

**Регрессионные модели оценки
урожайности
сельскохозяйственных
культур по спутниковым и
агрометеорологическим данным**



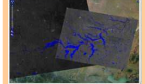
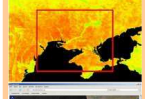
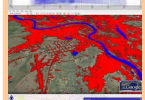
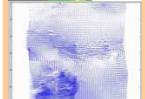
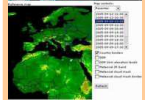
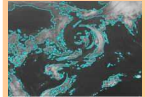
Куссуль Н. Н ., Кравченко А. Н ., Колотий А.
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Лавренюк А. Н ., Куссуль О. М., Грипич Ю. А.

Институт космических исследований НАНУ-
ГКАУ,
Киев, Украина



Context – CRDF project

- **Title:**
 - Analysis of Climate Change & Food Security based on Remote Sensing & In Situ Data Sets
- **Project duration:**
 - 2 years (2011-2012)
- **US Principal Investigator**
 - **Felix Kogan**, PhD, National Oceanic and Atmospheric Agency (NOAA)
- **Overall goal of the project**
 - to analyze climate change and food security based on remote sensing and in situ data sets.



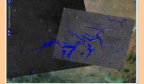
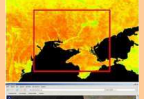
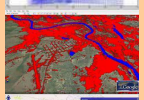
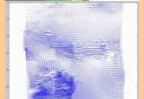
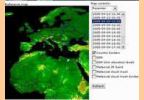
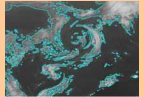
Results

Winter wheat yield forecasting using satellite data



Winter wheat yield forecasting

- Empirical regression model based on NDVI data derived from MODIS and statistical data on winter wheat
 - Time period: 2000-2009
- NDVI averaged for oblasts by crop masks
- Trend eliminated from yield
- Regression model
 - Robust regression
- Regression model is specific for each oblast

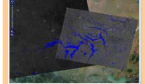
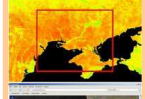
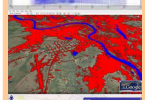
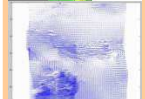
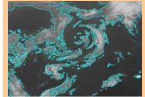


Модели для областей Украины. Выбор признаков

Днепропетровская обл.

Черниговская обл.

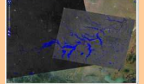
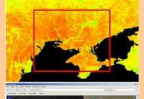
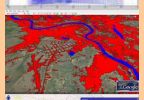
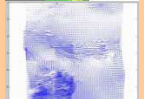
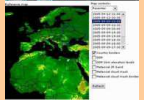
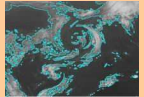
корреляция среднего урожая урожайности на протяжении года
(Днепропетровская и Черниговская области)



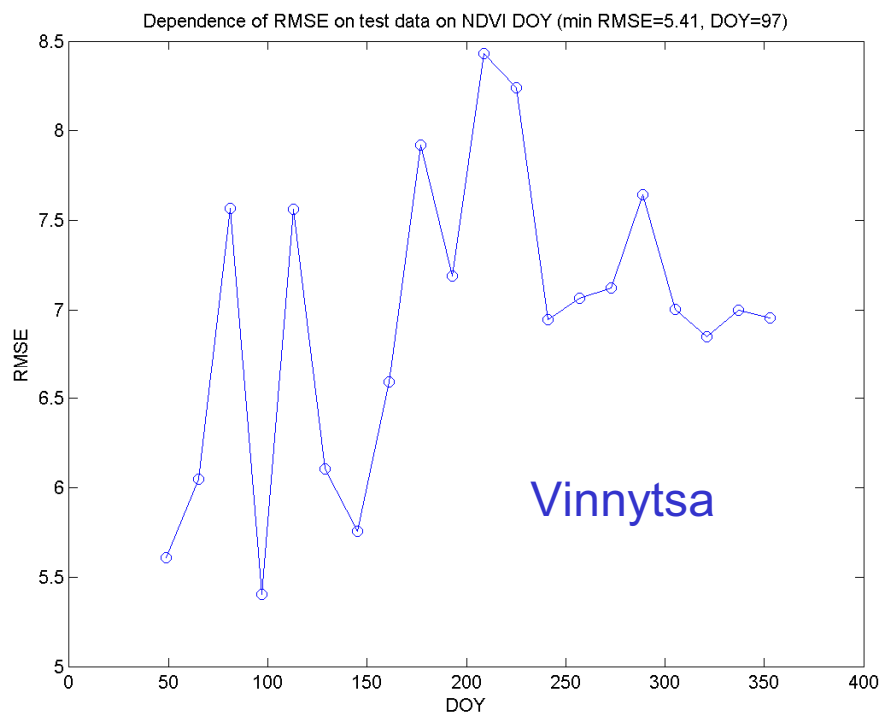
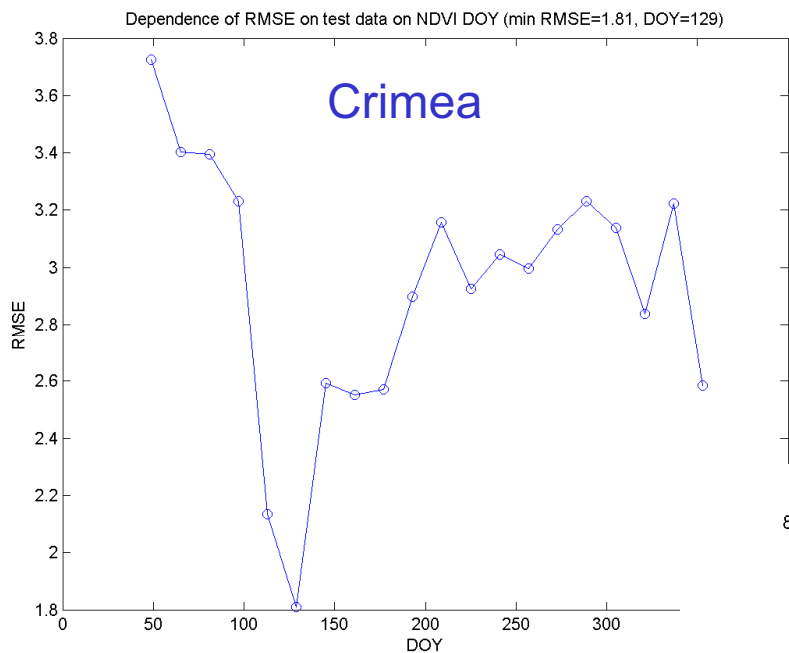
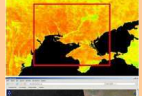
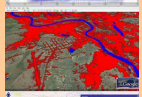
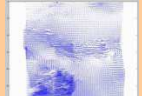
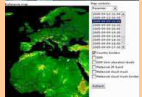
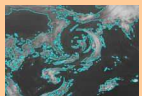
Уменьшение ошибки при использовании среднего $ndvi$ в качестве предиктора

