

itü



Remote sensing and Data Fusion for Digital Twin Applications

B. Berk ÜSTÜNDAĞ

Professor at Istanbul Technical University / Vice President of ISAM



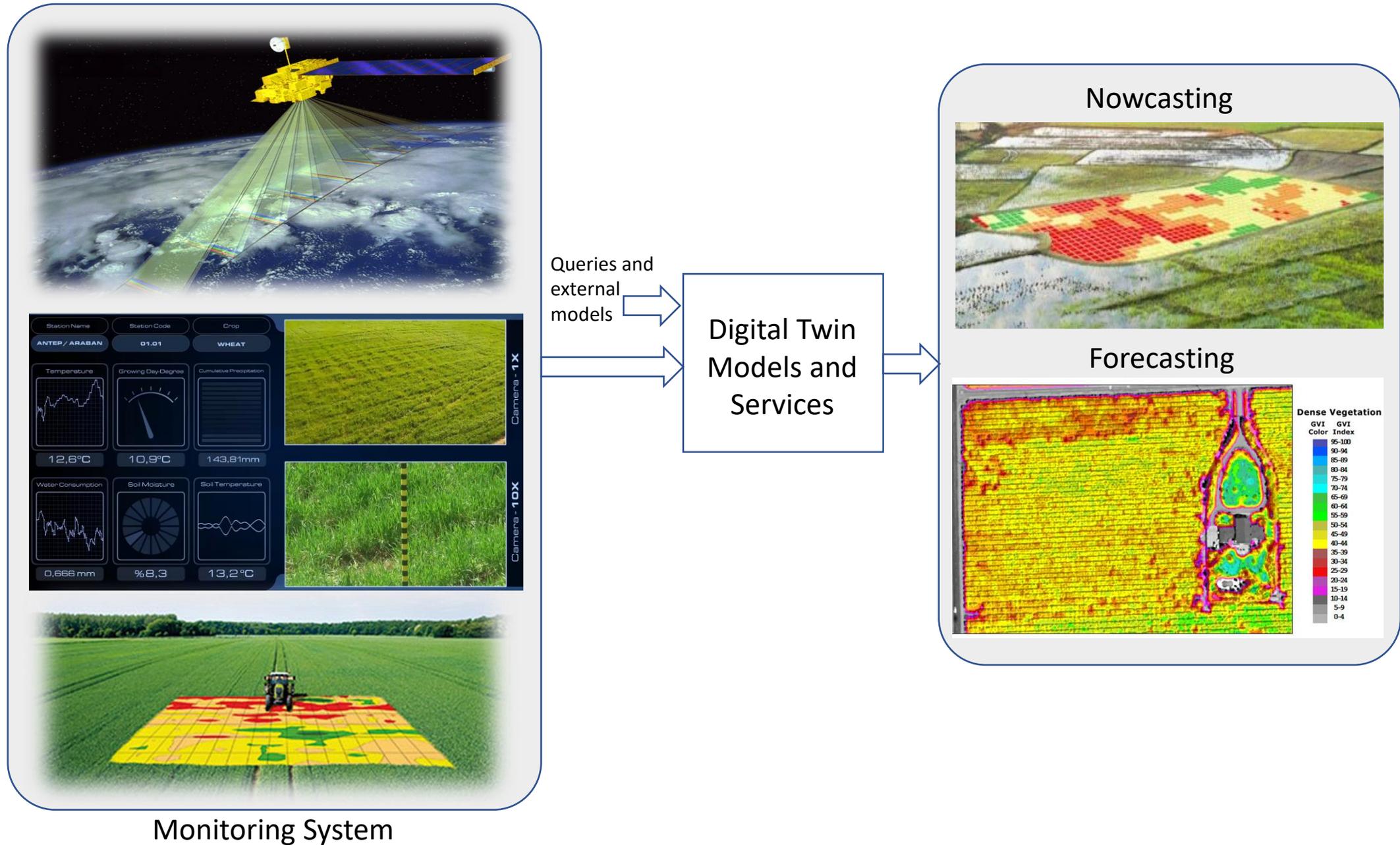
International Society of Agromatics (ISAM)

www.agromatics.org

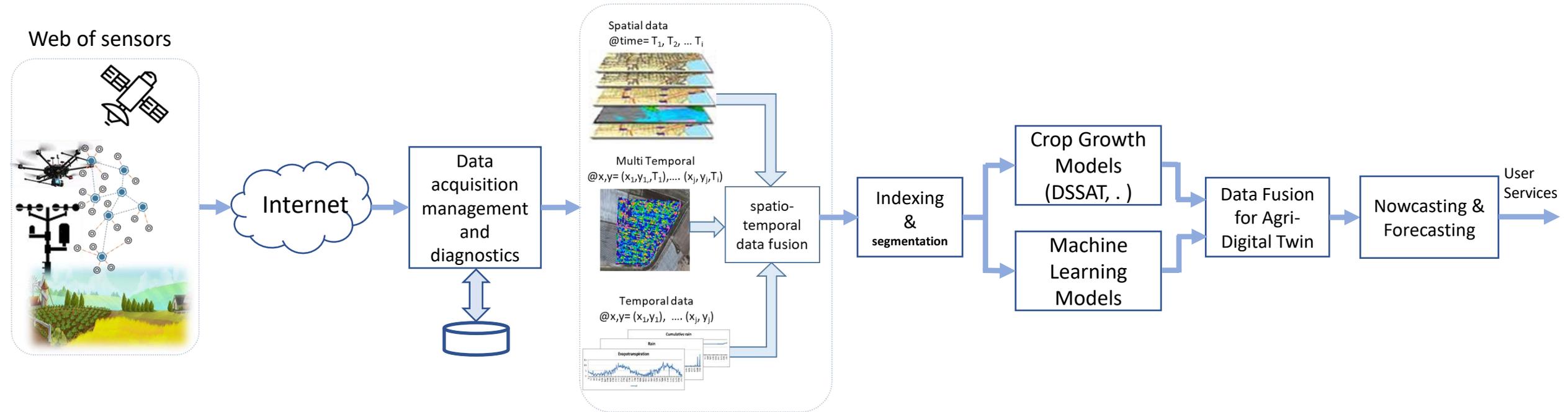
Eurisy Workshop
7th June 2022

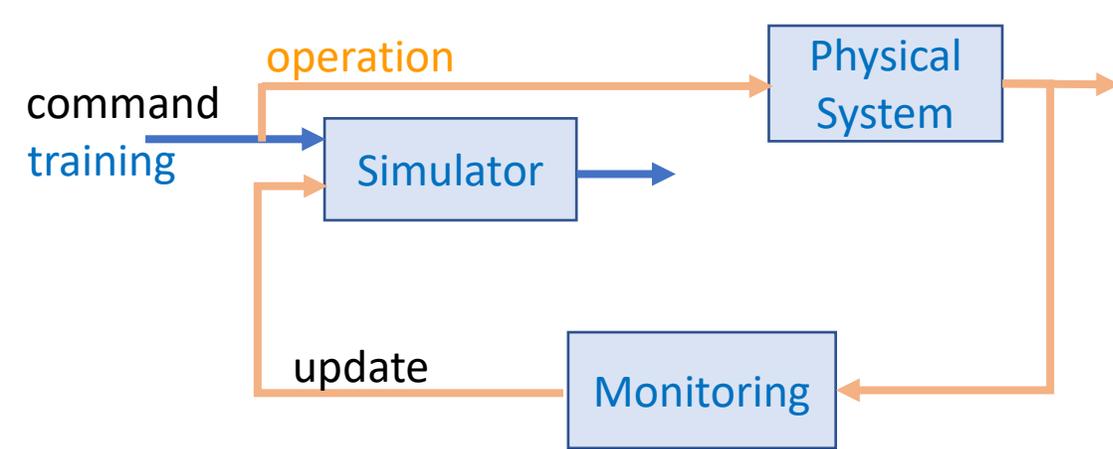


Digital Twinning Concept for Nowcasting and the Forecasting



Digital Twinning Process Block Diagram of the AgriTwin





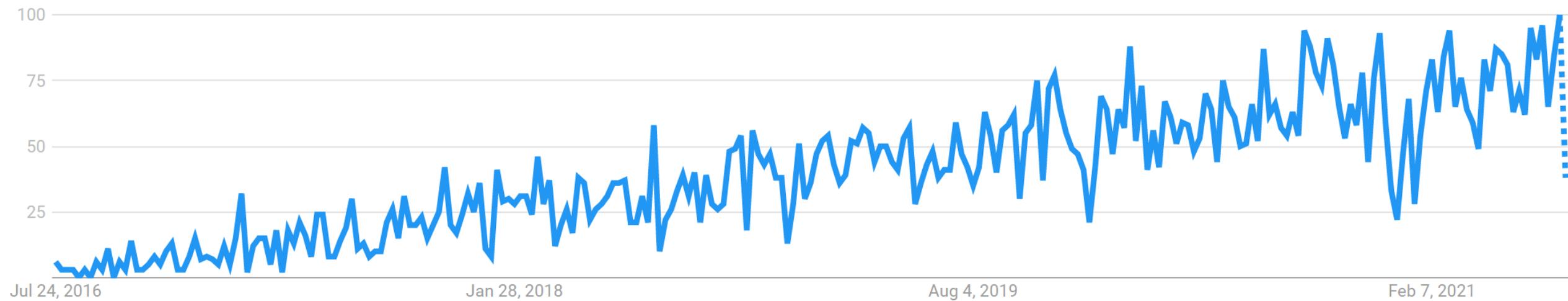
1970



- Simulators that enable system twinning helped to prevent disasters in Apollo 13 mission
- Apollo 13 mission was probably the first use of a system's virtual twin. NASA mission controllers were able to rapidly adapt and modify the simulations, to match conditions on the real-life faulty spacecraft, so that they could research, reject, and perfect the strategies required to bring the astronauts home.

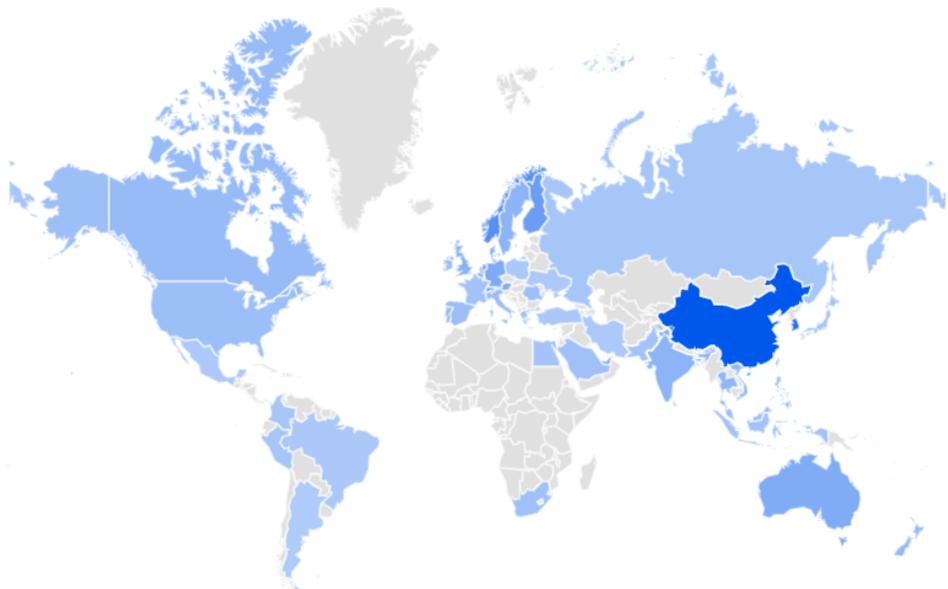


Interest over Time to Digital Twin

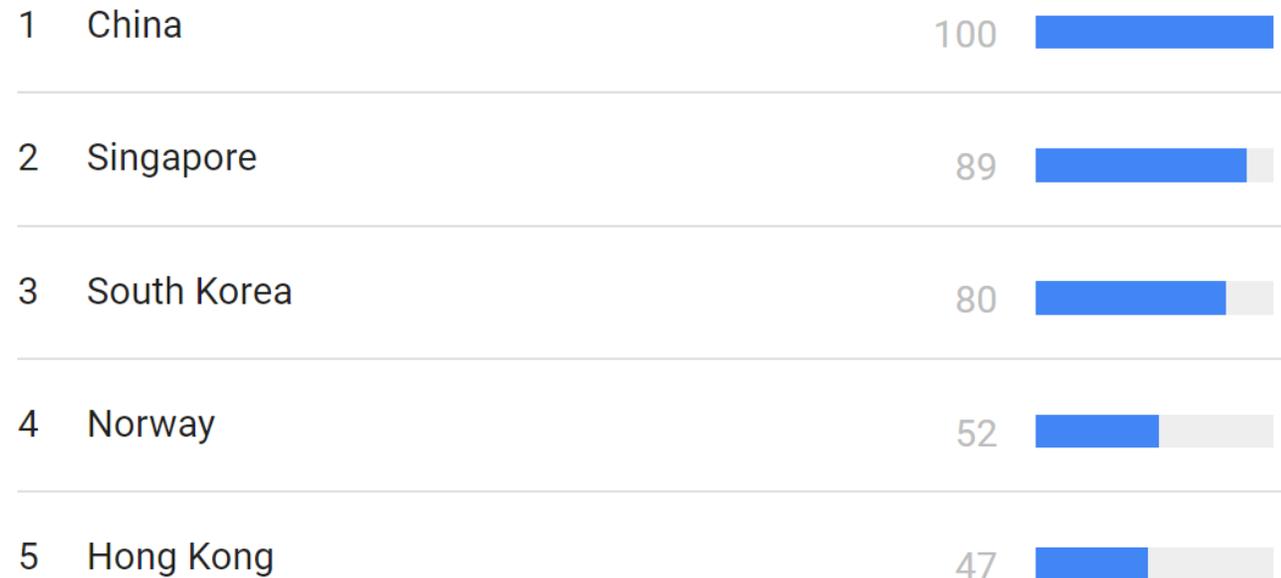


(Google analytics)

Interest by Countries



What is Digital Twin ?



Is “Digital Twinning” new or a magic ?

Formal Definition:

“A digital twin is a virtual representation of a physical system, object, or process

that serves as its real-time digital counterpart

by using simulation, machine learning and reasoning

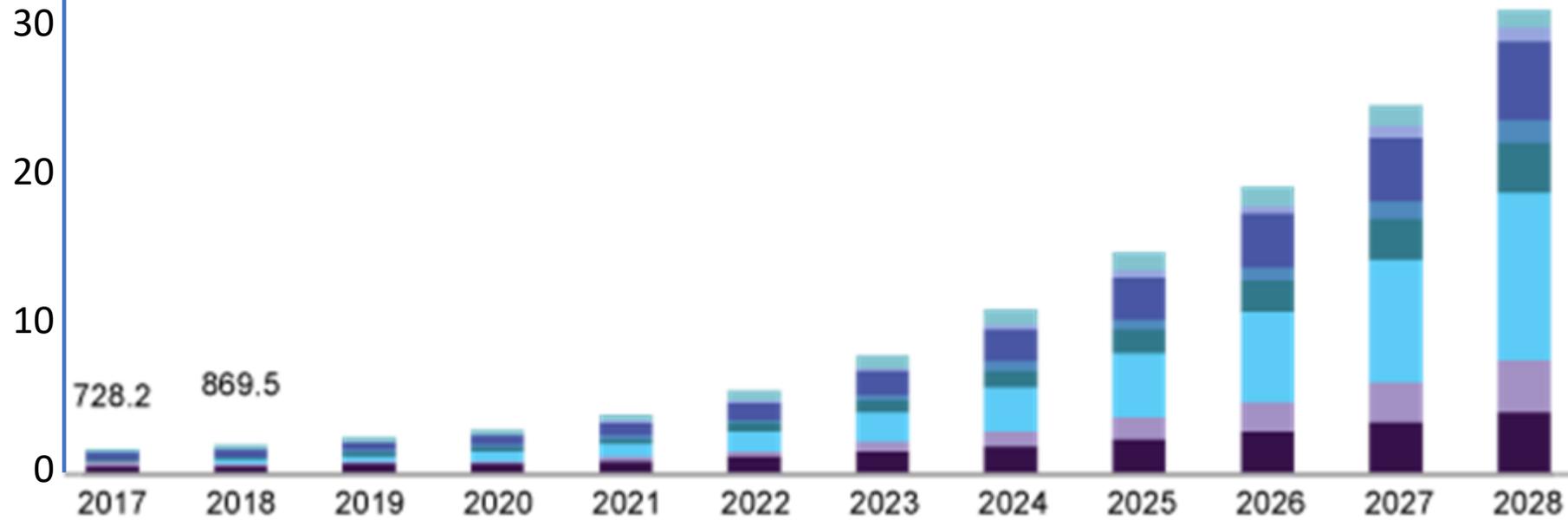
through real-time updating its states

for decision-support purposes”

Definition is new but exists in nature since the evolution of our brains

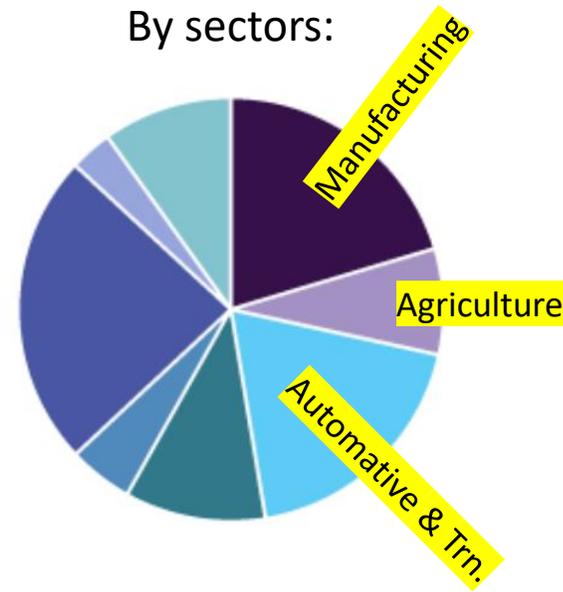
USD (Billion)

