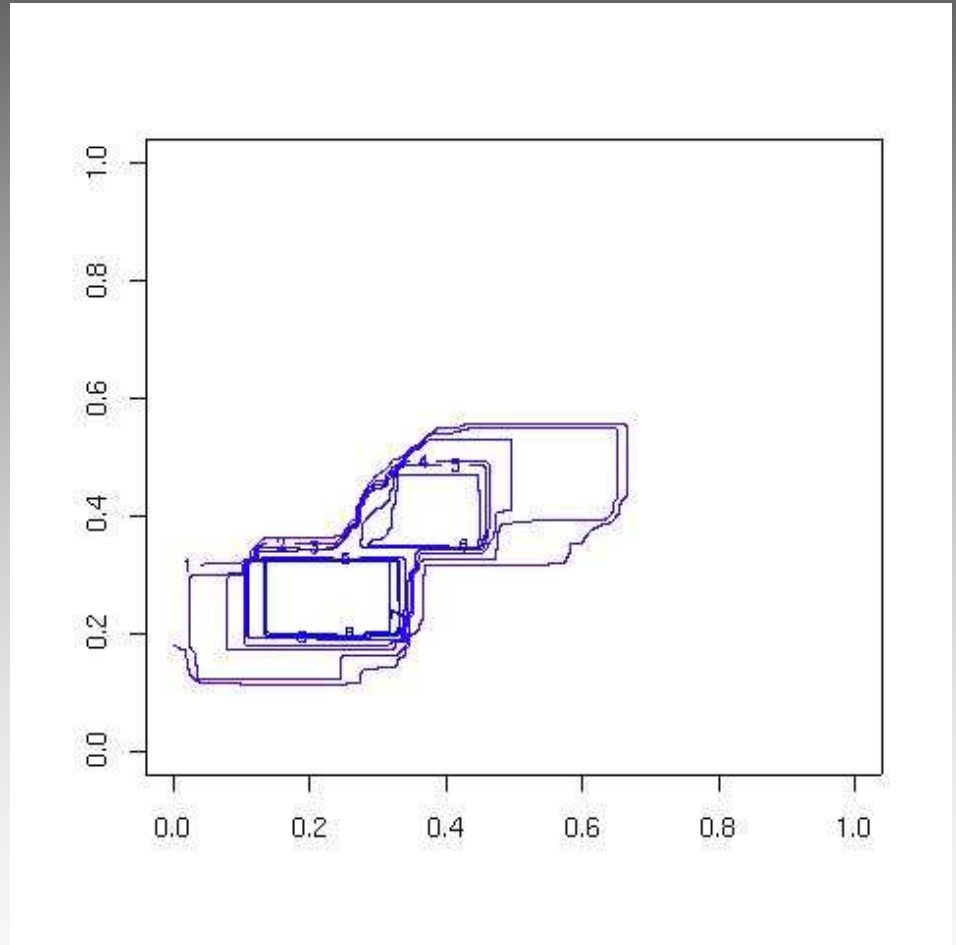
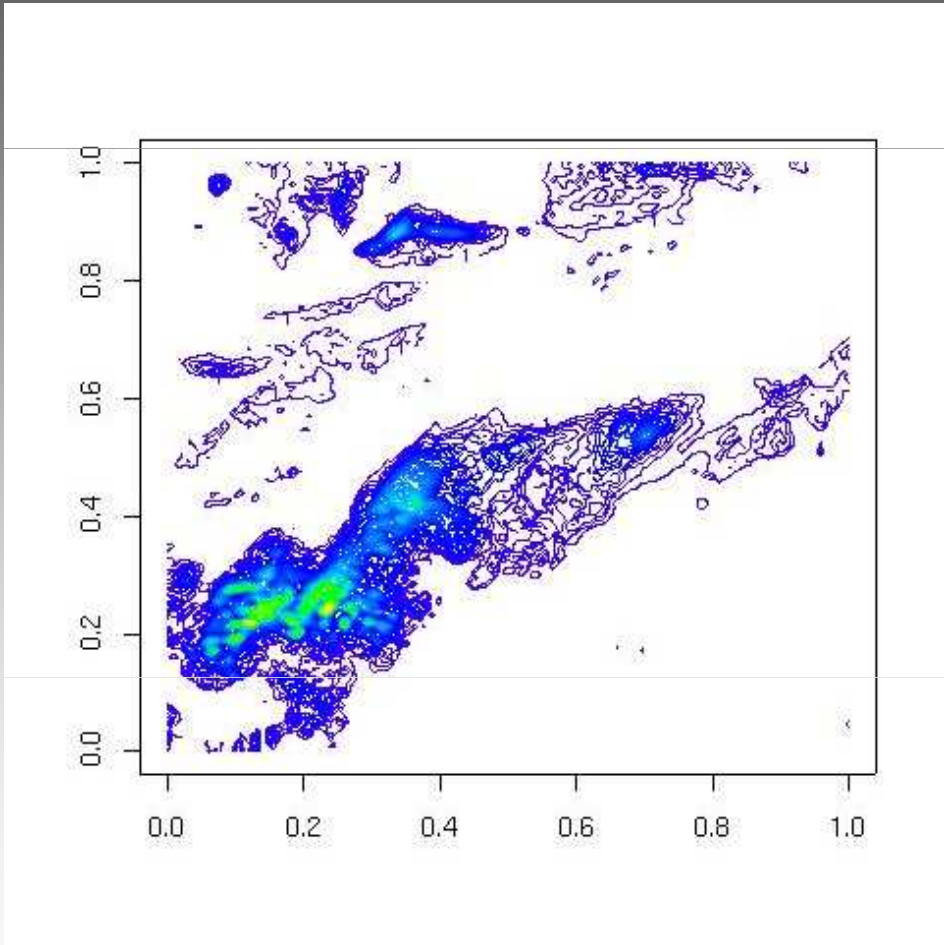
2-D filtering example – original vs 6-point features filtered out



2-D filtering example – original vs 10-point features filtered out



2-D filtering example – original vs 20-point features filtered out



CWM Filters – open questions I

- *Is CWM filter an idempotent operator ?*
 - ✓ *idempotence is desired property for filters*
 - ✓ *this question needs detailed mathematical investigation yet*
 - ✓ *CWM filter applied to set of primitive shapes shows such property*

CWM Filters – open questions II

- *Is directional dependence of 2-D CWM filter procedure a deficiency ?*
 - ✓ *directional dependence of the procedure can be overcome by taking minimum or average of two filter passes: x axis - oriented and y axis – oriented ones*
 - ✓ *even one-pass CWM filter applied to some primitive shapes shows no directional dependence of results - both passes give the same*
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CWM Filters – deficiencies I

- *For efficient filtering mean derivative of analysed field averaged over domain should be close to zero:*
 - ✓ *for large domains, bigger than analysed patterns, it isn't serious problem at all*
 - ✓ *for small domains boundary conditions should be posed properly*
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CWM Filters – deficiencies II

- *High computational cost of large scale features removing*
 - ✓ *forget about “cheap” statistics - modern statistical methods need more and more computational power*
 - ✓ *it is tool rather for research purposes*
 - ✓ *small scale patterns removal is not expensive*

CWM Filters – future research

- *Methods of data analysis which can fully exploit potential of CWM filters and specificity of their results*
 - ✓ *CWM filters rather eliminate finer features than average them*
- *3-D filtering*

CWM Filters – end of tour

Thank you for your attention