Winter wheat yield forecasting

- Cross-validation
 - leave-one-out cross-validation (**LOOCV**)
 - using a single observation from the original sample as the testing data, and the remaining observations as the training data
- Criteria
 - RMSE on testing data



Feature selection



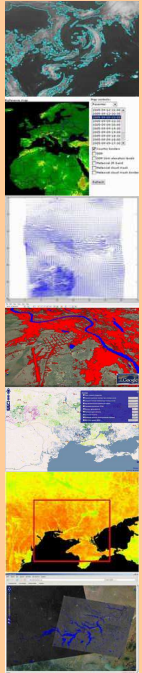
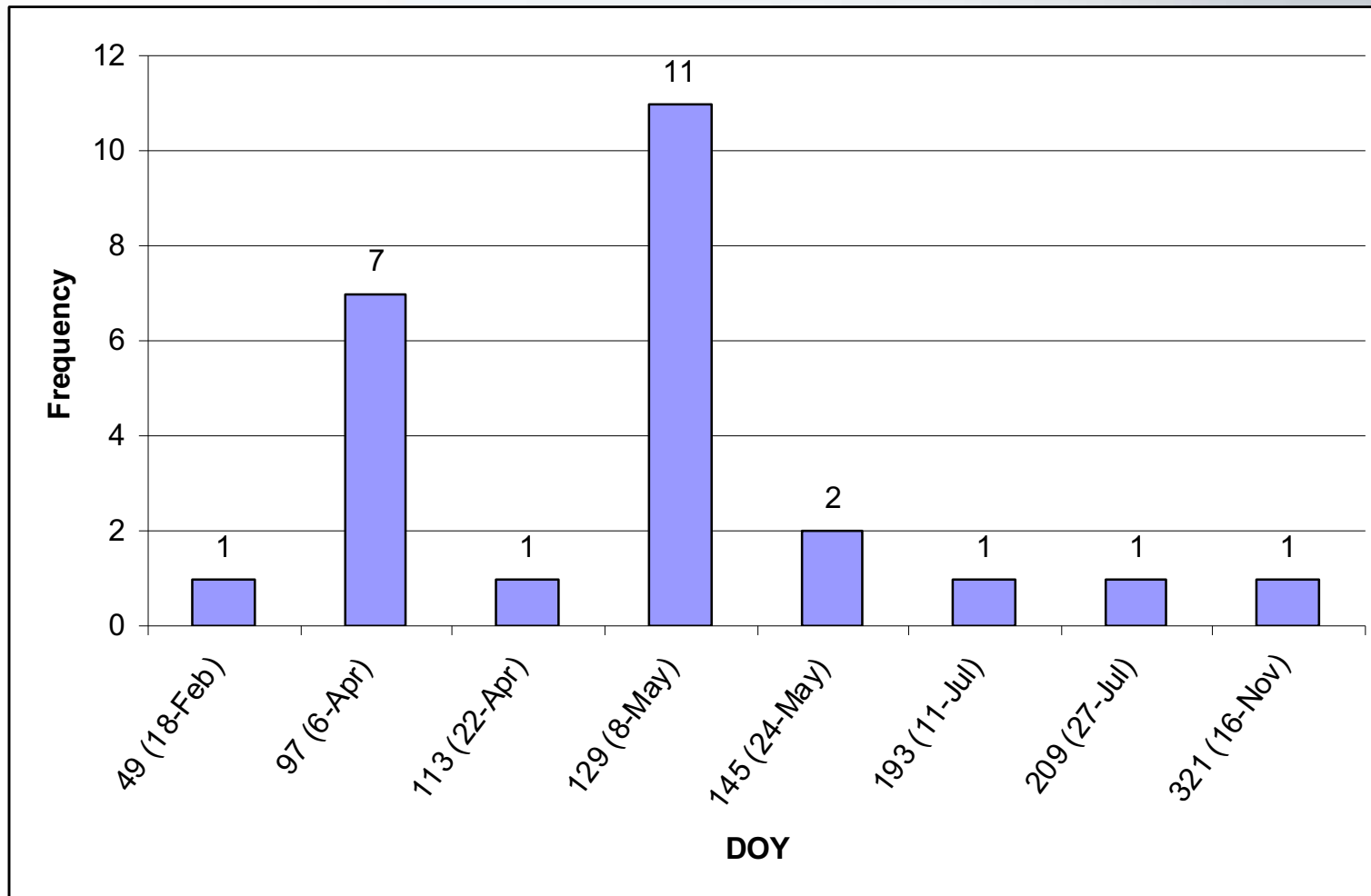
Oblast	DOY	RMSE	Eff
Volyn Oblast	209	3.274	1.018
Zhytomyr Oblast	97	2.898	1.598
Zakarpattia Oblast	321	2.694	1.716
Ivano-Frankivsk Oblast	193	2.771	1.22
Lviv Oblast	113	2.486	0.993
Rivne Oblast	97	3.411	1.214
Chernihiv Oblast	97	3.366	1.267
Vinnitsia Oblast	97	5.405	1.114
Kiev Oblast	97	4.083	1.616
Poltava Oblast	129	4.286	2.09
Sumy Oblast	145	3.758	1.766
Ternopil Oblast	97	3.914	1.214
Kharkiv Oblast	129	3.846	2.443
Khmelnytskyi Oblast	49	3.868	1.421
Cherkasy Oblast	129	6.473	1.35
Chernihiv Oblast	97	3.366	1.267
Dnipropetrovsk Oblast	145	5.302	2.048
Donetsk Oblast	129	4.41	1.871
Zaporizhia Oblast	129	3.797	1.947
Kirovohrad Oblast	129	4.506	2.324
Luhansk Oblast	129	4.189	1.829
Mykolaiv Oblast	129	4.086	2.116
Odessa Oblast	129	5.321	1.589
Kherson Oblast	129	3.927	1.796
Crimea	129	1.809	1.424