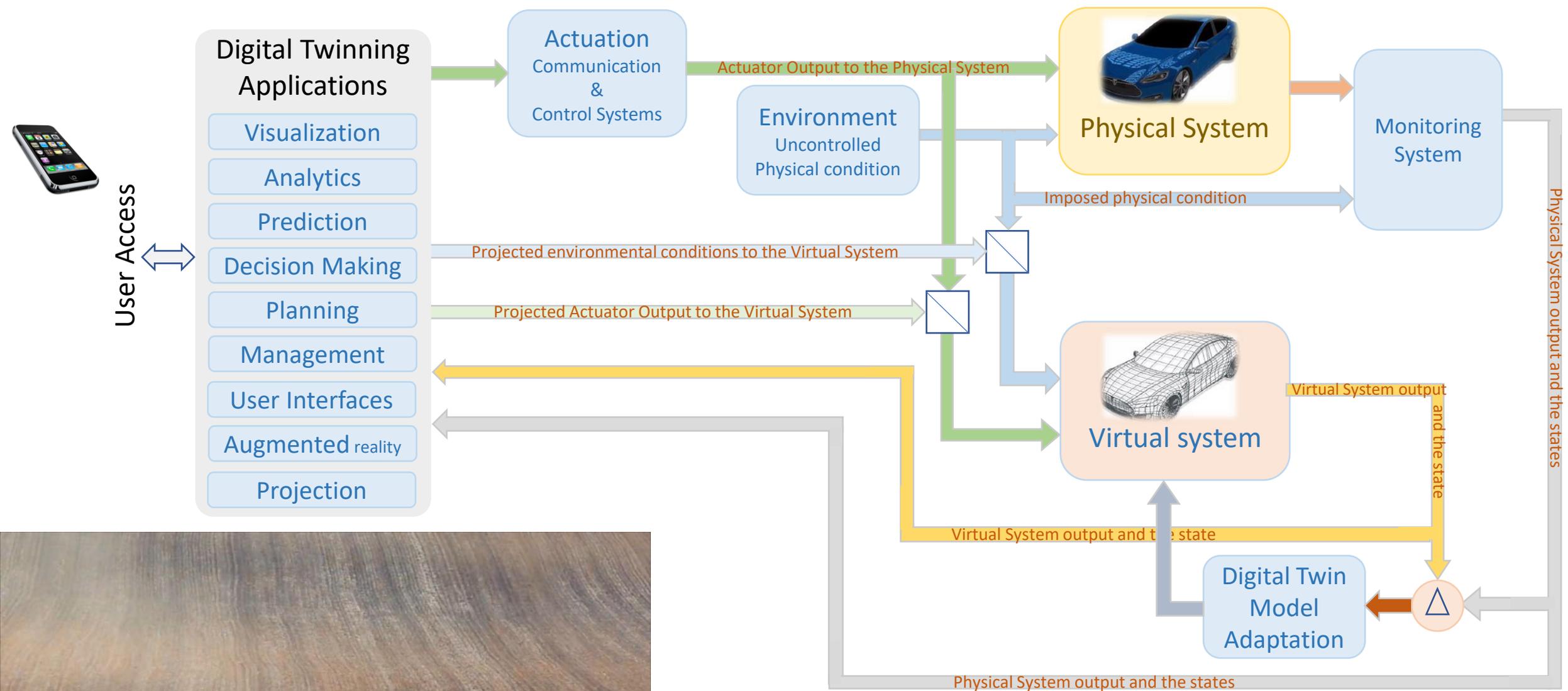
Digital Twin End User Market Forecast in US



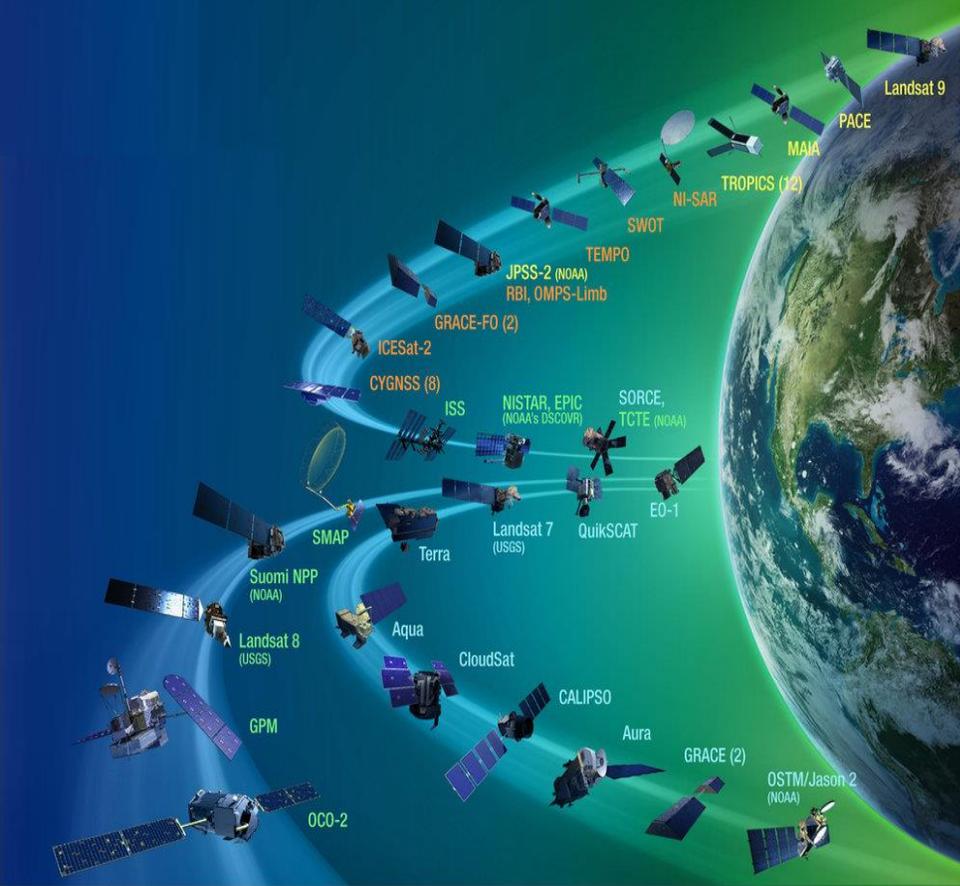
- Agriculture
- Manufacturing
- Automotive & Transport
- Energy & Utilities
- Healthcare & Life Sciences
- Residential & Commercial
- Retail & Consumer Goods
- Others

By sectors:





Digital Twinning Model



Business process requires spatio-temporal data with known accuracy and tolerance.

Data driven models require labeled data in relatively short term that satisfies:

Time varying system

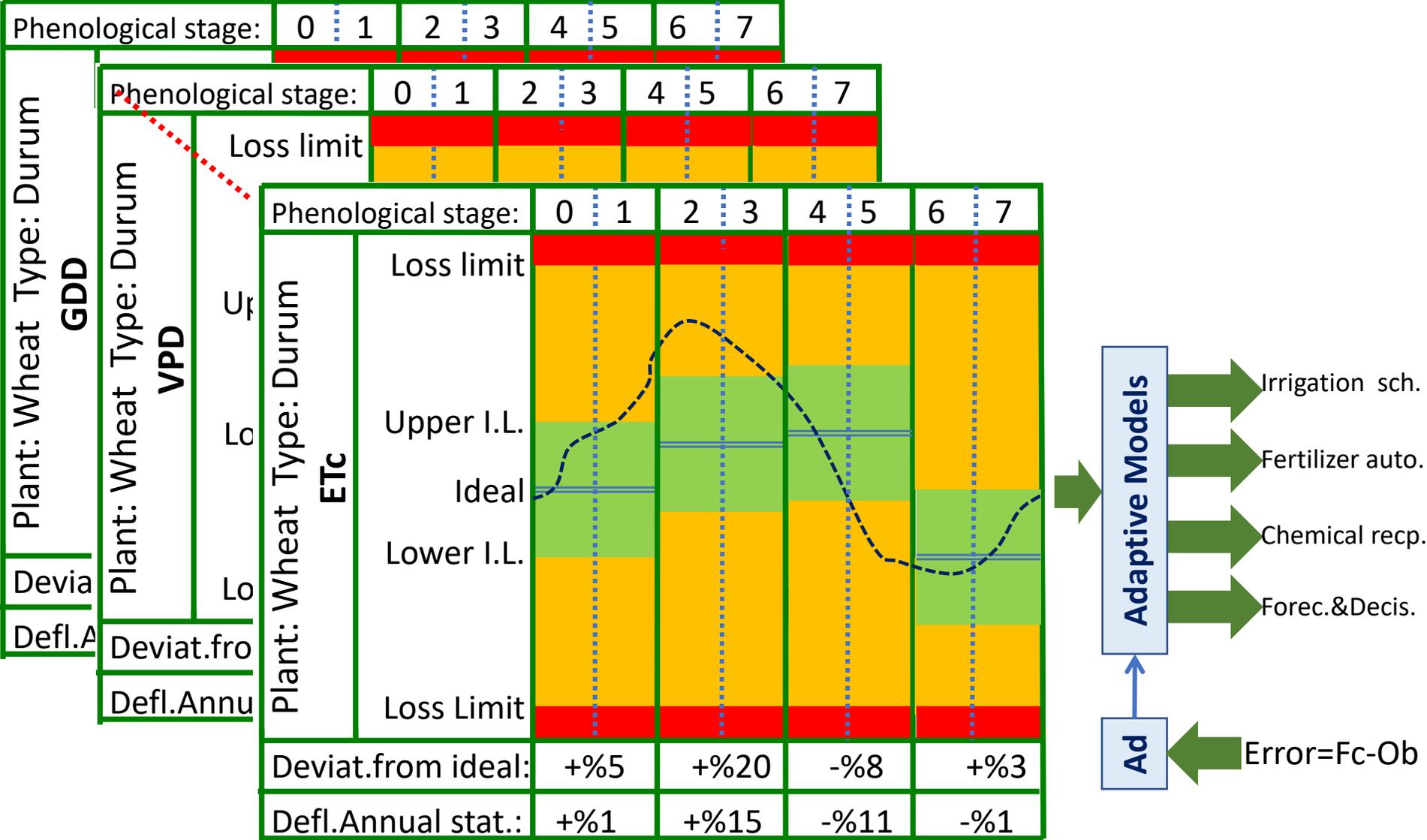
Non-linear

Stochastic

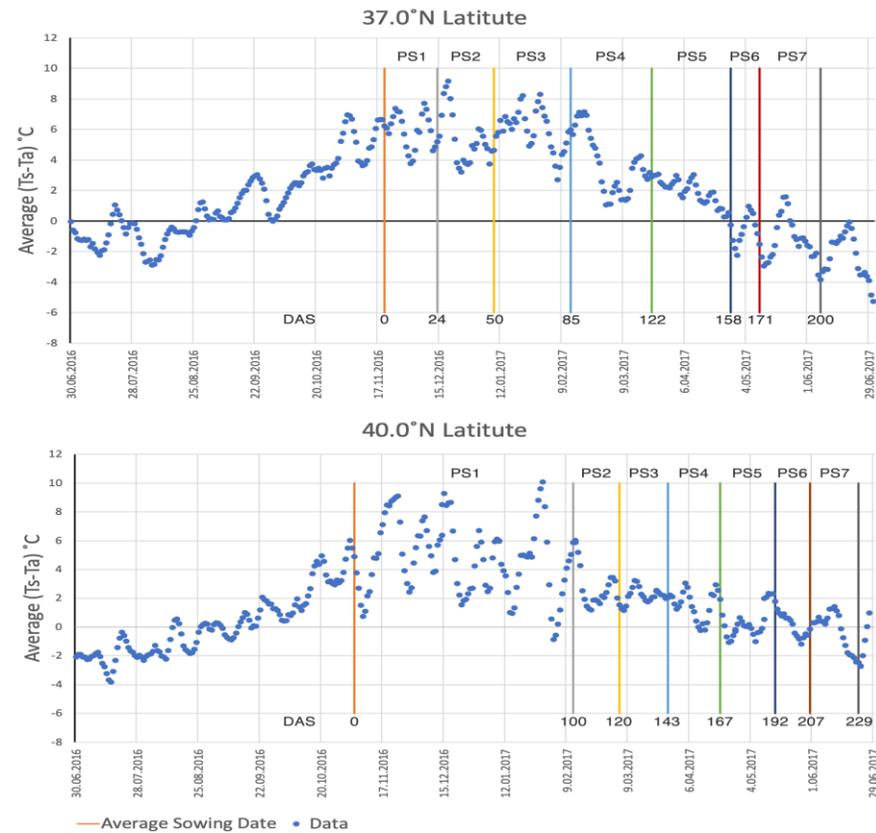
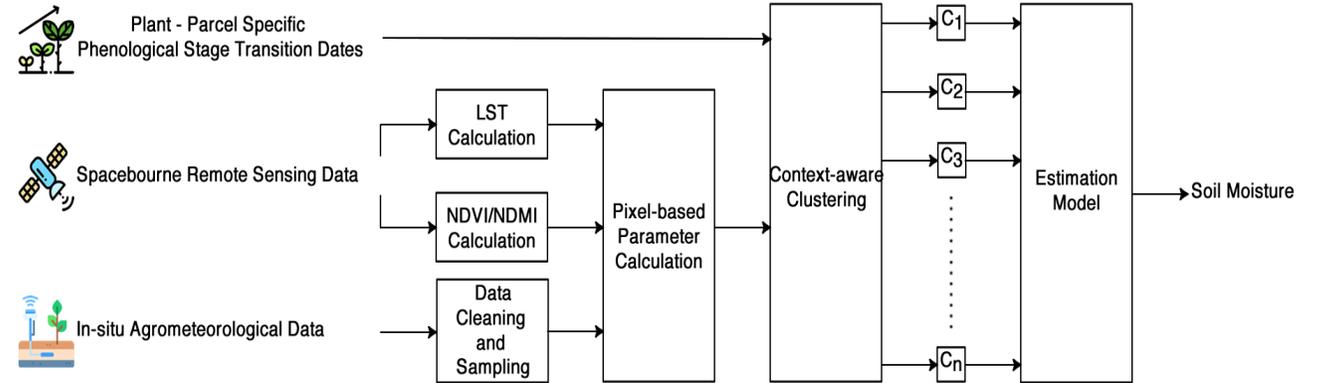
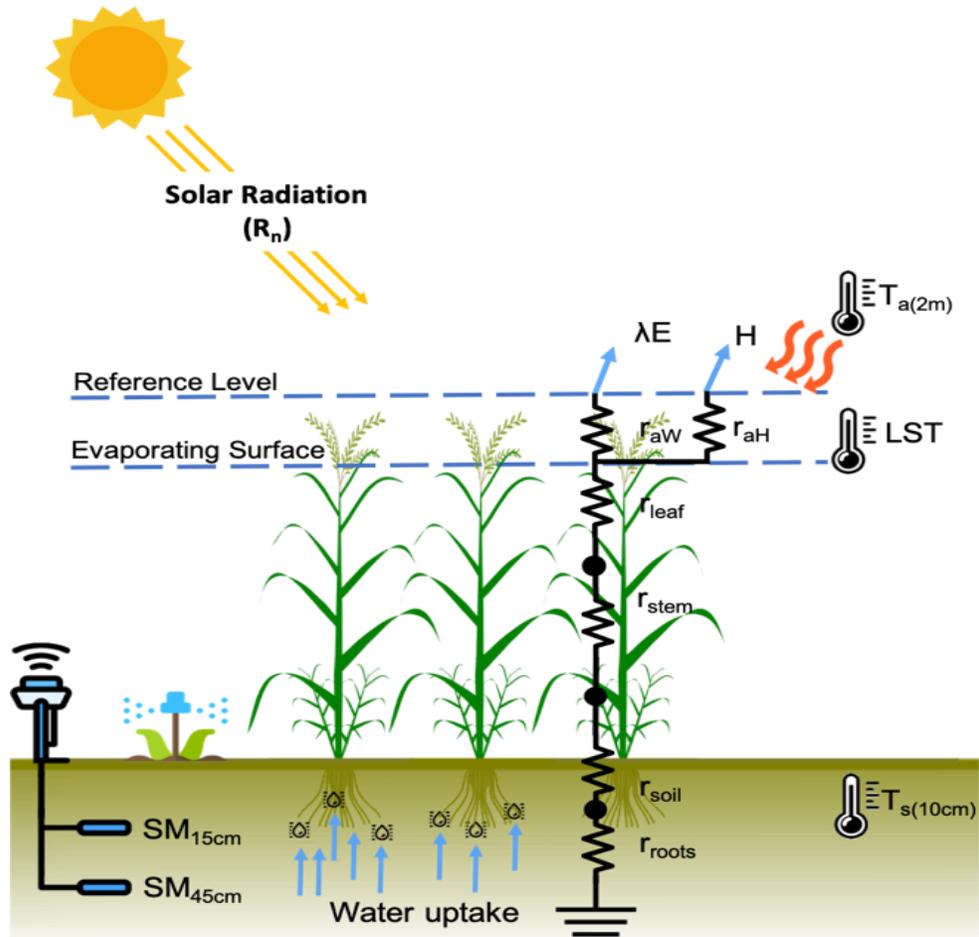
Combinatorial derivatives



Decision Making / Planning Requirements:



1- Context Matters: Plant as a sensor, context (PS, (Ta-LST))



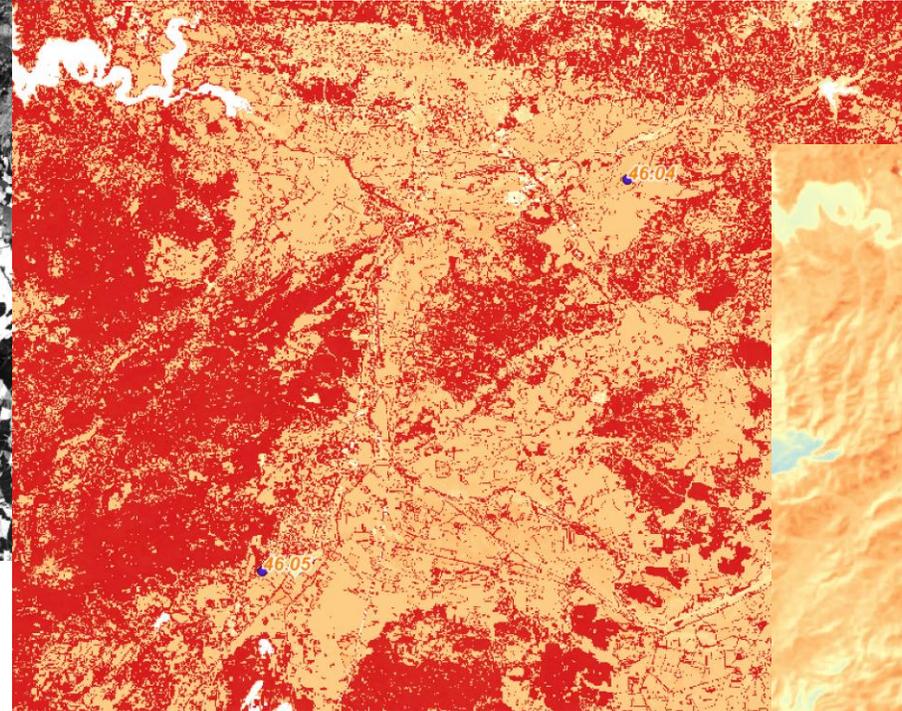
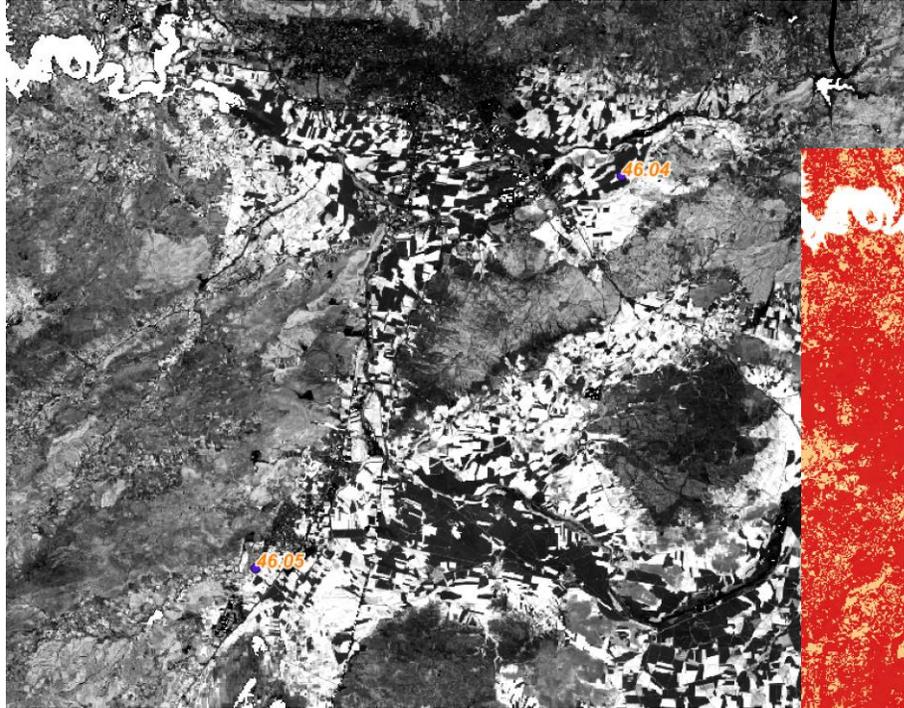
Vegetation Proportion

$$P_v = \left(\frac{NDVI - NDVI_{min}}{NDVI_{max} - NDVI_{min}} \right)^2$$

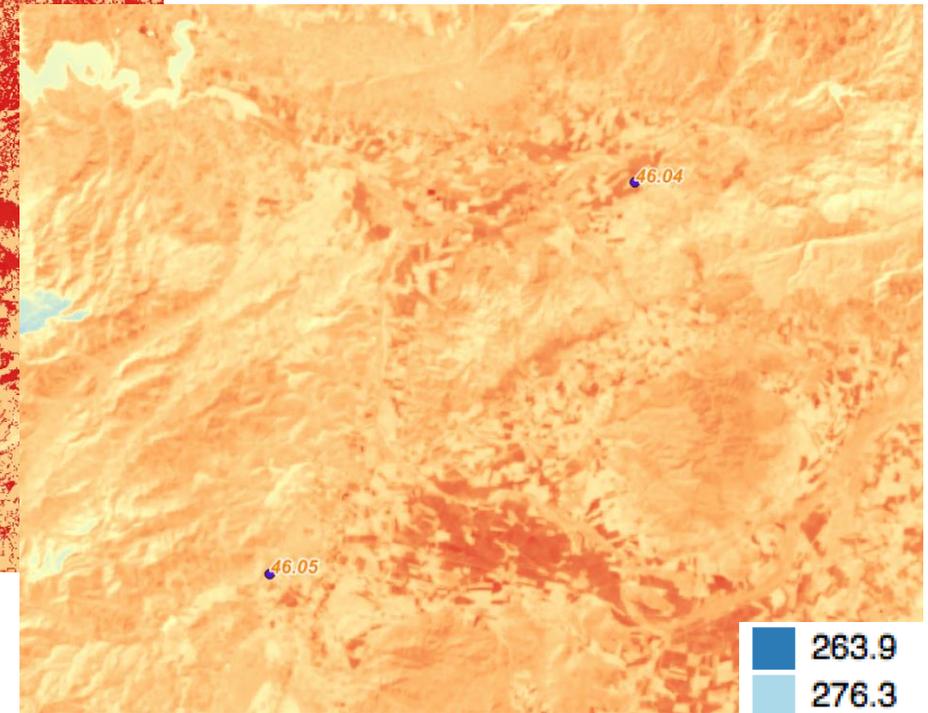
$$f(\epsilon_i) = \begin{cases} a_i \rho_{red} + b_i, & NDVI < 0.2 \\ \epsilon_{v,i} P_v + \epsilon_{s,i} (1 - P_v) + C_i, & 0.2 \leq NDVI \leq 0.5 \\ \epsilon_{v,i} + C_i, & NDVI > 0.5 \end{cases}$$

$$C_i = (1 - \epsilon_{s,i}) \epsilon_{v,i} * F' * (1 - P_v)$$

Emissivity



Brightness Temperature (K)



■ -0.466667
2.03333

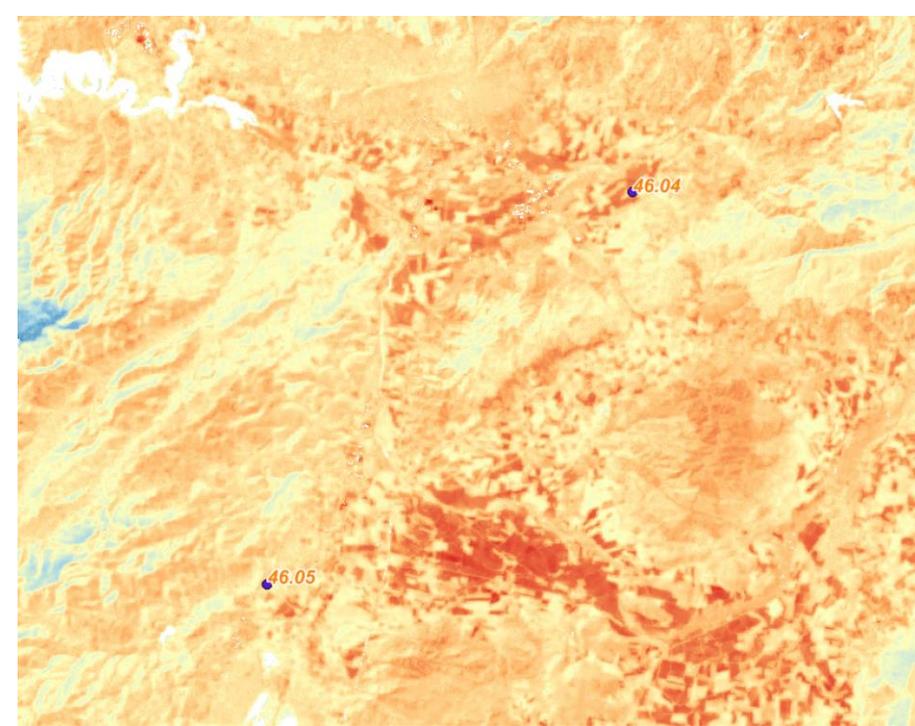
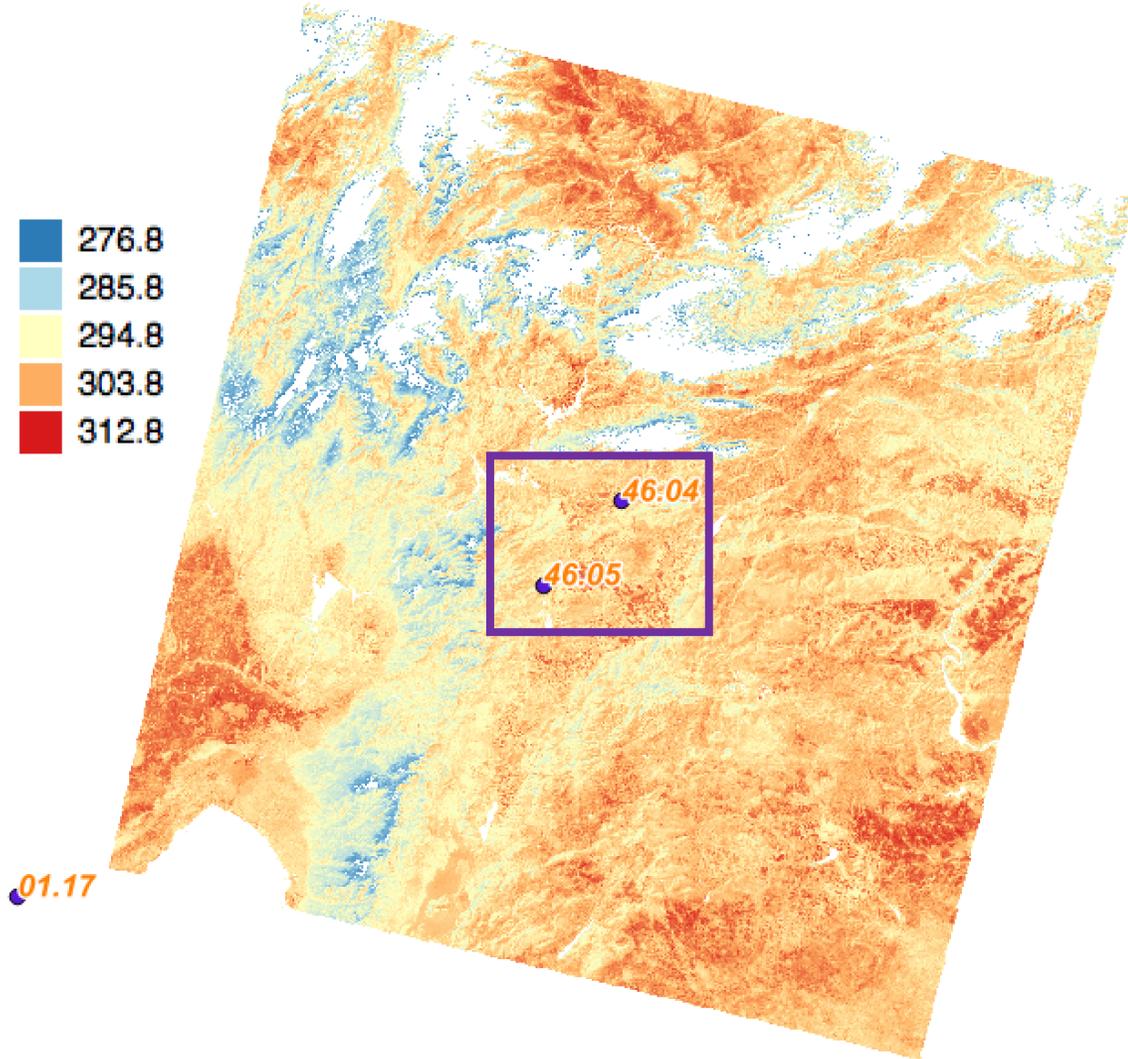
■ 0.9298
■ 0.9439
■ 0.9581
■ 0.9722
■ 0.9863

■ 263.9
■ 276.3
■ 288.6
■ 301
■ 313.4

Land Surface Temperature

$$\frac{T_B}{1 + \left(\lambda * \frac{T_B}{p}\right) * \ln(\varepsilon)}$$

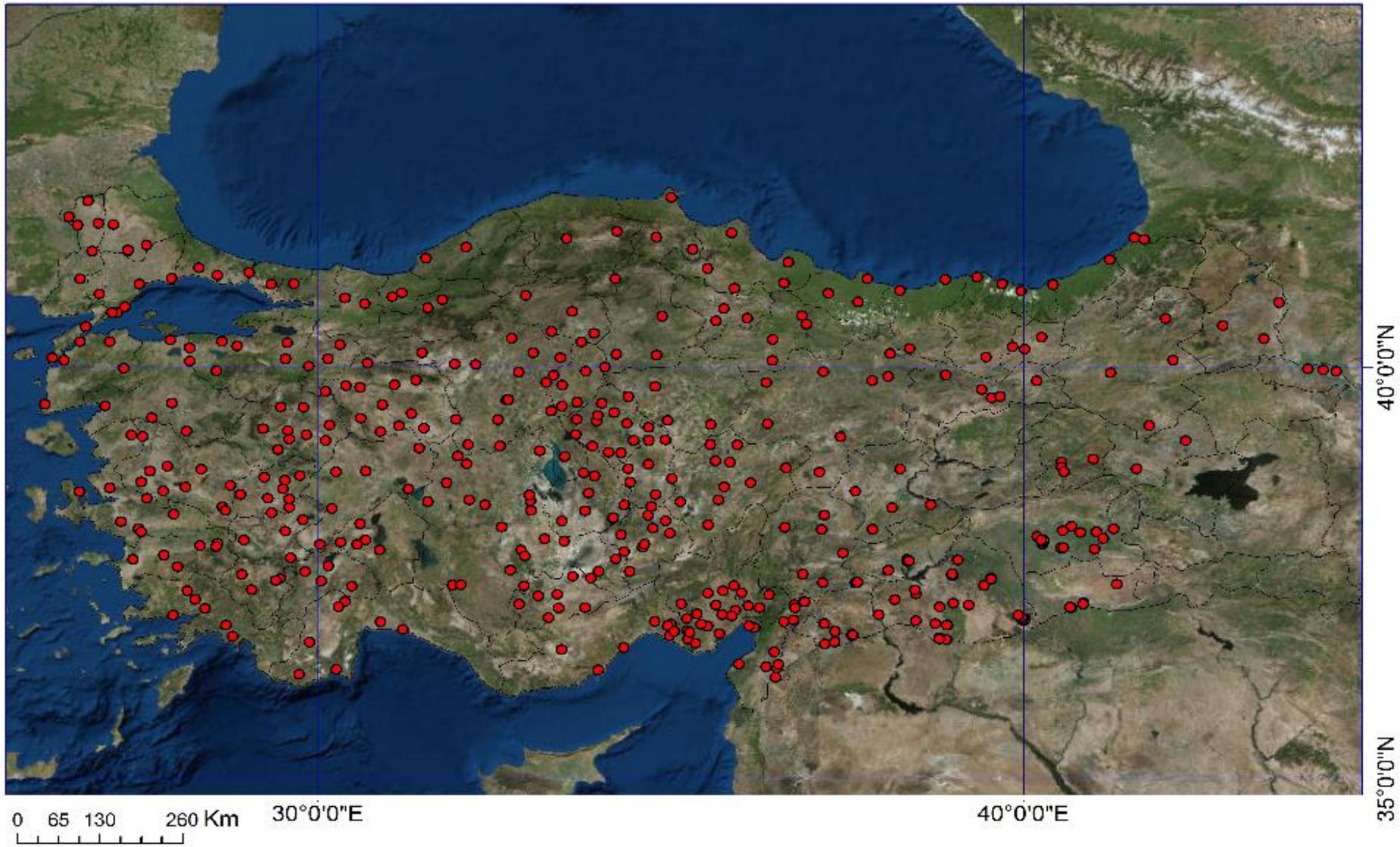
- 276.8
- 285.8
- 294.8
- 303.8
- 312.8



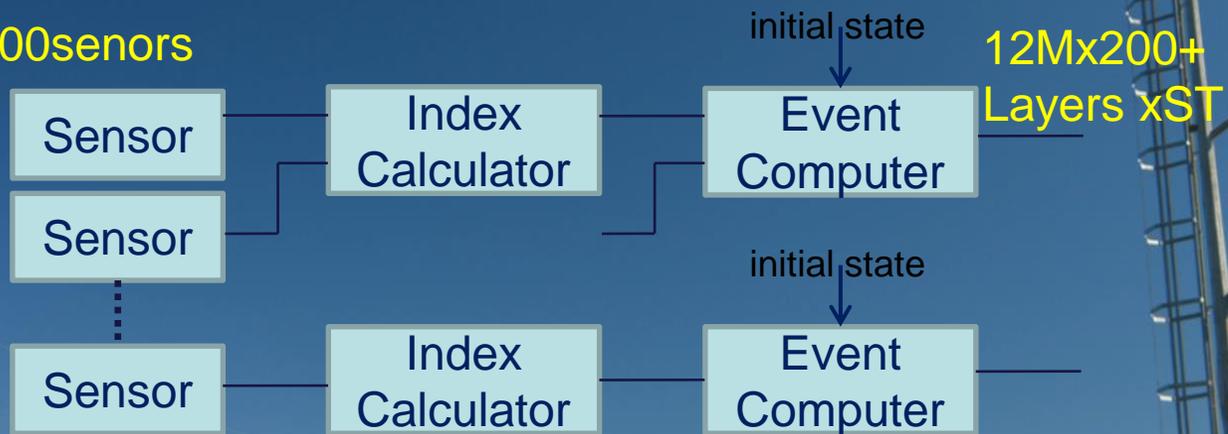
fx | =VLOOKUP(D2;'/Users/aydaaktas/Documents/Doktora/Agro-Geo2018/Data/46_04/[46_04_10mins.xlsx]Sampling'!\$L\$3:\$Q\$224;6;0)