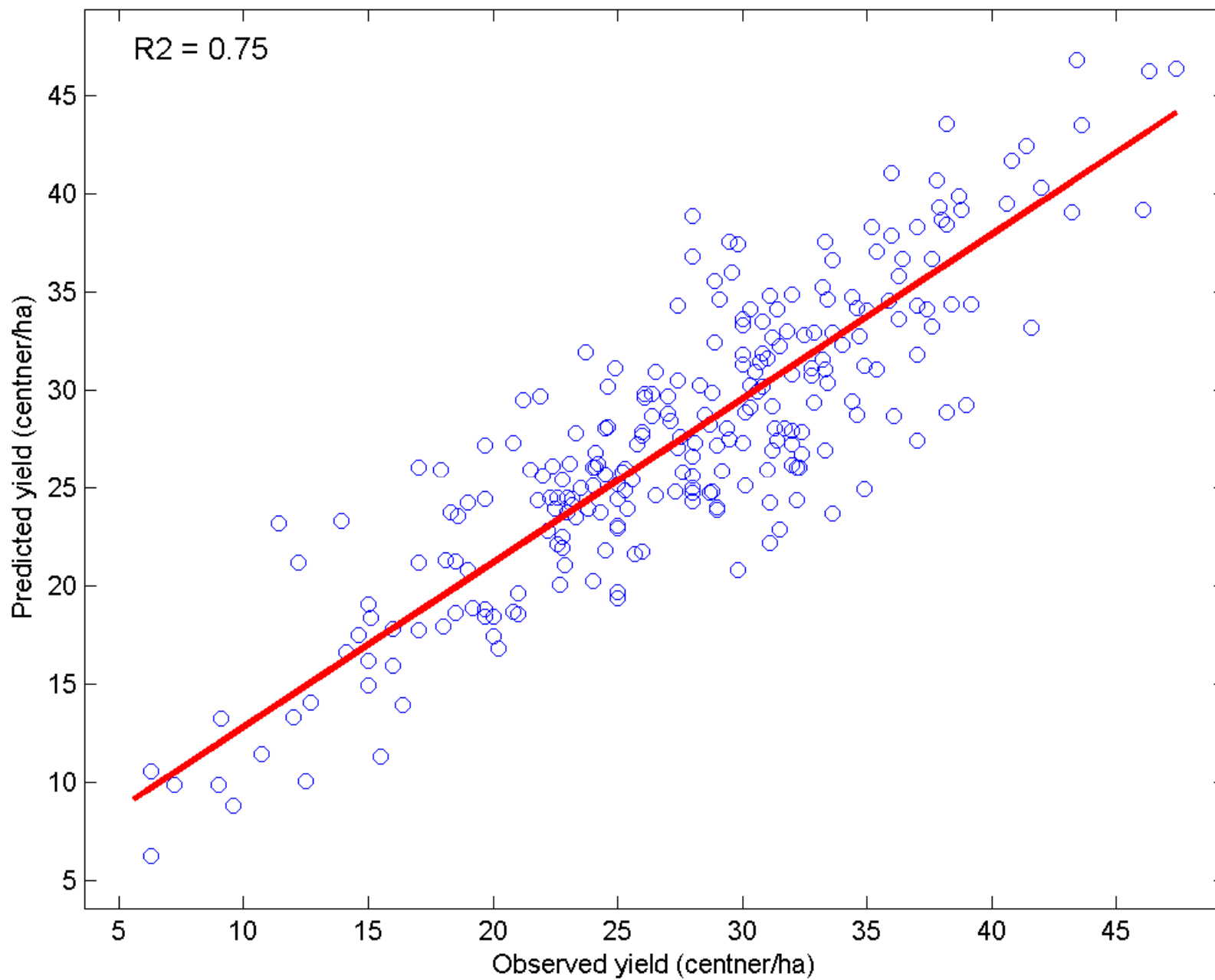
	Zone	Av. eff.
	Plane Polissya	1.289
	Forest-Steppe	1.587
	Steppe	1.883

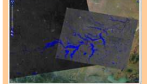
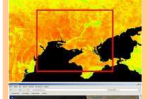
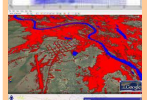
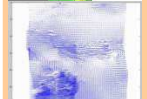
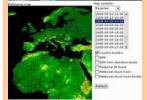
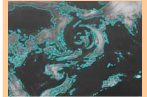
Distribution of DOY

- Distribution of DOYs over oblasts that minimize RMSE on test data



Observed vs predicted yield on test data (for all oblasts)





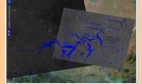
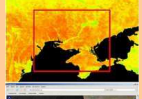
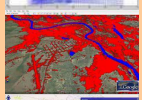
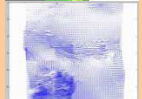
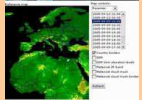
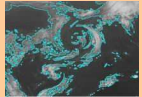
Results

Winter wheat yield estimation using agrometeorological data



Objective

- To model the relationship between crop productivity (in particular winter wheat) and main climatic parameters
 - Maximum temperature
 - Minimum temperature
 - Average temperature
 - Precipitation
 - Soil moisture
 - 0-20 cm depth
 - Available for months: Sept, Oct, Apr, May, June
- Methodology
 - Correlation analysis
 - Linear multivariate regression
 - Non-linear multivariate regression





Correlation between maximum temperature and yield

No	Crim	Dnepr	Don	Zap	Kirov	Lug	Mykol	Odessa	Kherson
7	0.028394	-0.48934	-0.19916	-0.35864	-0.24458	-0.031357	-0.33754	0.14953	-0.45106
8	0.68393	0.15036	-0.21285	-0.01219	0.22505	-0.18567	0.52422	0.47785	0.16628
9	0.12035	-0.31091	0.19388	0.0040819	-0.58689	0.0042976	-0.33512	-0.52793	0.025162
10	0.15613	0.0050704	-0.20699	-0.047397	-0.091975	-0.31676	0.041256	0.070695	0.14046
11	-0.2756	0.3727	0.047867	-0.12432	-0.33289	0.18936	0.019755	-0.11093	0.045587
12	0.097253	0.29823	0.099358	0.13126	0.17035	-0.19041	0.28216	0.16529	0.20881
1	-0.16606	0.63236	0.51598	0.24708	0.25303	0.391	0.35301	0.26233	0.38443
2	0.53168	0.39994	0.54325	0.3951	0.6185	0.40004	0.6659	0.73938	0.51766
3	0.66576	0.62146	0.65599	0.78493	0.51085	0.41348	0.5907	0.85968	0.73434
4	0.59713	-0.016151	-0.13286	0.17282	-0.70793	-0.042479	-0.24879	-0.11057	0.47367
5	-0.31848	-0.64692	-0.50068	-0.62155	-0.70971	-0.31977	-0.84414	-0.63519	-0.71678
6	-0.16975	-0.52878	-0.41524	-0.38899	-0.45837	-0.4801	-0.32603	-0.10596	-0.47221