A	B	D	E	F	G	H	I	J	K	L	M	N
Landsat Product Identifier	Landsat Scene Identifier	Acquisition Date	PS	Pixel Quality	LST (K)	LST @	Ts-Ta (SSDM)					
LC08_L1TP_174034_20161111	LC81740342016322LGN01	17.11.2016	1	Clear	289,13	17,98	3,36					
LE07_L1TP_174034_20161112	LE71740342016330NPA00	25.11.2016	1	Clear	290,91	19,76	6,78					
LC08_L1TP_174034_2016120	LC81740342016338LGN01	3.12.2016	1	Clear	285,95	14,80	2,57					
LE07_L1TP_174034_2016121	LE71740342016346NPA00	11.12.2016	2	Clear	286,95	15,80	3,80					
LC08_L1TP_174034_2016121	LC81740342016354LGN02	19.12.2016	2	Clear	280,79	9,64	3,62					
LE07_L1GT_174034_2016122	LE71740342016362NPA00	27.12.2016	2									
LC08_L1TP_174034_2017010	LC81740342017004LGN01	4.01.2017	2	Cloud	279,23							
LE07_L1TP_174034_2017011	LE71740342017012NPA00	12.01.2017	2	Clear	279,93	8,78	2,53					
LC08_L1TP_174034_2017012	LC81740342017020LGN01	20.01.2017	3	Clear	285,79	14,64	4,02					
LE07_L1TP_174034_2017012	LE71740342017028NPA00	28.01.2017	3	Snow								
LC08_L1TP_174034_2017020	LC81740342017036LGN00	5.02.2017	3	Cloud	272,21							
LE07_L1TP_174034_2017021	LE71740342017044NPA00	13.02.2017	3	Clear	287,91	16,76	6,41					
LC08_L1TP_174034_2017022	LC81740342017052LGN00	21.02.2017	3	Clear	293,59	22,44	8,26					
LE07_L1TP_174034_2017030	LE71740342017060NPA00	1.03.2017	4									
LC08_L1TP_174034_2017030	LC81740342017068LGN00	9.03.2017	4	Cloud	288,23	17,08	5,27					
LE07_L1GT_174034_2017031	LE71740342017076SG100	17.03.2017	4									
LC08_L1TP_174034_2017032	LC81740342017084LGN00	25.03.2017	4	Clear	299,45	28,30	9,07					
LE07_L1GT_174034_2017040	LE71740342017092SG100	2.04.2017	5	Cloud	241,40							
LC08_L1TP_174034_2017041	LC81740342017100LGN00	10.04.2017	5	Clear	294,20	23,05	7,74					
LE07_L1TP_174034_2017041	LE71740342017108SG100	18.04.2017	5	Cloud	249,41							
LC08_L1TP_174034_2017042	LC81740342017116LGN00	26.04.2017	5	Clear	306,18	35,03	10,43					
LE07_L1TP_174034_2017050	LE71740342017124SG100	4.05.2017	6	Clear	297,19	26,04						
LC08_L1TP_174034_2017051	LC81740342017132LGN00	12.05.2017	7	Clear	308,82	37,67	9,75					
LE07_L1GT_174034_2017052	LE71740342017140SG100	20.05.2017	7									
LC08_L1TP_174034_2017052	LC81740342017148LGN00	28.05.2017	7	Clear	306,28	35,13						
LE07_L1TP_174034_2017060	LE71740342017156SG100	5.06.2017	7	Clear	310,55	39,40	8,17					
LC08_L1TP_174034_2017061	LC81740342017164LGN00	13.06.2017		Clear	311,70	40,55	9,75					
LE07_L1TP_174034_2017062	LE71740342017172SG100	21.06.2017		Cloud	265,38							

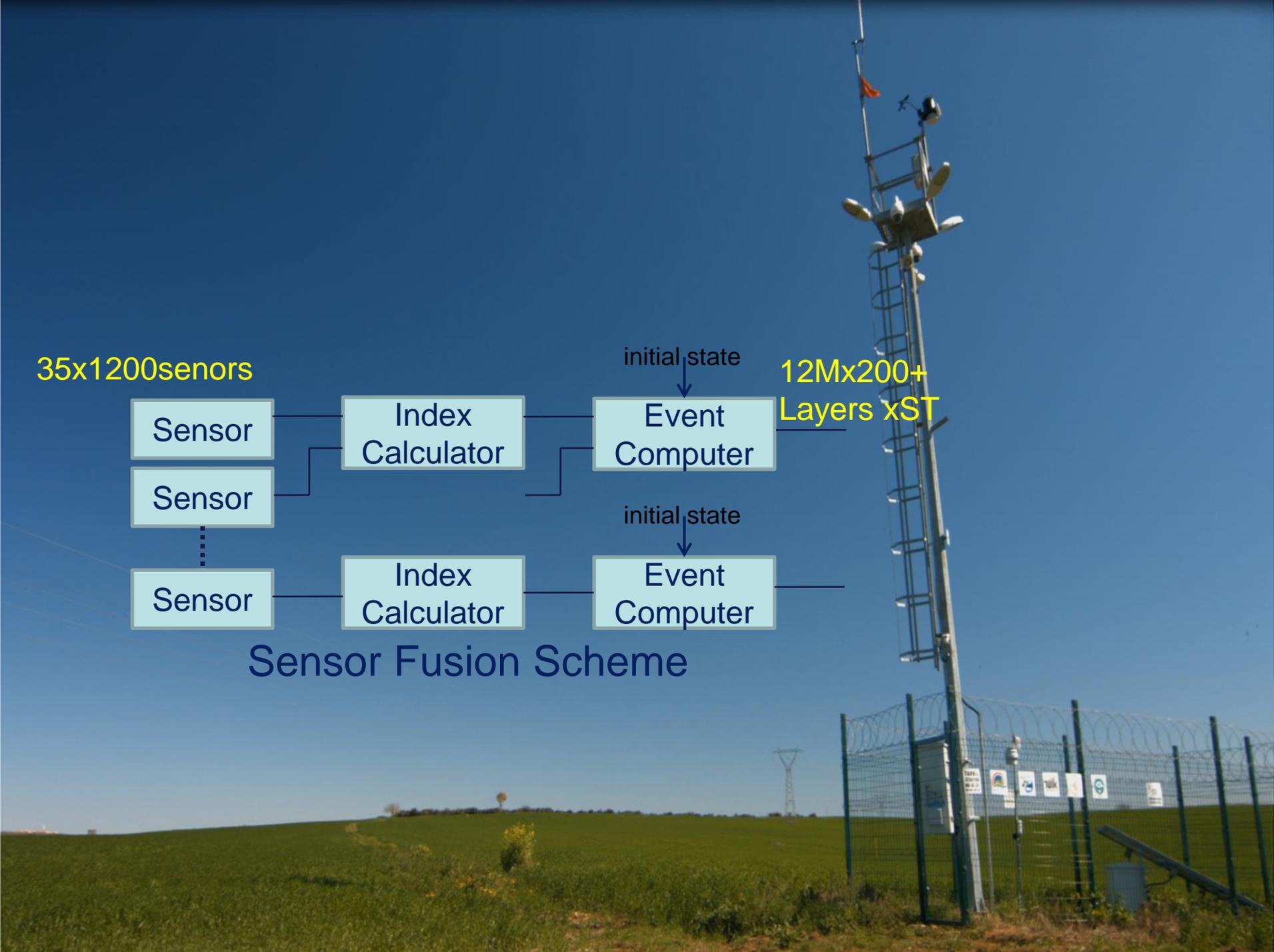
Distribution of monitoring stations in Turkey
(first 440 of 1200)



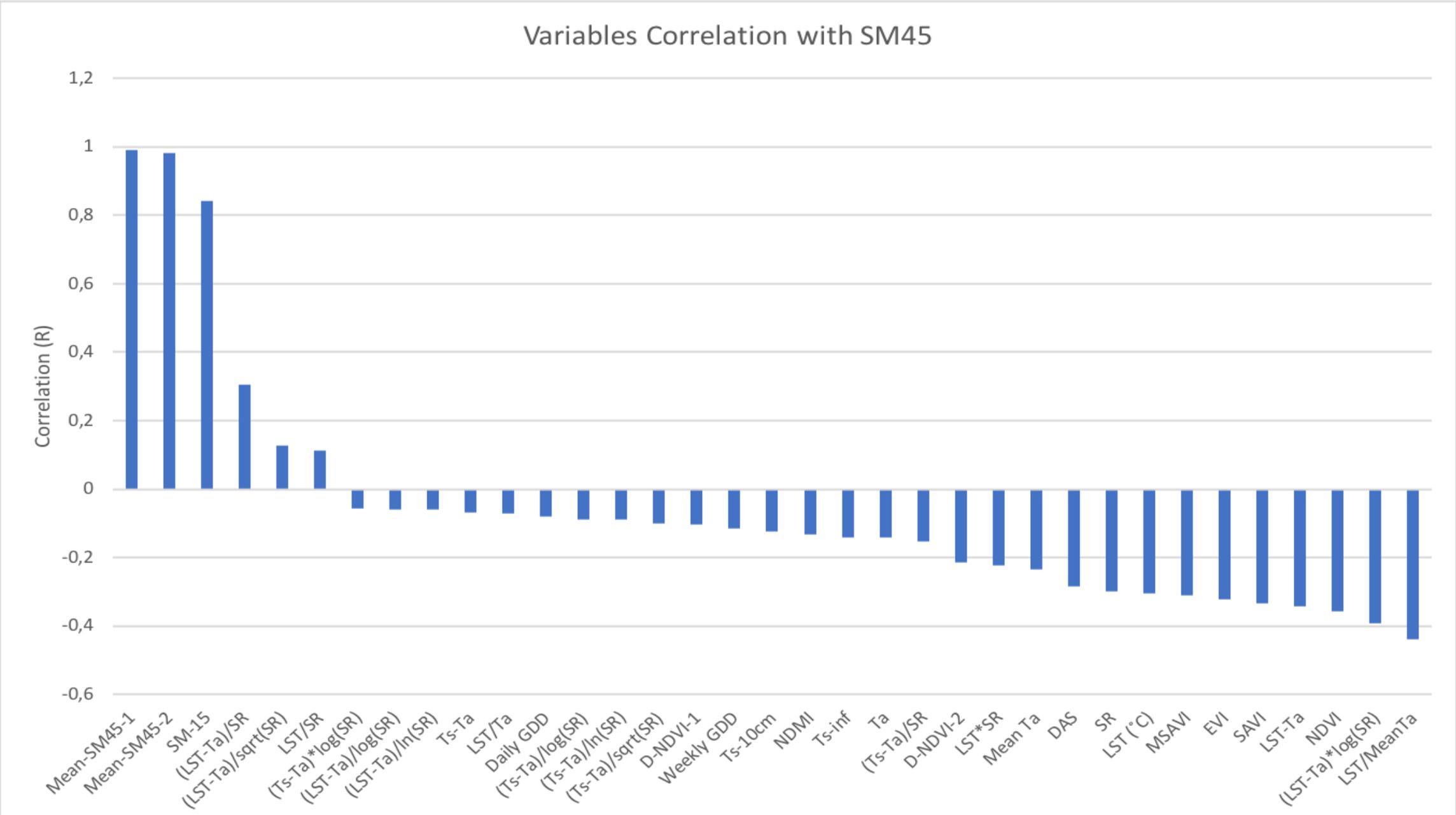
35x1200senors



Sensor Fusion Scheme

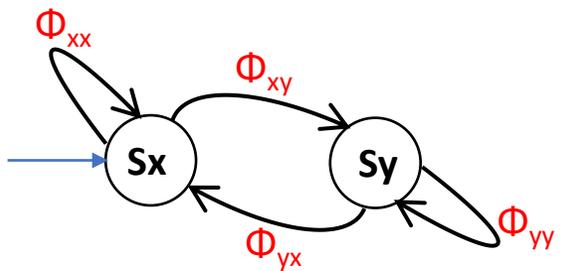
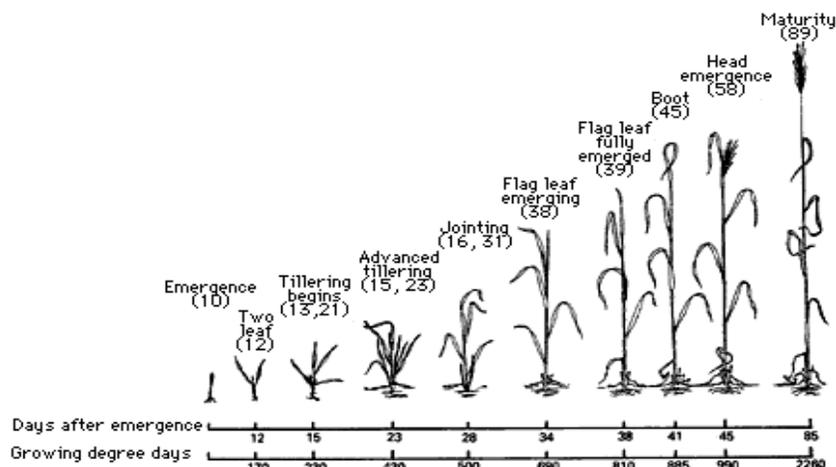