Correlation between minimum temperature and yield

No	Crim	Dnepr	Don	Zap	Kirov	Lug	Mykol	Odessa	Kherson
7	-0.41662	0.2813	0.15103	0.012797	0.21299	0.5339	0.3039	0.24178	-0.042668
8	0.29938	-0.26698	0.22591	0.062019	-0.4164	0.065315	-0.029812	-0.39759	0.0028998
9	-0.23182	0.060864	-0.14665	-0.23218	0.31486	0.070909	-0.13485	-0.35152	-0.020444
10	0.15472	-0.26534	-0.026552	-0.33768	-0.11899	0.11331	-0.13834	-0.26121	-0.25858
11	-0.32862	0.12824	0.084218	-0.1159	0.12995	0.016175	0.21904	0.17784	0.011755
12	0.71624	0.16197	0.1543	0.19255	0.32885	0.0451	0.43278	0.25153	0.36252
1	-0.17515	0.56079	0.2928	0.13224	0.38556	0.55532	0.20403	0.13767	0.096442
2	0.083637	0.27467	0.26774	0.10612	0.54175	0.30203	0.54004	0.36172	0.36274
3	0.14942	0.30964	0.50986	0.16943	-0.015934	0.3601	0.047862	0.19136	-0.16842
4	0.35385	-0.13256	-0.24131	-0.019246	-0.23256	-0.27213	-0.17588	-0.096503	-0.0088771
5	-0.35995	0.12722	0.15198	0.036682	-0.039337	0.36516	-0.068365	0.069377	-0.47783
6	-0.20339	-0.066609	0.14397	0.16411	-0.45289	0.055353	-0.30575	-0.74256	-0.1294



Correlation between average temperature and yield

No	Crim	Dnepr	Don	Zap	Kirov	Lug	Mykol	Odessa	Kherson
7	-0.29942	-0.35126	-0.24725	-0.29955	-0.35319	-0.14199	-0.34068	-0.29236	-0.35027
8	0.34911	0.1372	0.11163	0.13711	0.25645	0.13735	0.35313	0.53017	0.21105
9	-0.20768	-0.54585	-0.5522	-0.45706	-0.58285	-0.31288	-0.6098	-0.5268	-0.47981
10	-0.11309	0.033319	-0.11353	-0.059982	0.24779	-0.28074	0.11742	-0.018076	-0.0026652
11	-0.56186	0.026467	-0.085765	-0.10321	0.029687	-0.07696	0.050242	-0.15798	-0.25548
12	0.73051	0.32459	0.25071	0.38424	0.32825	0.015717	0.30901	0.30539	0.4968
1	-0.25626	0.17545	0.2493	0.06493	0.14072	0.30679	-0.017265	-0.019324	-0.046139
2	0.41033	0.48671	0.49646	0.41528	0.55871	0.3154	0.6022	0.63751	0.53412
3	0.69726	0.51643	0.58657	0.54978	0.47441	0.35182	0.5327	0.69129	0.64766
4	0.81435	0.37751	0.31695	0.4645	0.36667	0.2552	0.37446	0.49336	0.60399
5	-0.79492	-0.68007	-0.44278	-0.71048	-0.78131	-0.14451	-0.87952	-0.87296	-0.84323
6	-0.2975	-0.24814	-0.17757	-0.33558	-0.42477	-0.34793	-0.44075	-0.5314	-0.39467



Correlation between precipitation and yield

No	Crim	Dnepr	Don	Zap	Kirov	Lug	Mykol	Odess a	Kherson
7	-0.44383	-0.05185	-0.17968	-0.076432	0.32844	-0.062767	0.35819	0.094325	0.29632
8	-0.33229	-0.071638	0.28205	0.12939	-0.26735	-0.035257	-0.31541	-0.50435	-0.078802
9	-0.59345	-0.001362	-0.037385	-0.32289	0.12356	0.40803	0.35673	0.33836	-0.09701
10	-0.44322	-0.3653	-0.7031	-0.53118	-0.27822	-0.5492	-0.1809	-0.17297	-0.36161
11	0.12658	-0.2417	-0.32216	-0.3028	0.19192	-0.27842	-0.093623	0.17322	-0.16728
12	0.45188	0.1798	-0.22143	0.21909	-0.0036298	-0.19411	0.10072	0.19536	0.22254
1	-0.28507	-0.48694	-0.091946	-0.066088	-0.37517	-0.24441	-0.2021	-0.26254	-0.22865
2	-0.20136	0.28383	0.052988	0.13561	0.033568	-0.020791	-0.071412	-0.45491	-0.22081
3	0.32626	0.15665	-0.16078	-0.090452	0.20544	-0.057225	0.40517	0.4879	0.20706
4	0.28161	0.40718	0.44374	0.34788	0.19804	0.53928	0.0048345	0.034649	0.49343
5	0.4525	0.42638	0.41399	0.48975	0.75568	0.45475	0.69528	0.50943	0.52469
6	0.5692	-0.073222	-0.08988	0.31157	0.38893	-0.18582	0.70188	0.40401	0.34927