Nowcast or forecast of Invisible : Root zone soil moisture example



	SM-45	SM-15	DAS	Daily GDD	Weekly GDD	Mean Ta	LST/MeanTa	D-NDVI-1	Mean-SM45-1	D-NDVI-2	Mean-SM45-2	NDVI	EVI	SAVI	MSAVI	NDMI	Ts-infj	Ta	SR	Ts-10cm	Ts-Ta	(Ts-Ta)/SR	Ts-Ta)/log(SR)	Ts-Ta)/ln(SR)	LST (°C)	LST-Ta	(LST-Ta)/SR	ST-Ta)/log(SR)	ST-Ta)/ln(SR)	(SR*Ts-Ta)*log(SR)	LST*SR	LST/SR	LST/Ta	Ts-Ta)/sqrt((SR*ST-Ta)/sqrt(SR))		
SM-45	1																																			
SM-15	0.66516852	1																																		
DAS	0.72348061	0.35826524	1																																	
Daily GDD	-0.3080048	0.03665946	-0.6242423	1																																
Weekly GDD	-0.4201387	-0.0728364	-0.5610309	0.73704909	1																															
Mean Ta	-0.6262016	-0.2468563	-0.7103212	0.80262676	0.90209283	1																														
LST/MeanTa	-0.3475101	0.21872037	-0.2613882	0.32311785	0.3111267	0.41576053	1																													
D-NDVI-1	-0.376261	0.12620578	-0.3973817	0.09779169	0.01562527	0.0860722	0.01455259	1																												
Mean-SM45	0.98722114	0.62168421	0.73980289	-0.3411336	-0.4412476	-0.6486856	-0.3759203	-0.3385097	1																											
D-NDVI-2	-0.0858949	0.41278744	-0.3108008	0.10432127	0.01786673	-0.0212358	-0.0765479	0.89615879	-0.0983653	1																										
Mean-SM45	0.9824091	0.61151336	0.73842043	-0.3373671	-0.438957	-0.6508344	-0.3823654	-0.3321934	0.99937828	-0.0991339	1																									
NDVI	-0.4128661	-0.0138345	-0.5454967	-0.0546087	-0.0115806	-0.0055315	0.07564698	0.79387237	-0.4009328	0.77294764	-0.3959694	1																								
EVI	-0.4310242	-0.091013	-0.5915214	-0.0086263	0.03791218	0.03986958	-0.1340884	0.79996715	-0.4313044	0.81150096	-0.4280079	0.94718831	1																							
SAVI	-0.3183441	0.01226298	-0.5120801	-0.0928504	-0.0768815	-0.085171	-0.0923523	0.79812497	-0.3094518	0.81966941	-0.3055452	0.97587376	0.97897738	1																						
MSAVI	-0.1771784	0.09796461	-0.4579616	-0.1052205	-0.0779619	-0.1360798	-0.1537081	0.66923693	-0.1995751	0.79850551	-0.2001052	0.925716	0.94461772	0.96569618	1																					
NDMI	0.80287442	0.58081254	0.56753999	-0.1851849	-0.2728218	-0.4253238	0.07881733	-0.4017349	0.83791866	-0.2740833	0.8406764	-0.3674043	-0.5446823	-0.380447	-0.3237039	1																				
Ts-inf	-0.4843896	-0.1095561	-0.8032544	0.90605889	0.85054508	0.90643664	0.337141	0.07856674	-0.538771	0.10757129	-0.541518	0.08807302	0.16585322	0.05237927	0.06533543	-0.3878513	1																			
Ta	-0.3411092	-0.0104438	-0.5594171	0.87044491	0.630748	0.83406945	0.34992618	0.13846689	-0.3604686	0.05109729	-0.3651478	-0.1365134	-0.1065729	-0.1832391	-0.2395443	-0.1747102	0.79245294	1																		
SR	-0.4950262	-0.2080795	-0.7168362	0.92167999	0.65578295	0.79216793	0.13753019	0.19870627	-0.5402875	0.17331212	-0.5387709	0.04781506	0.17025566	0.03750583	0.02215241	-0.5111266	0.89133482	0.82126993	1																	
Ts-10cm	-0.47894	0.05387788	-0.6077567	0.87357192	0.80649359	0.91395801	0.46638781	0.23924011	-0.5146201	0.18417524	-0.5191663	-0.019642	0.03603079	-0.0734399	-0.1298047	-0.3608552	0.87299353	0.89797448	0.83129718	1																
Ts-Ta	-0.4206372	-0.1639327	-0.7039684	0.5471344	0.70581504	0.58525176	0.17673394	-0.0175181	-0.4873906	0.11884063	-0.4869454	0.28123554	0.37438937	0.27250602	0.3512356	-0.4383004	0.77737283	0.23235901	0.57437011	0.46601472	1															
(Ts-Ta)/SR	-0.3939484	-0.025606	-0.7864339	0.44381666	0.56944255	0.45485032	0.06270615	0.41334342	-0.4306747	0.54068995	-0.4278287	0.67524634	0.74221301	0.68405986	0.72612676	-0.416746	0.65968278	0.18769357	0.48509838	0.38376772	0.85839016	1														
(Ts-Ta)/log(S)	-0.4263473	-0.1369474	-0.7446388	0.53007043	0.68929172	0.56541214	0.14897539	0.09400683	-0.4863566	0.23079864	-0.4851377	0.39537411	0.48318102	0.39033297	0.46167644	-0.4439656	0.76470861	0.22194804	0.56205712	0.45178527	0.99054212	0.92044381	1													
(Ts-Ta)/ln(SR)	-0.4263129	-0.1369158	-0.7446853	0.53020077	0.68931666	0.56545664	0.14887583	0.09408548	-0.4863266	0.23089837	-0.4851072	0.39538093	0.48321723	0.39035766	0.46170886	-0.4439885	0.76478704	0.22207866	0.56220618	0.45186774	0.99053247	0.92047535	0.99999997	1												
LST (°C)	-0.6387603	-0.1946217	-0.7030259	0.64567447	0.54096192	0.76979357	0.30414442	0.29318825	-0.68631	0.20226595	-0.6972562	0.08467609	0.22775089	0.09159102	0.02133124	-0.664693	0.75934273	0.73420727	0.76821513	0.83767455	0.45367888	0.38031662	0.44077586	0.44083152	1											
LST-Ta	-0.6184731	-0.2731297	-0.5139498	0.14776482	0.21356175	0.36126948	0.12417465	0.30025446	-0.6701062	0.24781797	-0.681783	0.24733227	0.42822054	0.29991918	0.24911951	-0.8076031	0.38415077	0.15853411	0.37057029	0.40186298	0.44908581	0.38296951	0.43979283	0.43975503	0.78673596	1										
(LST-Ta)/SR	0.39574681	0.46775289	0.46586306	-0.7533263	-0.6342829	-0.7118513	0.07442397	0.09145014	0.39216744	0.20250498	0.38133619	0.20391441	0.1201874	0.24495029	0.26438358	0.31570427	-0.6813769	-0.6912758	-0.8085353	-0.5902965	-0.373627	-0.2180043	-0.343078	-0.3432137	-0.4051	0.03947585	1									
(LST-Ta)/log(S)	-0.3291487	-0.0122527	-0.1841691	-0.2615031	-0.1342047	-0.0441552	0.12133943	0.2990285	-0.3724367	0.30112009	-0.3879128	0.28815327	0.40677073	0.35537061	0.31809289	-0.5384692	-0.0263638	-0.2047414	-0.0930924	0.04962043	0.16911412	0.18859883	0.17525081	0.17515275	0.47084225	0.87091515	0.52261575	1								
(LST-Ta)/ln(S)	-0.329206	-0.0121737	-0.1842051	-0.2615138	-0.1341582	-0.0440876	0.12142041	0.29923037	-0.3724802	0.30128214	-0.3879577	0.28828974	0.40686991	0.35548254	0.31817114	-0.5384667	-0.0263703	-0.2046567	-0.0931181	0.04970285	0.16901658	0.18861457	0.17518083	0.17508277	0.47087611	0.8708874	0.52265329	0.99999995	1							
(Ts-Ta)*log(S)	-0.4097925	-0.1697225	-0.6731505	0.5614745	0.70957228	0.59465489	0.19604735	-0.0839915	-0.4811889	0.05288398	-0.4814438	0.20351356	0.3007878	0.19394216	0.2765105	-0.4304174	0.78348204	0.24574965	0.58453192	0.47956709	0.99593195	0.81050807	0.97440742	0.97439382	0.46724412	0.45663083	-0.3889238	0.16879945	0.16868515	1						
(LST-Ta)*log(S)	-0.718513	-0.3850967	-0.6775105	0.39704667	0.40146217	0.56307346	0.13464848	0.26505302	-0.771461	0.19529665	-0.7791249	0.21773649	0.40672935	0.25229684	0.20379611	-0.860897	0.60580916	0.35902105	0.61770459	0.56720081	0.59578599	0.4873274	0.57950607	0.57950871	0.87880012	0.95161751	-0.2549434	0.68083374	0.68078676	0.60621671	1					
LST*SR	-0.5200991	-0.1679743	-0.761311	0.90941857	0.60438181	0.77561727	0.28788445	0.16811866	-0.5725454	0.13977788	-0.5737559	0.0555697	0.16011189	0.04330461	0.01852919	-0.4905807	0.90211619	0.80757412	0.96450215	0.83812543	0.60568749	0.48776097	0.58551179	0.58562536	0.83798765	0.48452368	-0.6758724	0.06264736	0.06260197	0.62236777	0.69974118	1				
LST/SR	-0.199424	0.19079318	-0.1168021	-0.0798746	-0.0131331	0.16814329	0.22006964	0.31946674	-0.2352182	0.28987476	-0.2582535	0.06465212	0.15502186	0.11443052	0.05313503	-0.3309876	0.06515856	0.21260596	0.00681296	0.32762755	-0.115361	-0.0435094	-0.1052228	-0.1052574	0.60628314	0.68836134	0.38169206	0.79117078	0.79129187	-0.1115579	0.51920861	0.12474064				

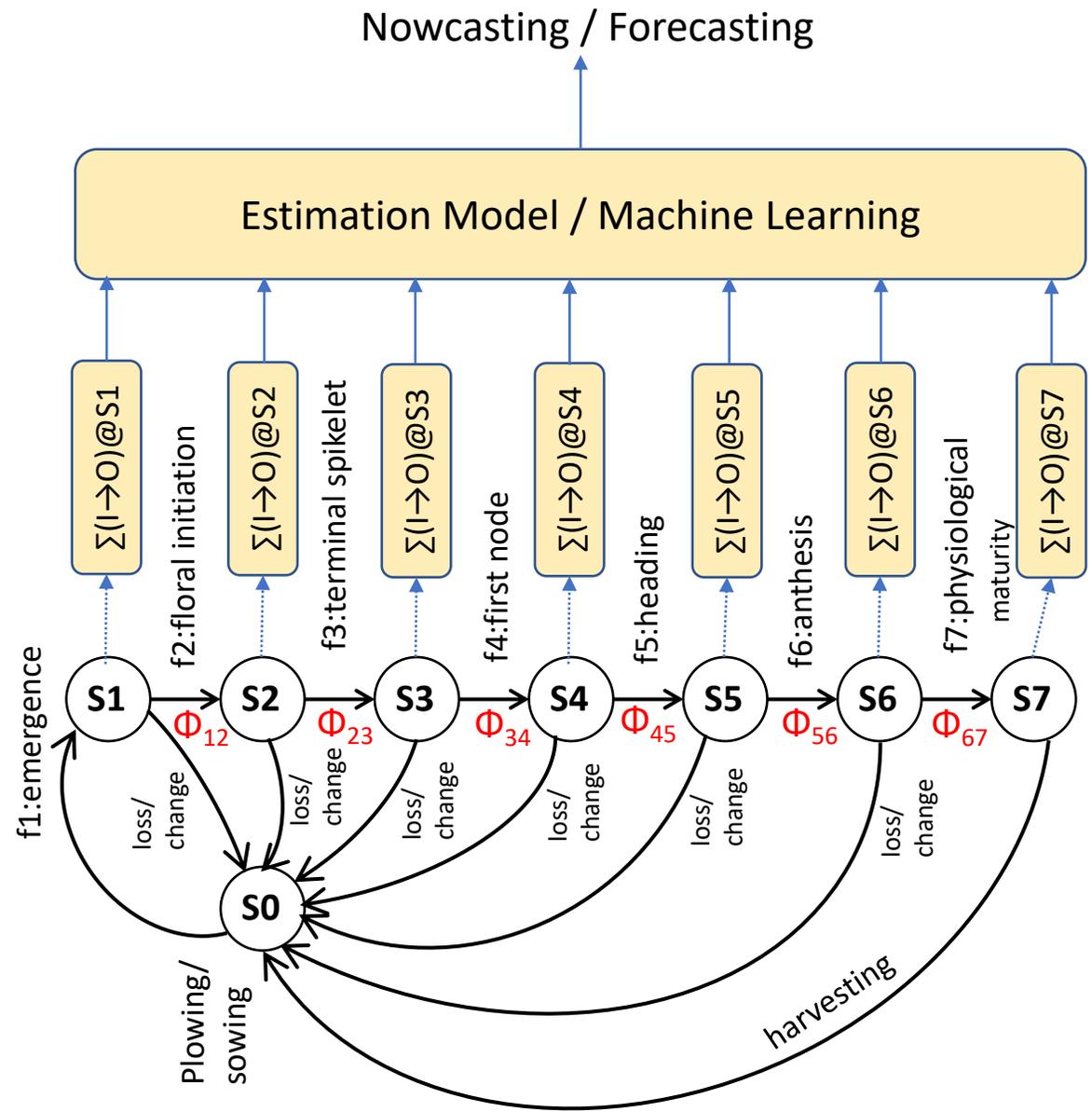
Markovian state representation of the phenological stages



$$O(x,n)=f_x(I(n),I(n-1),\dots,I(0))$$

$$O(y,n)=f_y(I(n),I(n-1),\dots,I(0))$$

$$S_x \rightarrow S_y : \Phi_{xy}(I(n),I(n-1),\dots,I(0))$$



Phenological stages can both be observed and nowcasted depending on monitoring schedule of the Digital Twinning system



Before sowing
12.11.2016



After sowing
12.11.2016



Phenological Stage 1
Emergence
07.12.2016



Phenological Stage 2
Floral Initiation
04.01.2017



Phenological Stage 3
Terminal Spikelet
15.02.2017



Phenological Stage 4
First Node
29.03.2017



Phenological Stage 5
Heading
02.05.2017



Phenological Stage 6
Anthesis
18.05.2017

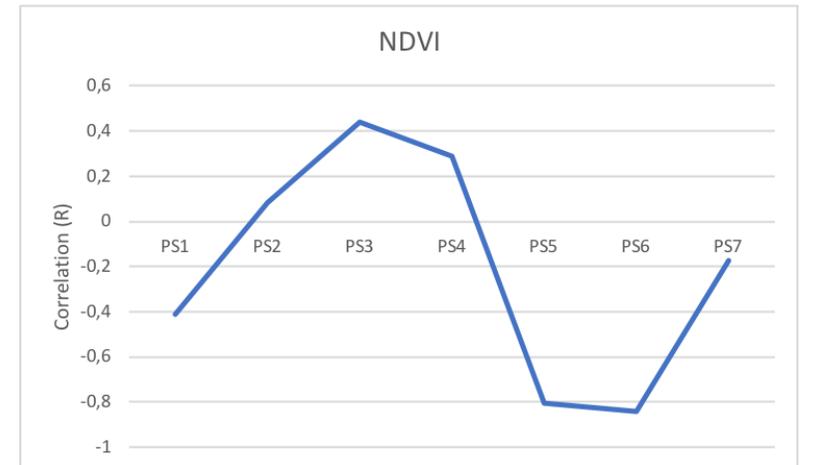
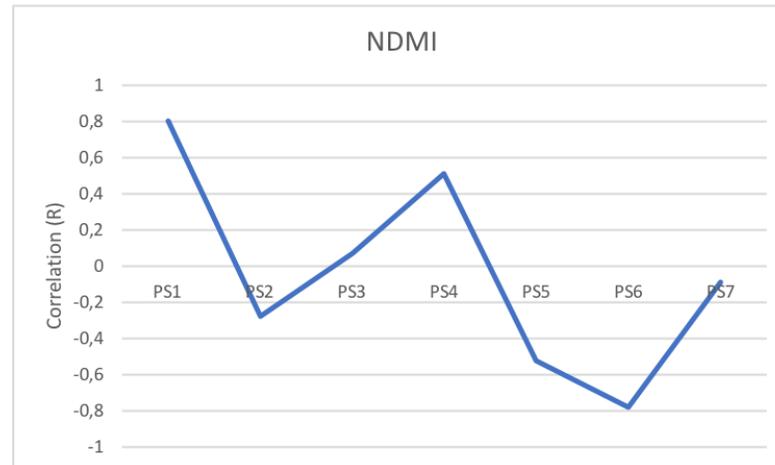
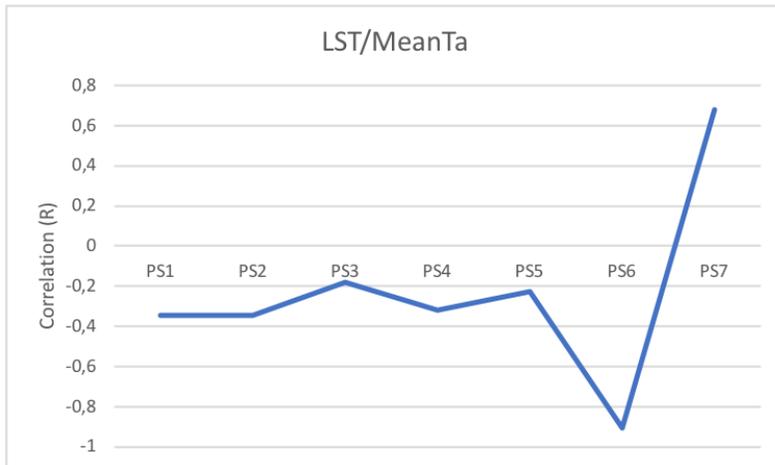
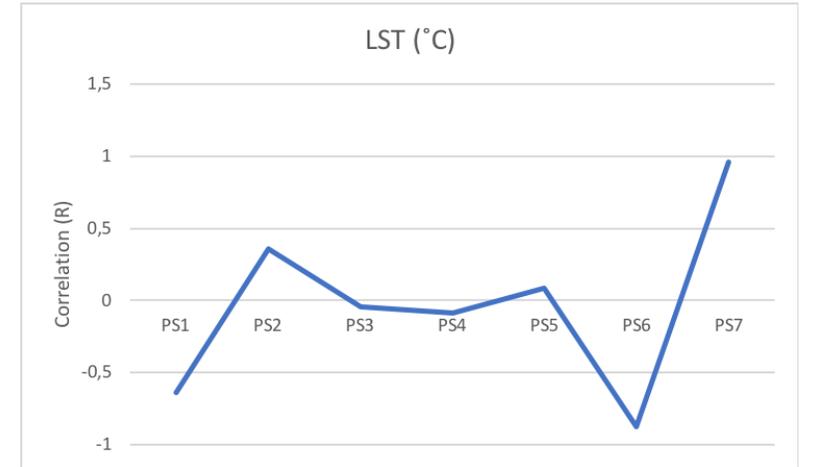
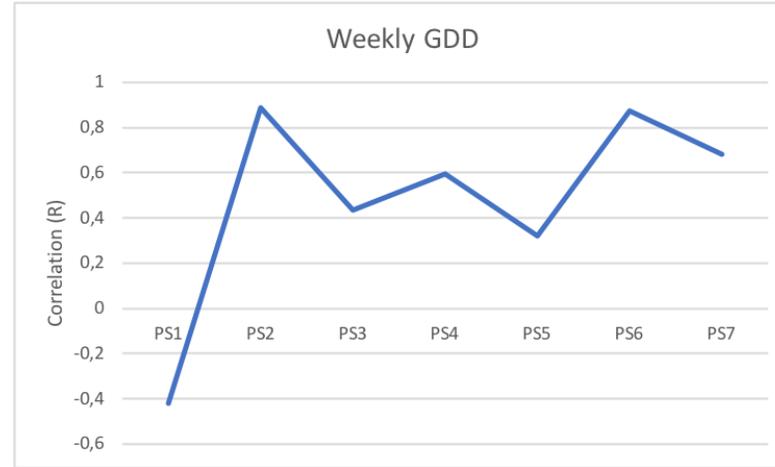
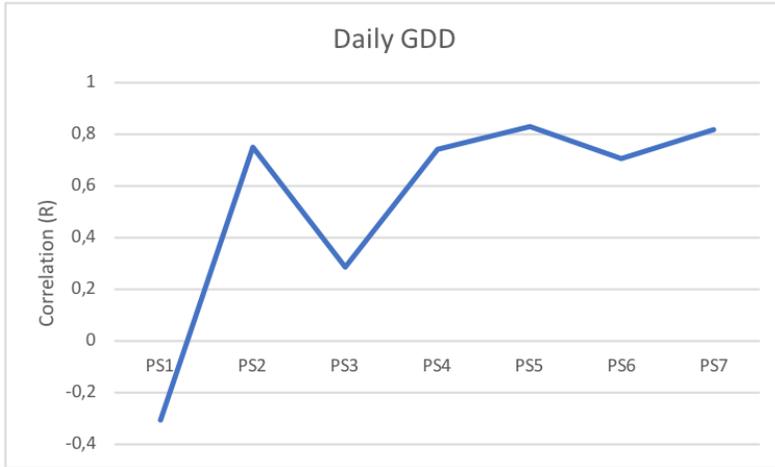