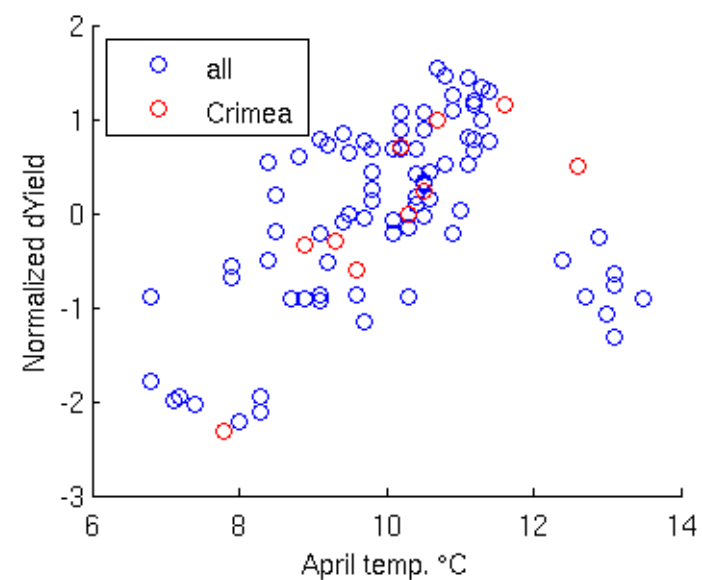
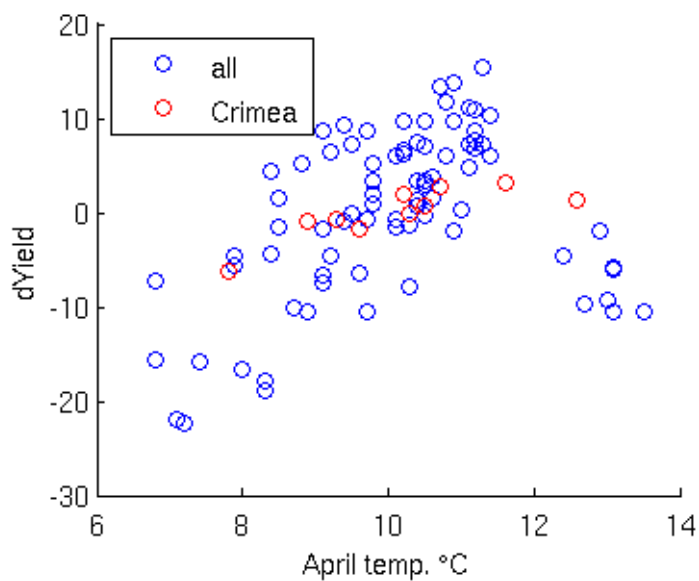
Correlation between soil moisture and yield

No	Crim	Dnepr	Don	Zap	Kirov	Lug	Mykol	Odessa	Kherson
4	-0.40062	-0.14184	0.19829	0.033671	-0.034916	0.48005	0.11103	-0.012341	0.0035585
5	-0.5799	-0.2638	-0.4639	-0.06519	-0.67087	-0.45136	0.3498	-0.26371	0.23095
6	0.35654	0.27268	0.36986	0.52899	-0.093773	0.18641	0.22475	0.0076216	0.39847
9	0.52945	0.3526	0.26894	0.31585	0.49607	0.44526	0.7434	0.41794	0.61782
10	0.74869	0.28547	-0.052434	0.22831	0.24574	-0.13895	0.68106	0.27564	0.41003

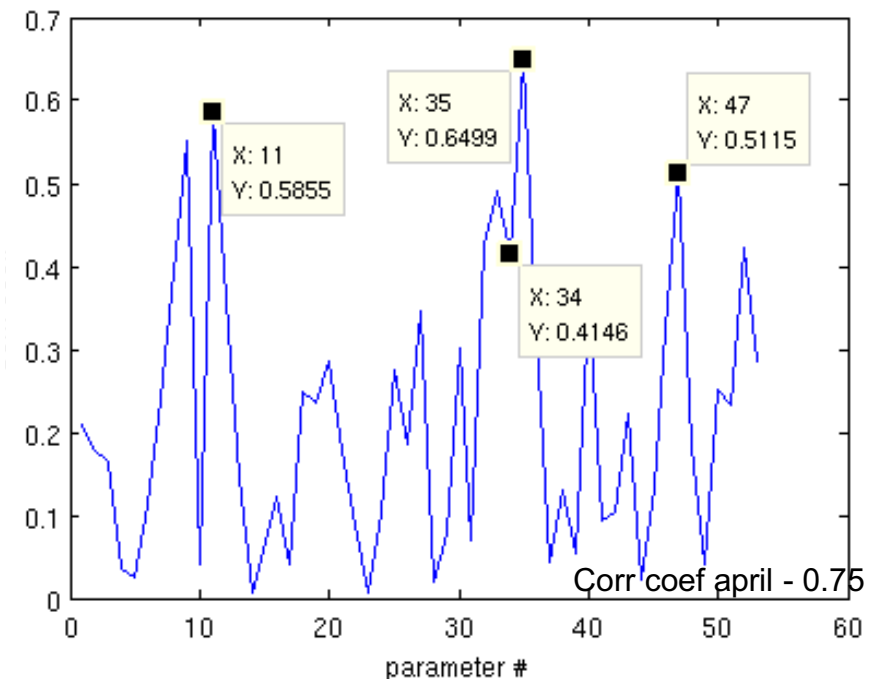
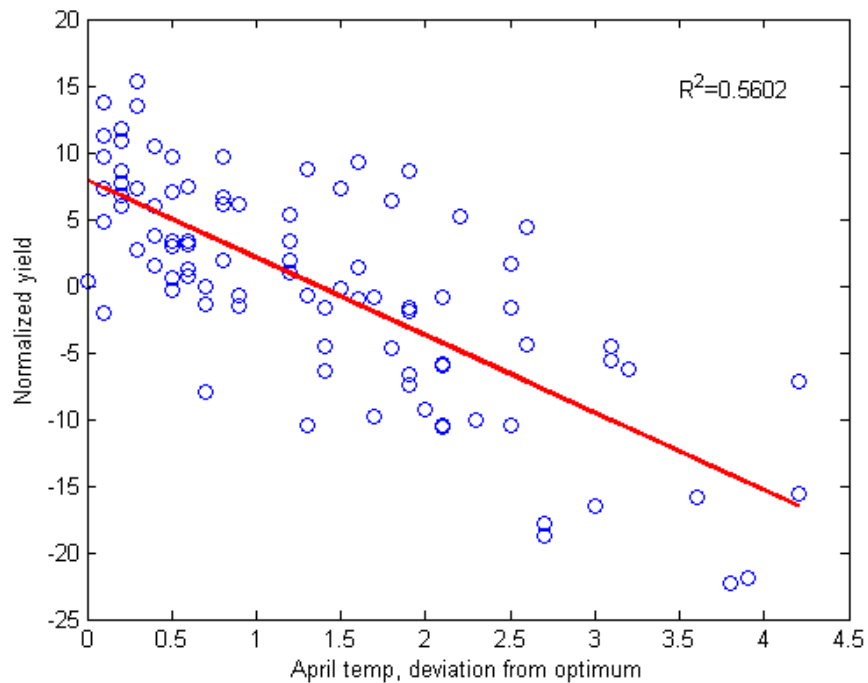
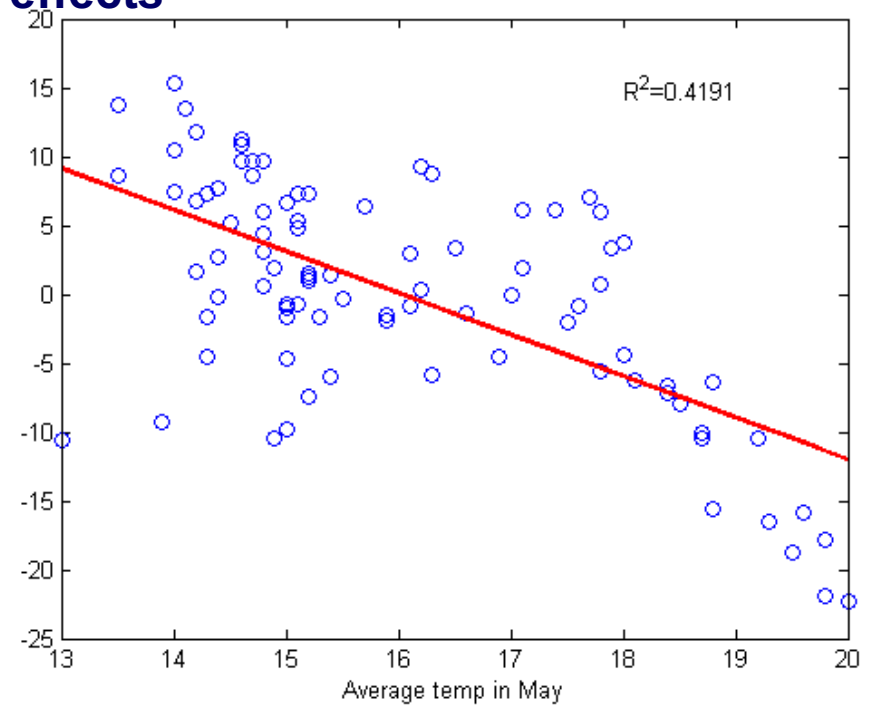
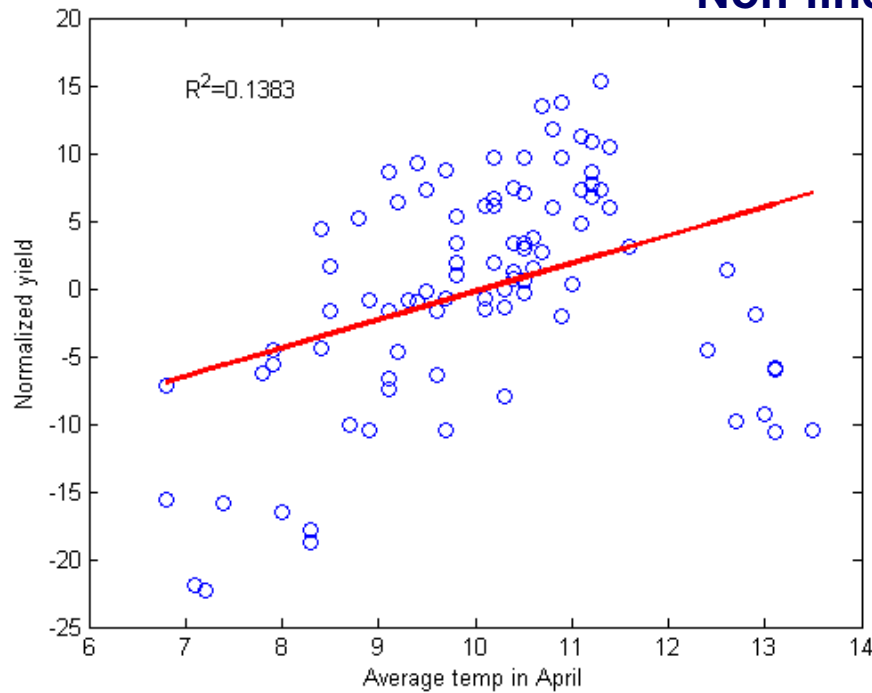


Regions normalization

- Split by agroclimatic zones
- Detrend
- Normalize by std

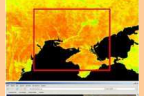
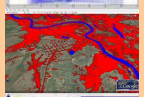
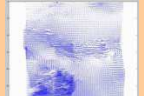
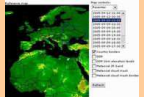
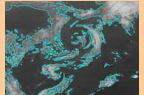