Phenological Stage 7
Maturity
05.06.2017



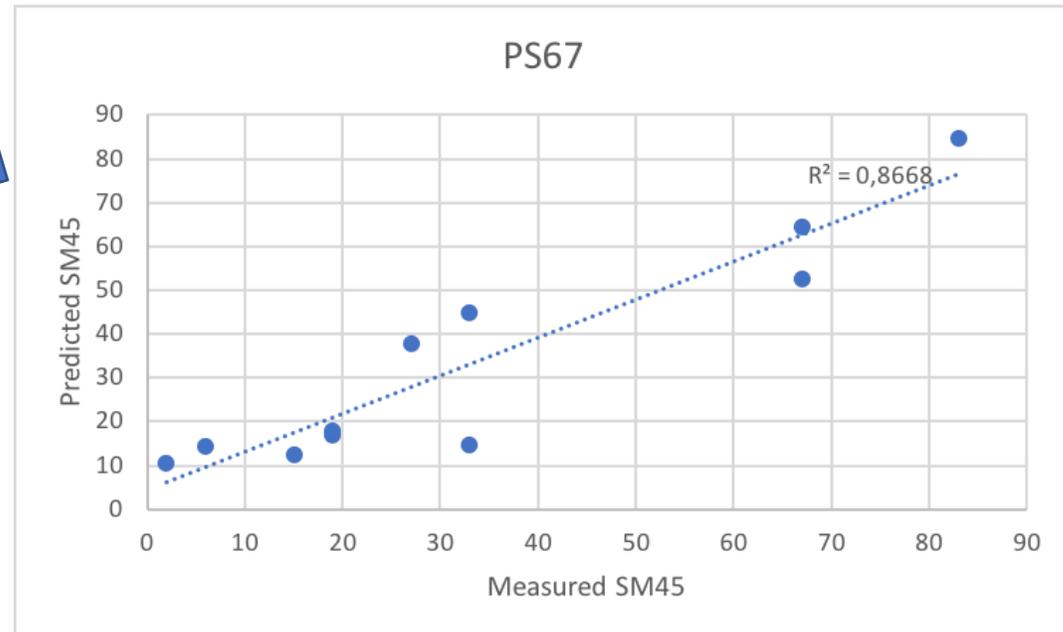
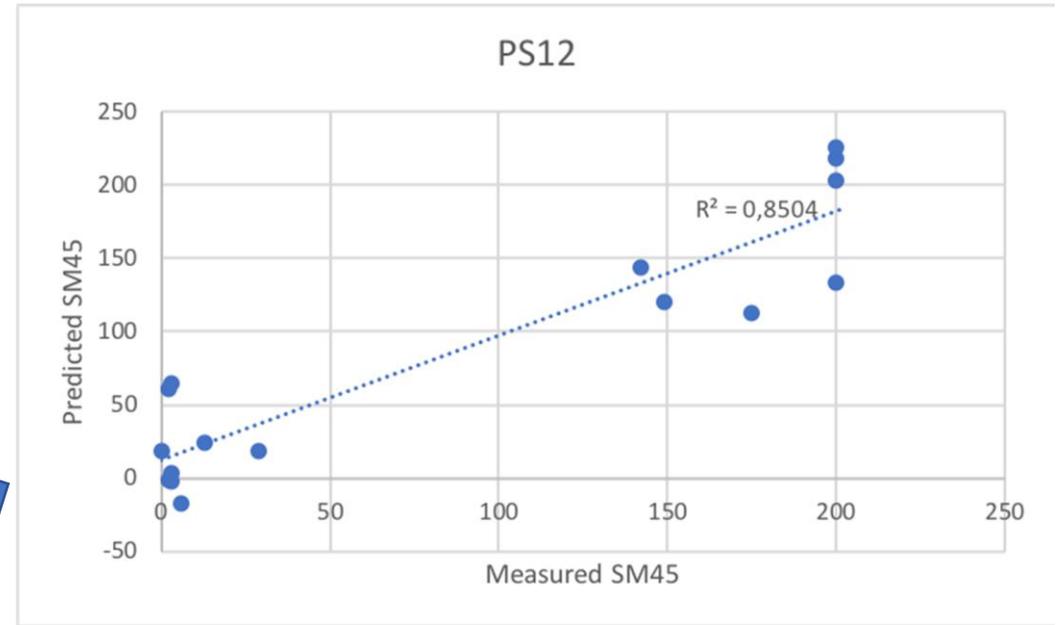
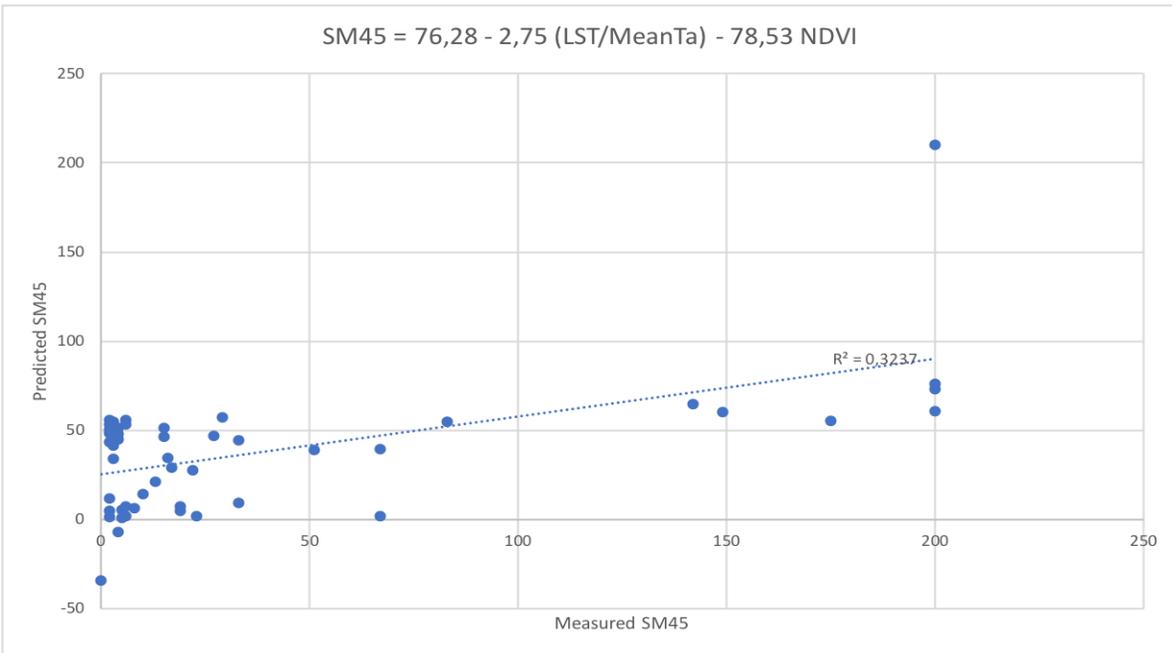
Harvest
20.06.2017

Correlation to the root zone soil moisture



2- ROLE of DATA SEGMENTATION

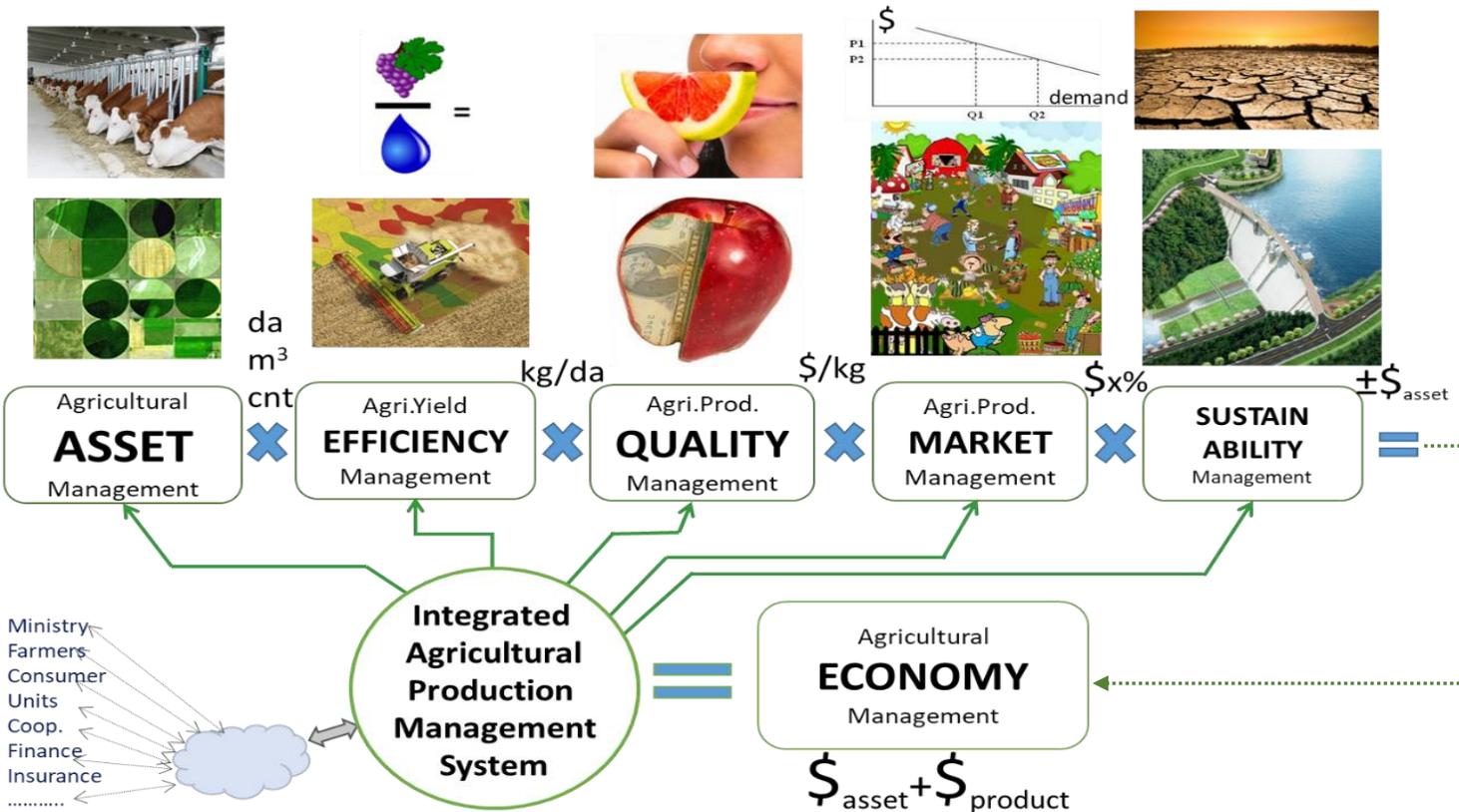
Without phenological stage segmentation:
Two highest correlating parameters $R^2=0.3237$



Potential of increasing the efficiency in a sustainable way:

Convergence through Digital Twinning

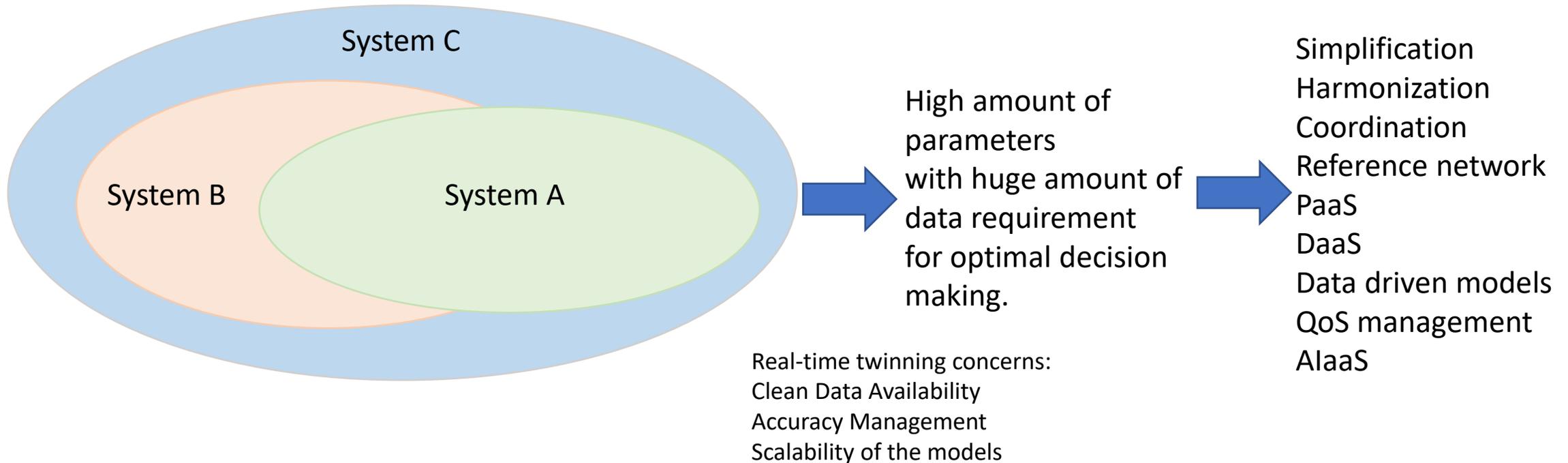
- 1- Integrated Agricultural Production and Food Supply Chain Management
- 2- Precision Agriculture
- 3- Novel production technologies (vertical farming, indoor farming)



Digital Twinning in Crop Production

- System A: Plant system: Non-linear time varying in interaction with soil and the atmosphere (time constant is in days range if we neglect variety development)
- System B: Non-linear time varying soil structure (time constant is around year range)
- System C: Climate and Ecology (Chaotic and Stochastic) (time constant is in decade range)
- System D: Cropping system: Rule-based, adaptive

Greenhouse agriculture and indoor farming systems differs from field conditions for digital twinning strategies



Real-time Continuous Monitoring

→ Real-time Diagnostics, Reconstruction, Fusion, Super resolution

Remote sensing problems:

- Cloudiness
- Surface disturbance / Snow coverage
- Resolution differences
- Viewing and altitude

Insitu monitoring problems:

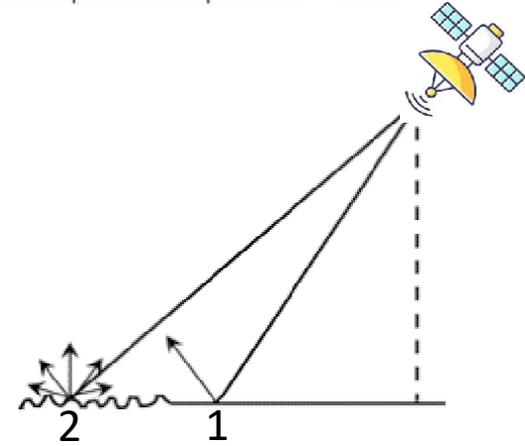
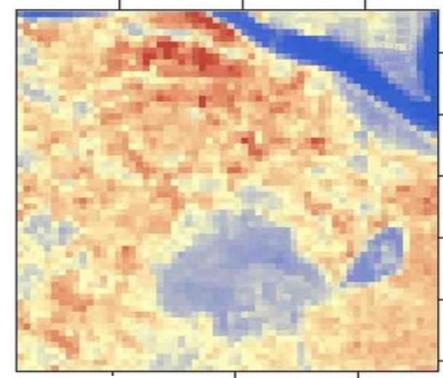
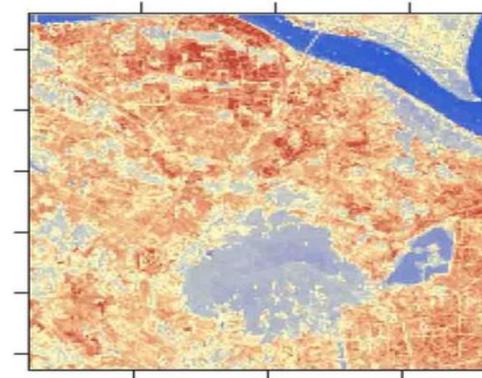
- Calibration requirement
- Sensor maintenance
- Disturbances
- Interruptions (Communication, maintenance, energy...)
- Temporal data is available at specific monitoring locations
- Distribution of the locations are non-uniform
- Multi-temporal / spatial data is available at some predefined instants

Sampling & Resolution differences:

- Temporal data is available at specific monitoring locations
- Distribution of the locations are non-uniform
- Multi-temporal / spatial data is available at some predefined instants

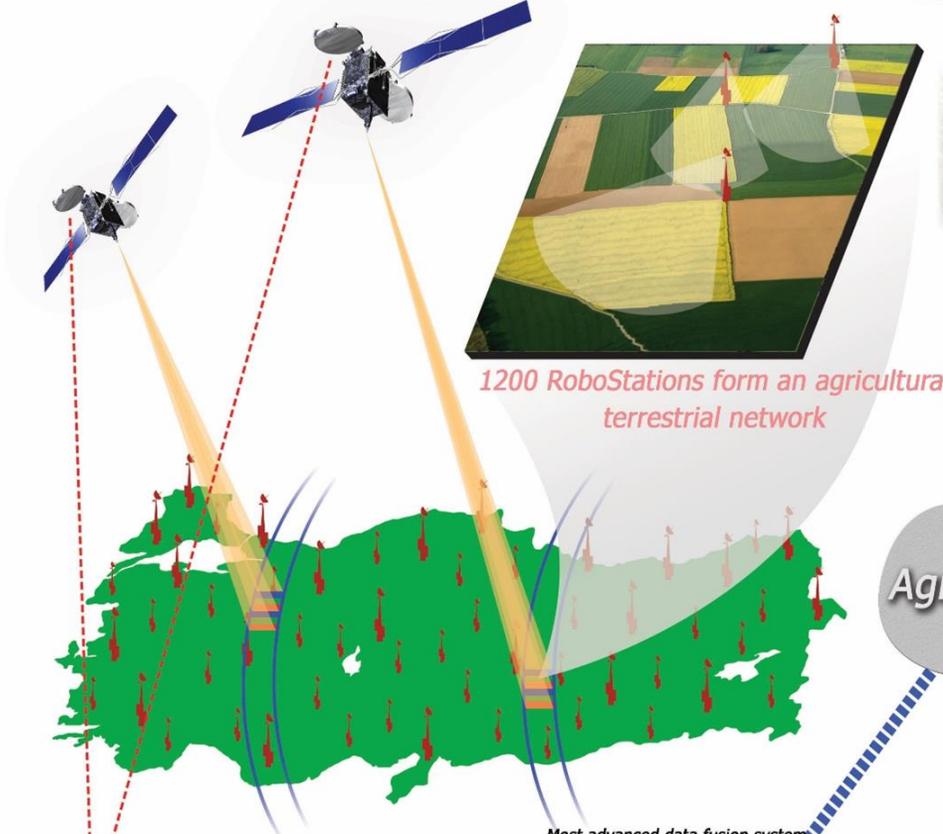
Farm registry problems:

- Human data entry failures
- Statistical sampling problems
- Farmers may not share data
- Data resource discontinuity



TARBIL

Remote sensing satellite fleet of 0.5m, 1.5m, 2.5m, 5m spatial high-resolution integrated in TARBIL



Agricultural experts access field and plant measurement data using TARBIL tablets. Through corresponding observations, experts provide farmers with prescriptions for most effective use of fertilizers/pesticides as well as the other agricultural activities.