Non-linear effects

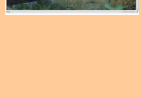
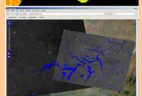
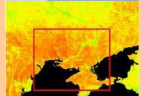
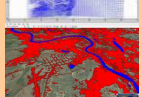
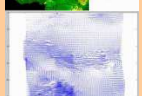
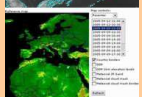
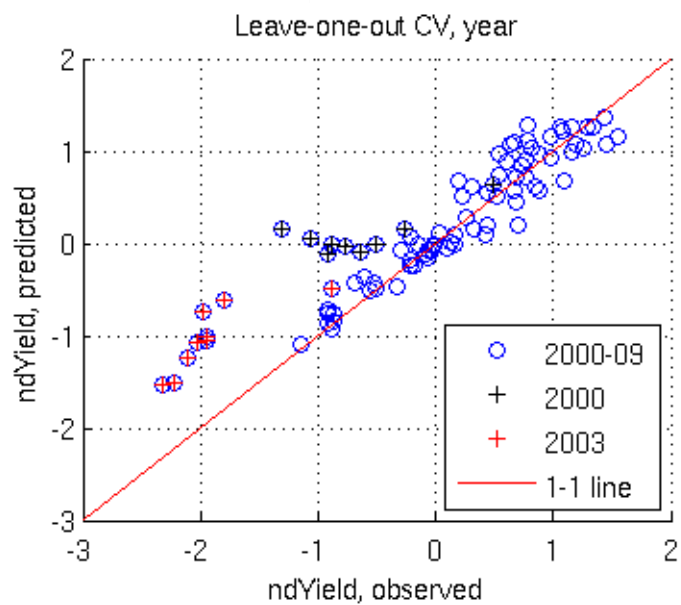
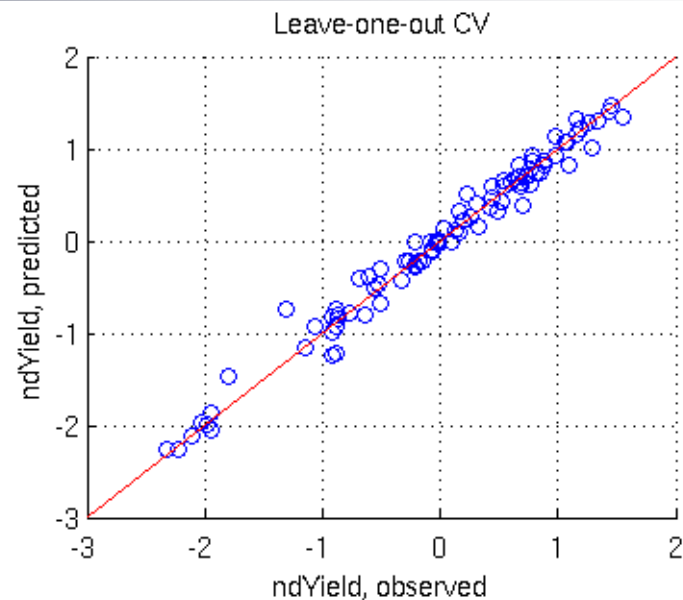
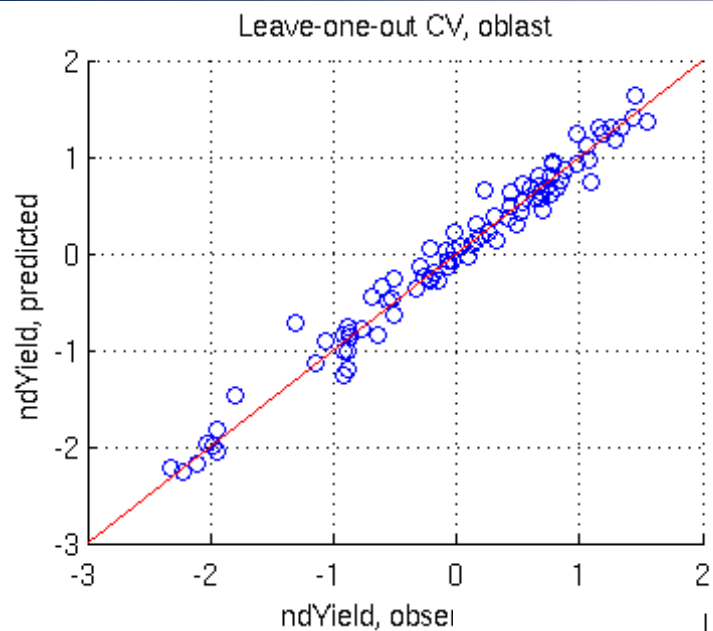


Nonlinear regression

- Neural Networks
- Kernel methods
 - Support Vector Machines
 - Gaussian processes regression



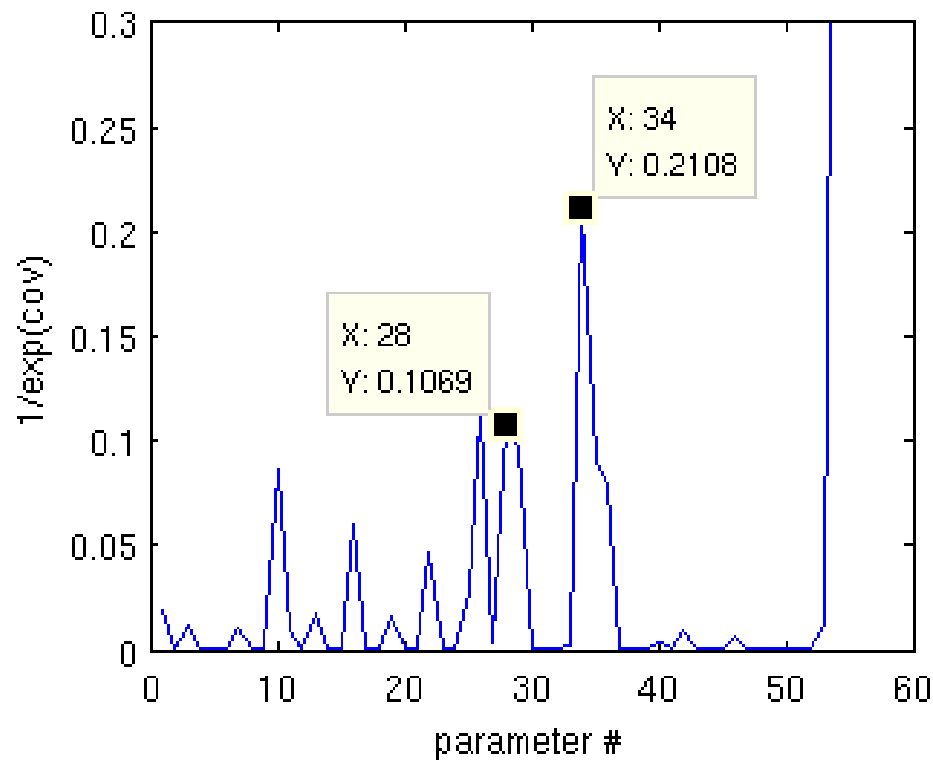
Gaussian processes regression



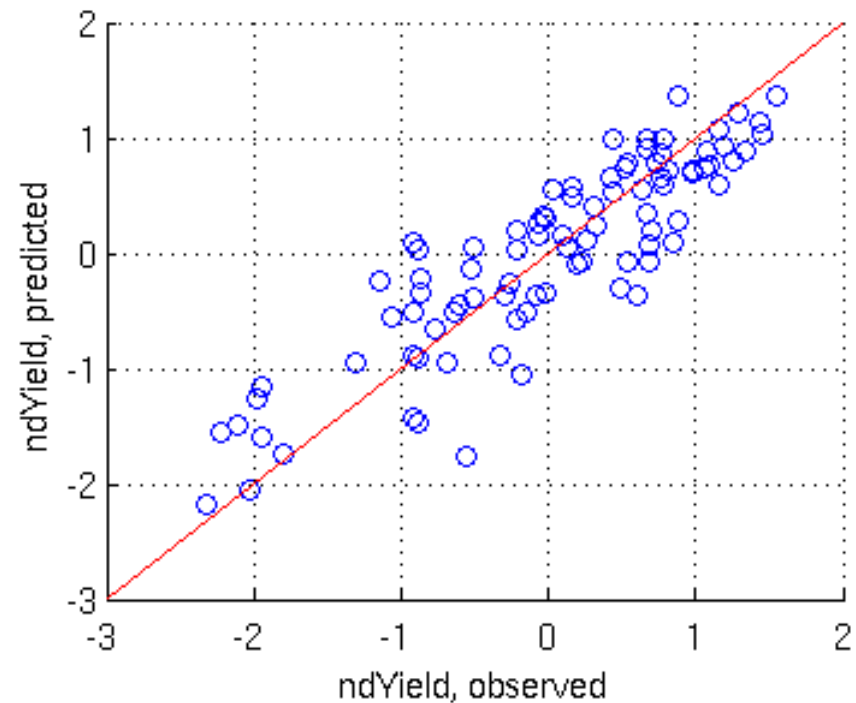
Feature selection

- Most relevant predictors
 - Av temp (August, October, April)