Farmers access TARBIL and Agricultural Information System (TBS) using mobile devices. In this way, farmers can follow the official (Ministry) procedures using TBS system and carry out their daily agricultural activities using Tarlamatik information support application.



Crop yield maps are created while reaper and combine harvesters are working on the field. Meanwhile remote sensing models are calibrated with the acquired yield data. In this way farmers can track the expected yield of their crops as from the sowing phase.



Agriculture experts of the Ministry (TARGEL) carry out the livestock registration and farmer support procedures in towns.



Depending on the actual plant conditions, automatic irrigation system's agenda is continuously rescheduled for economical water consumption.



Agricultural machinery and equipment utilize TARBIL data for precision agriculture applications during the planting, fertilization, disinfection, and maintenance phases.



Most advanced data fusion system for the agricultural development
TARBIL CENTER



Large aera coverage satellite ground station among top 5 in the world:
ITU UHUZAM (CSCRS)

First national satellite remote sensing management and archive system in Turkey:
UGIP



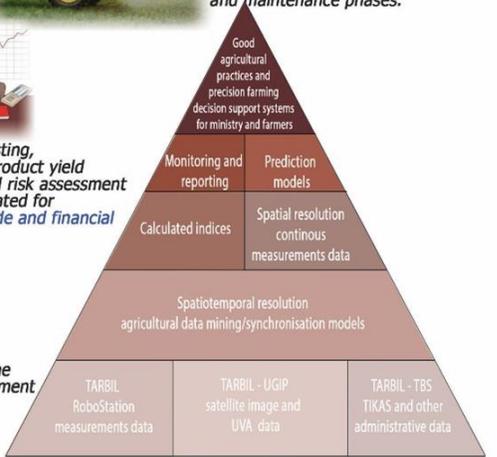
GPU and multi-core data processing system
TARBIL Command Control Center

Real-time agricultural computations in 12 million different points across Turkey through agricultural terrestrial station network equipped with more than 30 000 sensors together with UGIP data

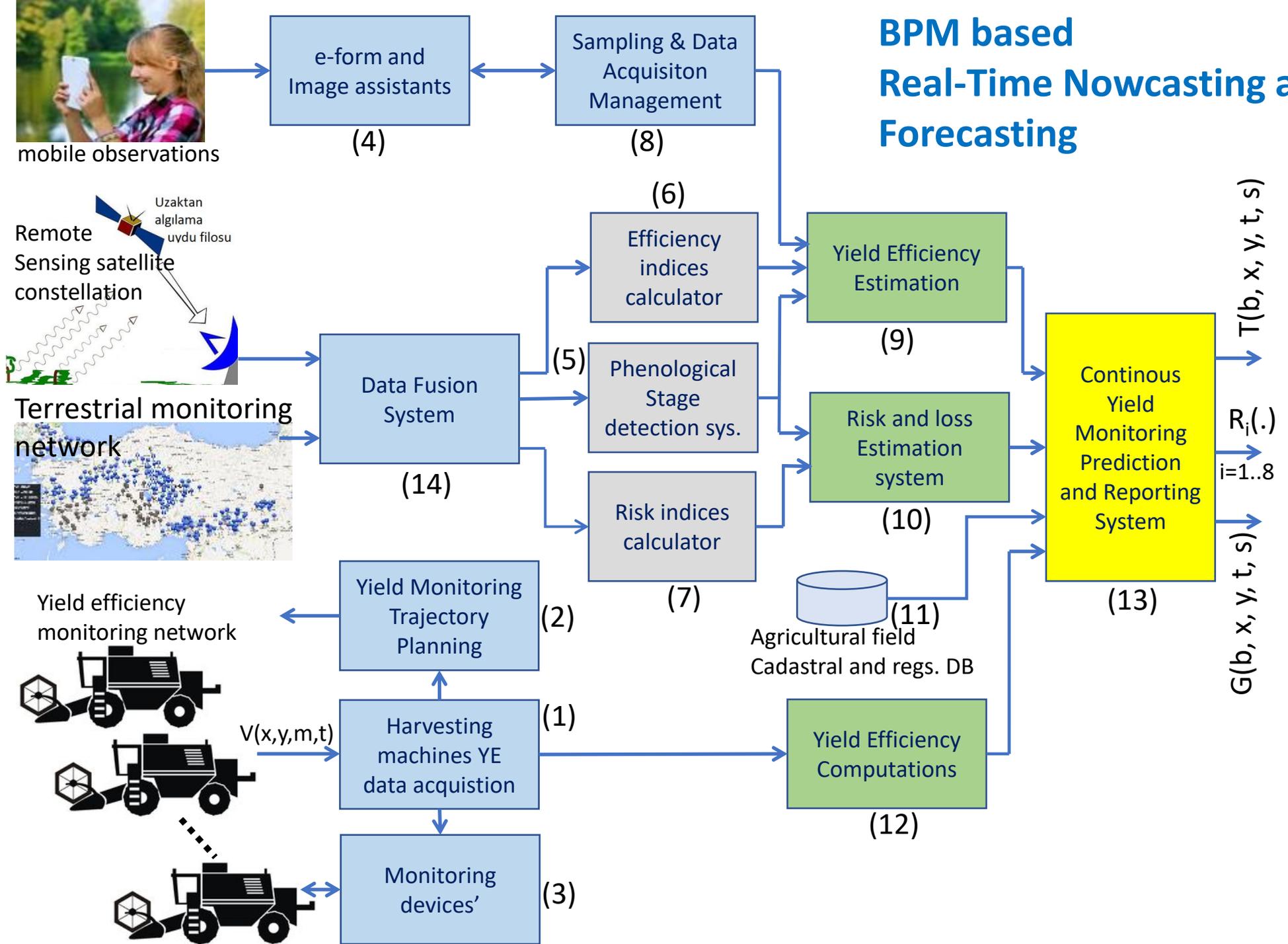


Harvest forecasting, region based product yield distribution and risk assessment data are generated for agricultural trade and financial management.

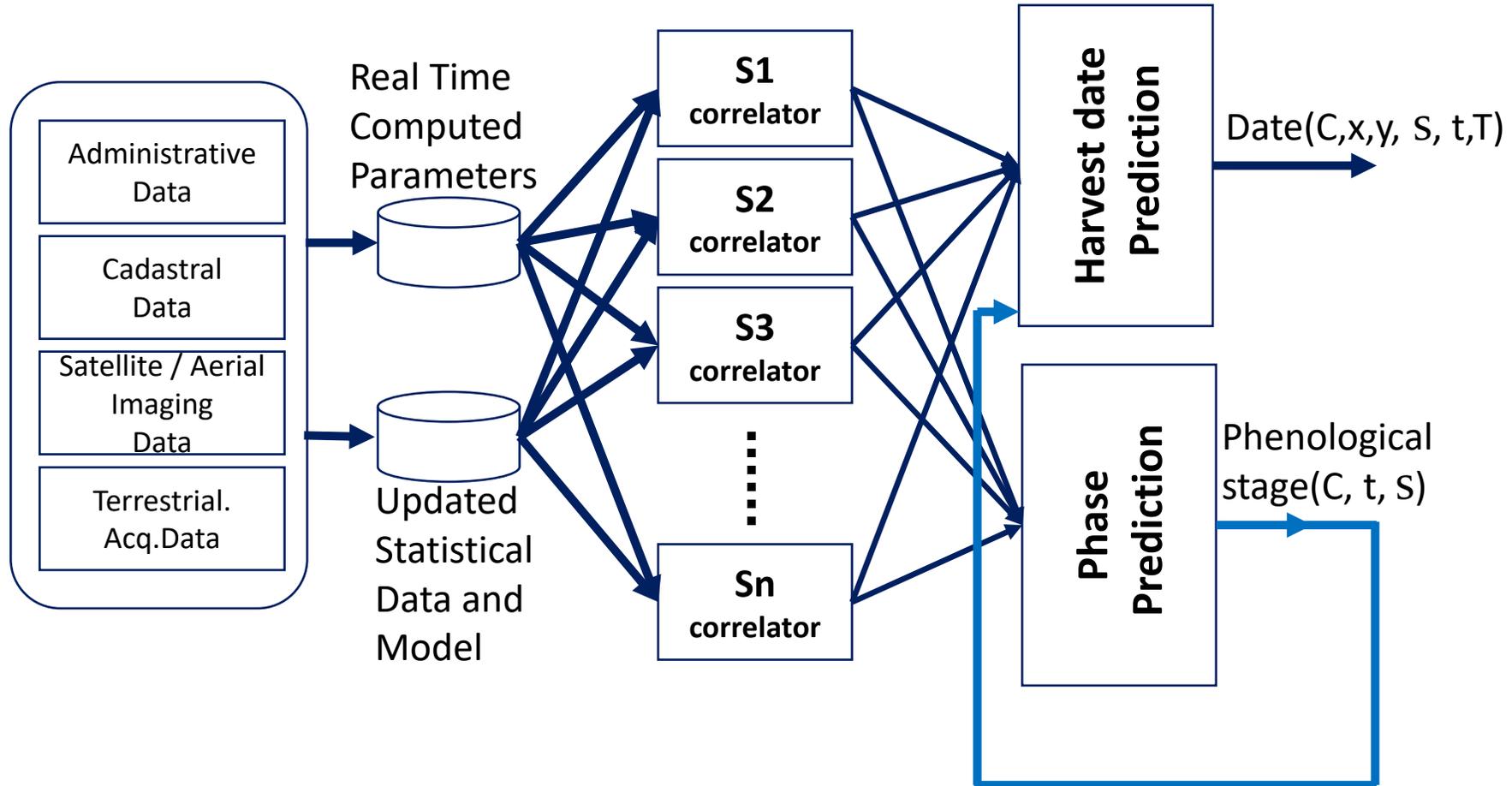
Ministry of Food, Agriculture and Livestock and related units use TARBIL monitoring and decision support systems together with TBS in determining the rational agricultural policies for the efficient management of agricultural economy and dissemination of good agricultural practices.



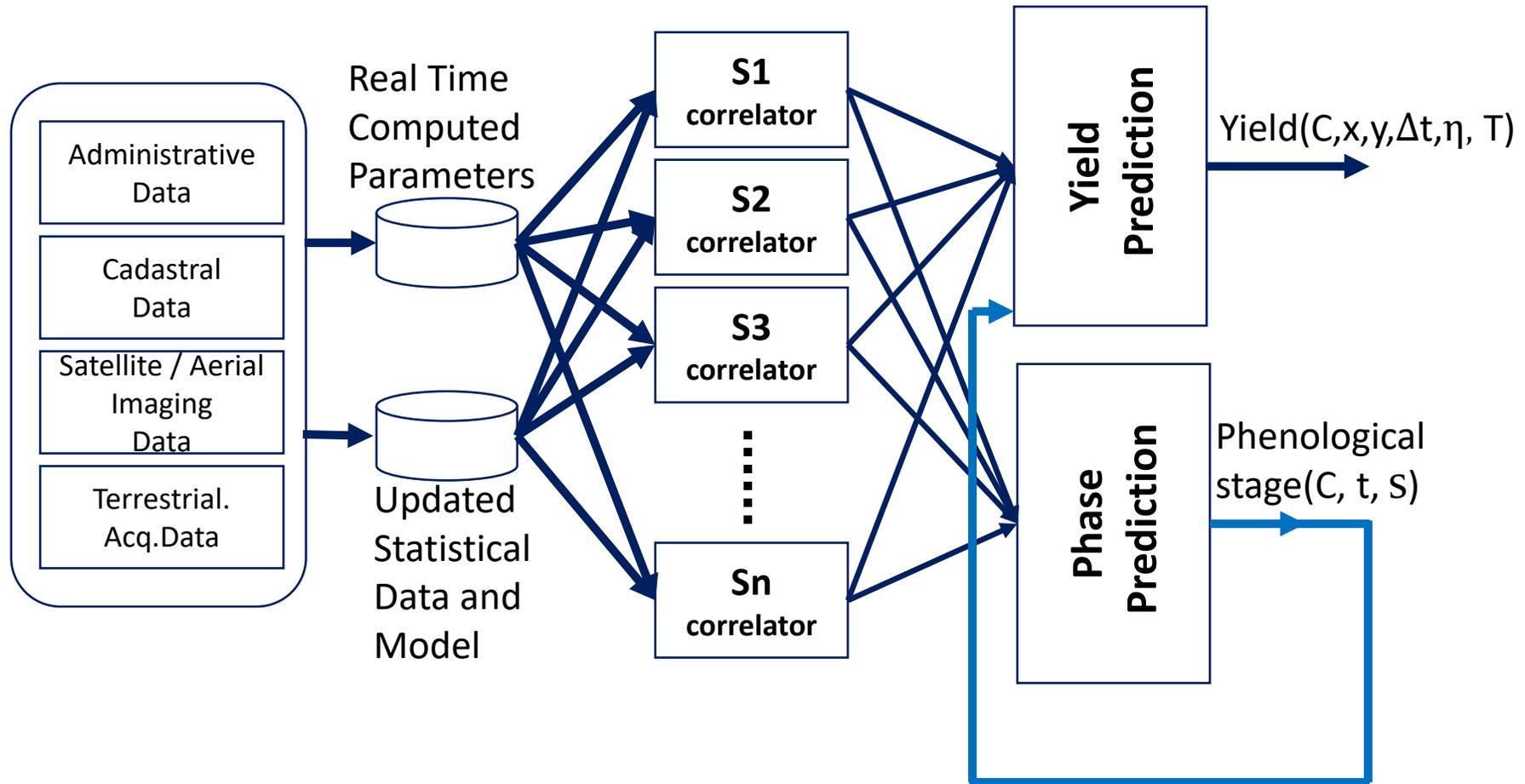
BPM based Real-Time Nowcasting and Forecasting



Phenological stage transition date estimation based on finite-state machine model