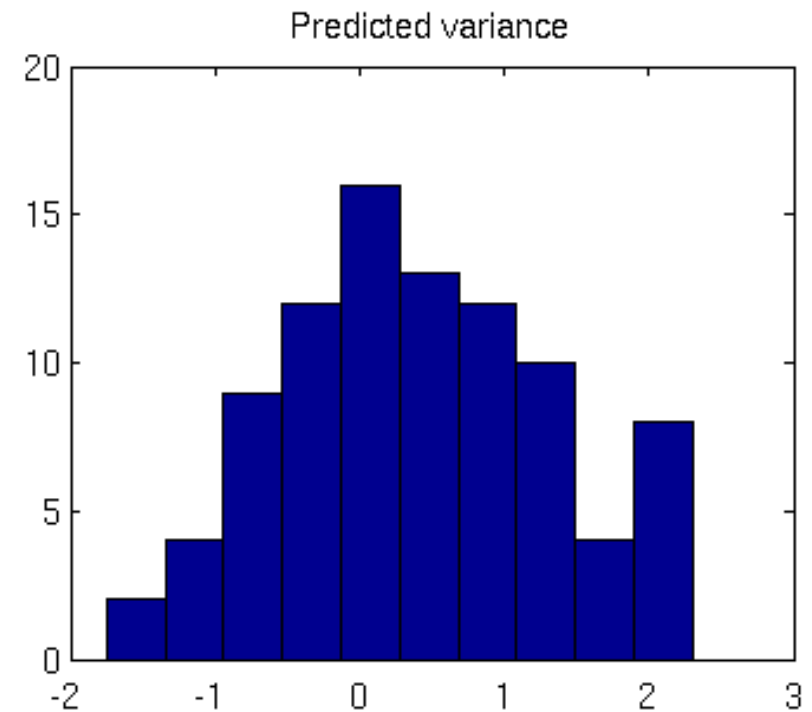
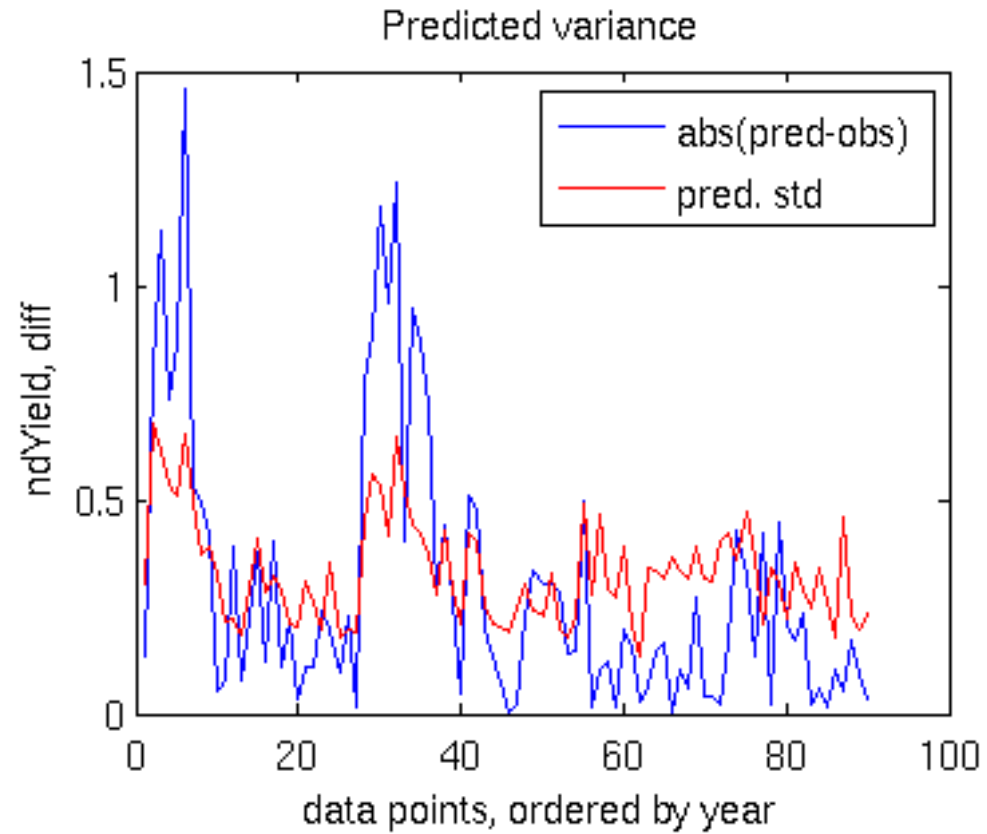
Parameter relevance



Leave-one-out CV. Set1

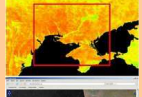
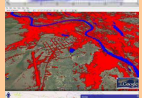
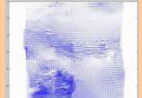
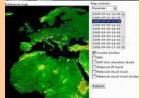
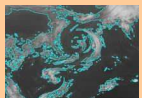


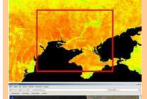
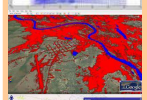
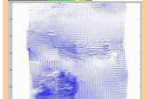
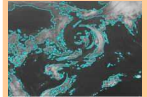
Variance prediction



Efficiency

Oblast	Eff.
Dnipropetrovsk Oblast	2.999
Donetsk Oblast	2.322
Zaporizhia Oblast	1.800
Kirovohrad Oblast	2.469
Luhansk Oblast	1.845
Mykolaiv Oblast	1.855
Odessa Oblast	2.511
Kherson Oblast	2.759
Crimea	3.592





Thank you!