Yield Nowcasting and Forecasting based on finite-state machine model

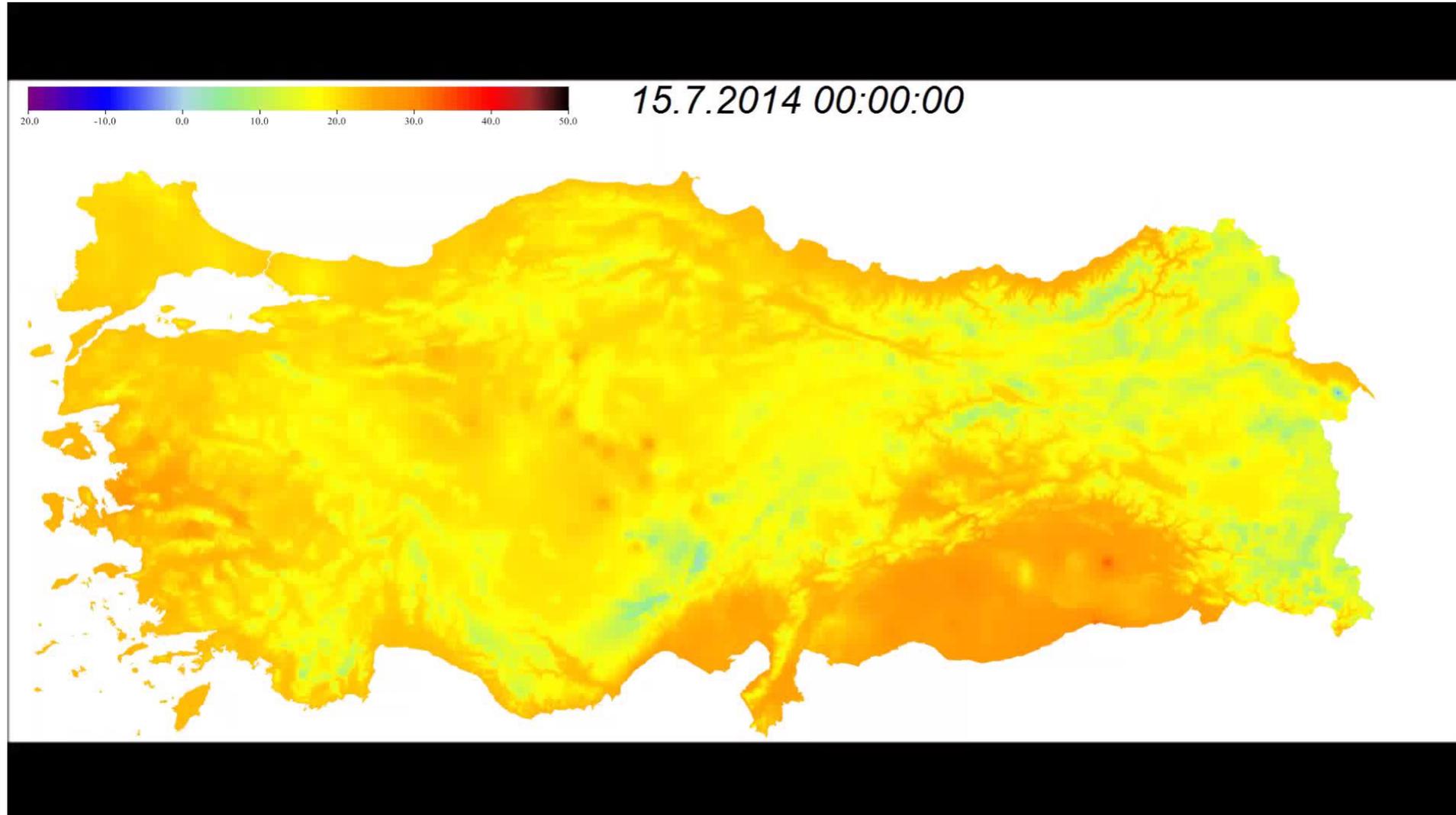


1200stations, 42.000sensors, 10min-30min sampling interval

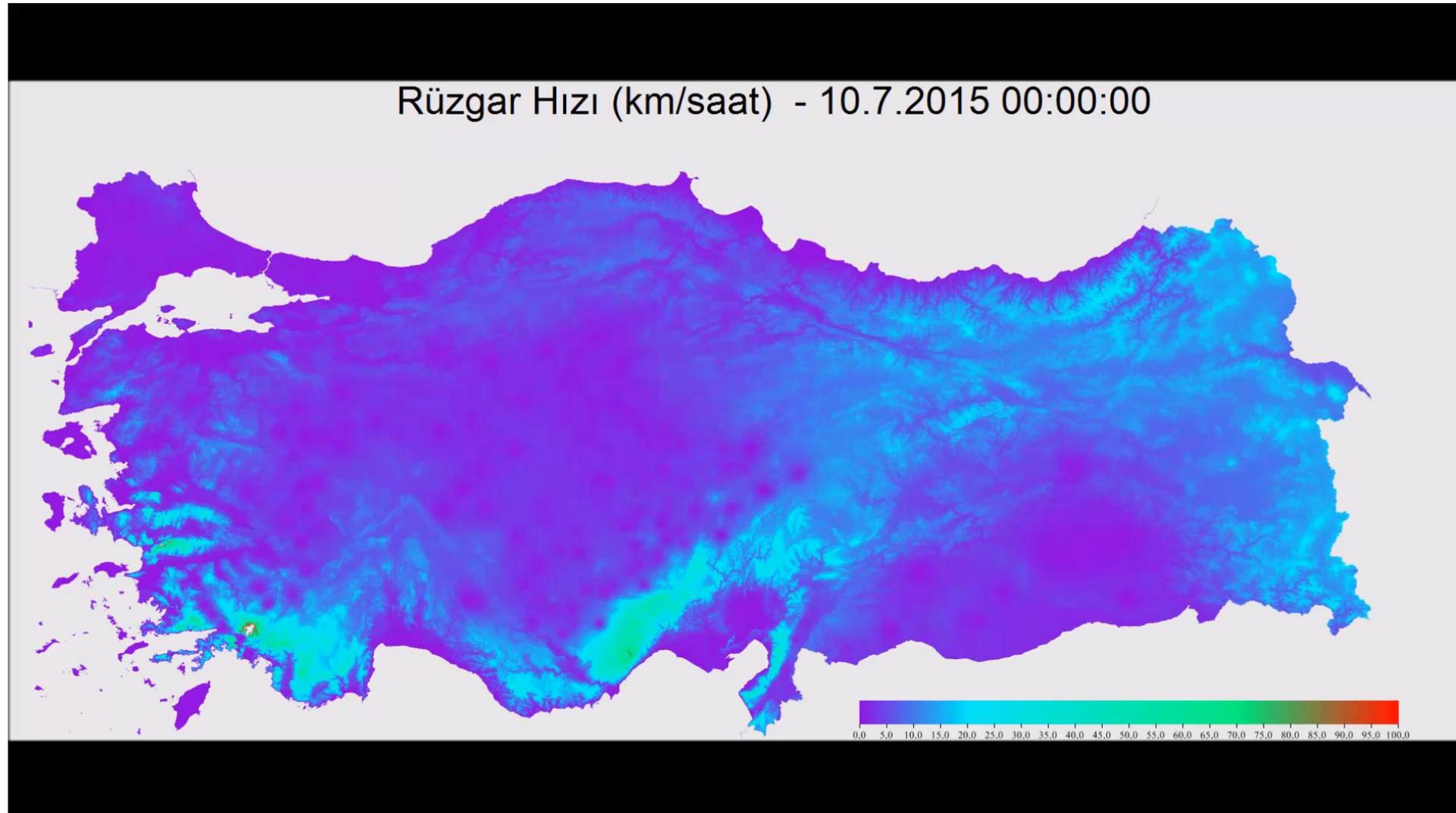
250mx250m model based interpolations @more than 12Million points 30mins..1day



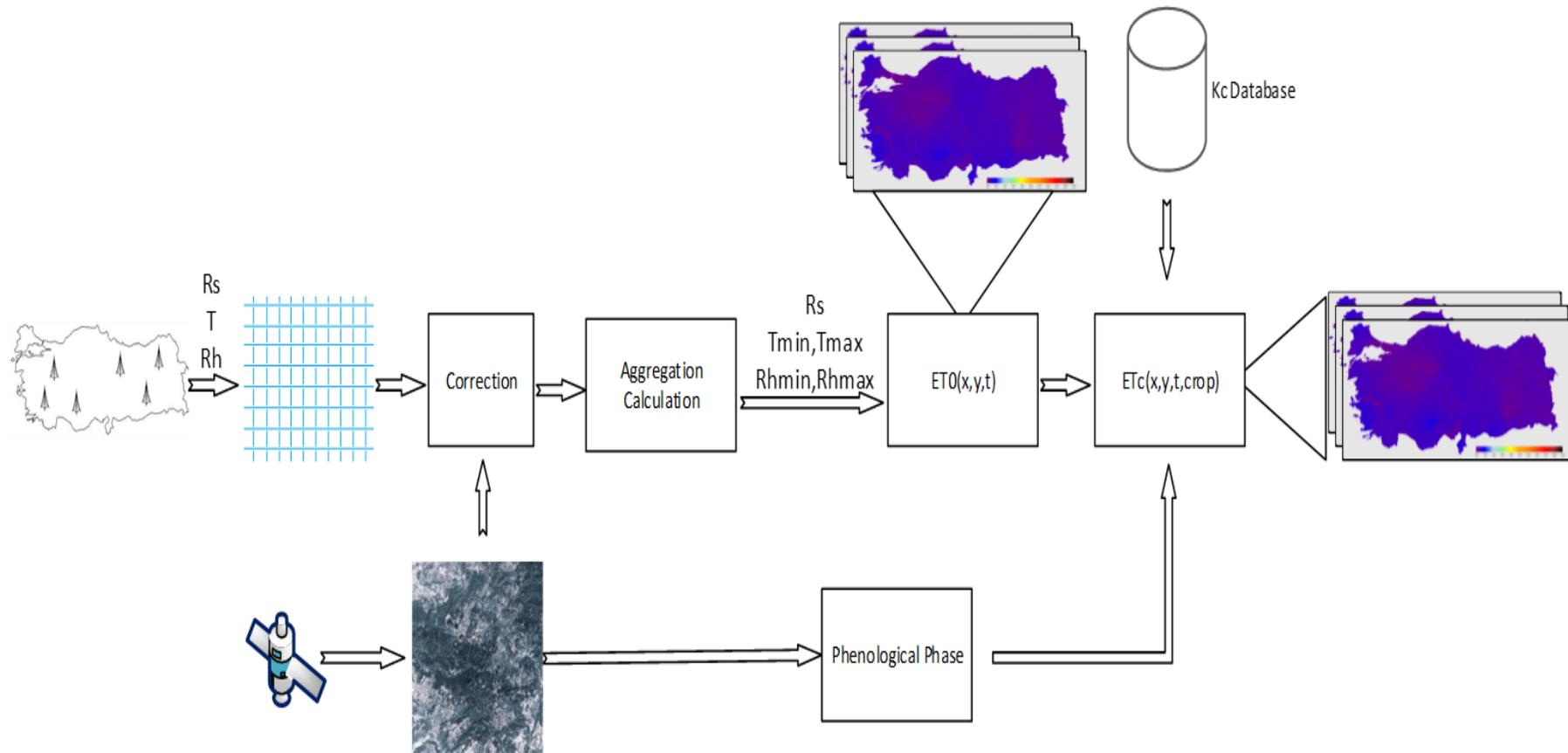
More than 230 parameters are computed at 250m resolution at 30min. time intervals. 35 different data is sampled from each station at 10-30min. intervals



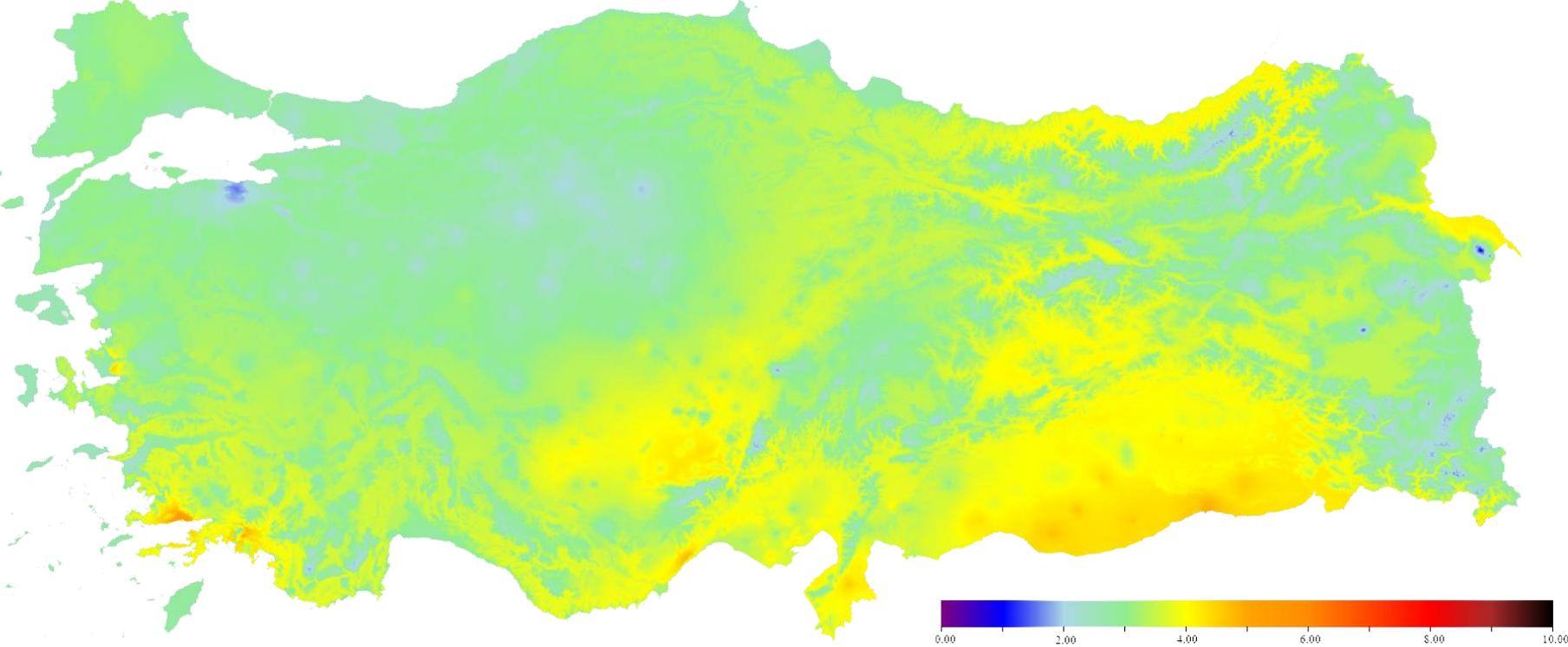
Monitoring of Wind speed over agricultural fields in Real Time



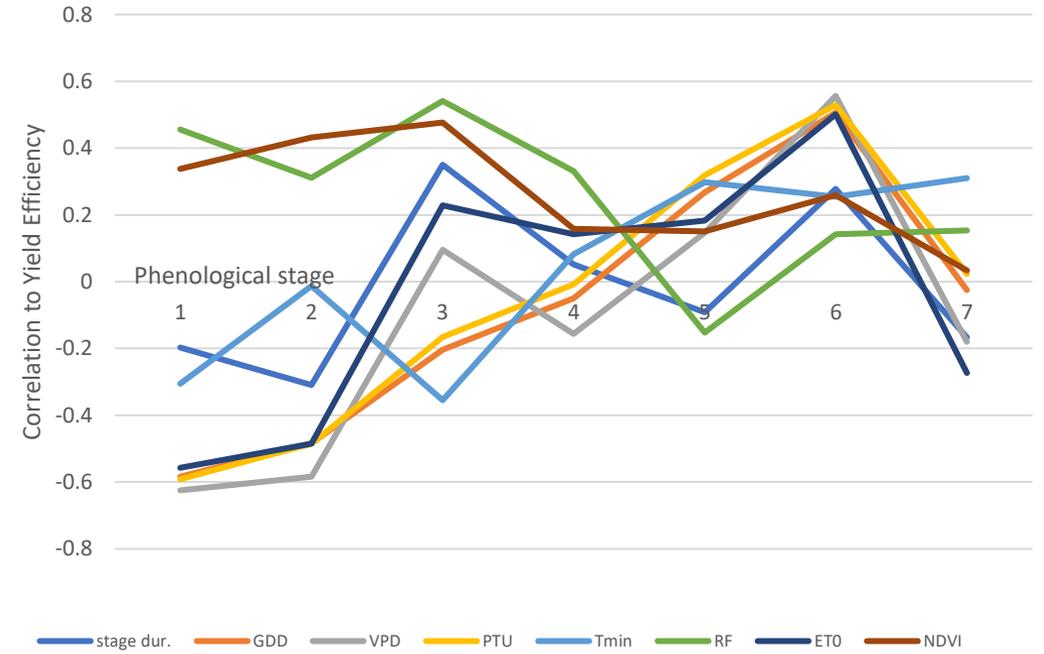
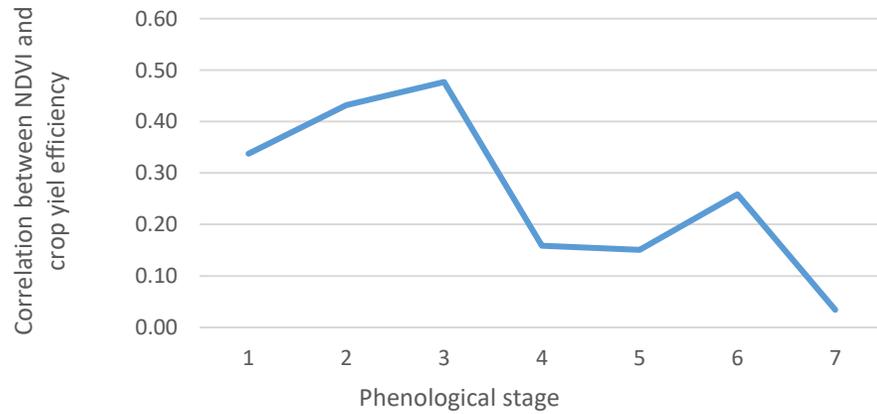
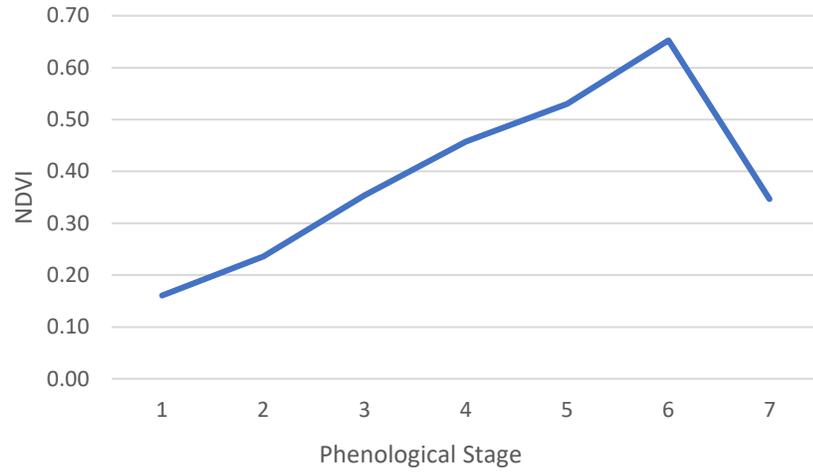
Real Time Spatio Temporal ET Mapping



ET0(mm) - 5.4.2015



Province	Town	Sowing Date	stage1 (days)								stage2 (days)								stage3 (days)								stage4 (days)							
			GDD	VPD	PTU	Tmin	RF	PET	NDVI	GDD	VPD	PTU	Tmin	RF	PET	NDVI	GDD	VPD	PTU	Tmin	RF	PET	NDVI	GDD	VPD	PTU	Tmin	RF	PET	NDVI				
Sanliurfa	Suruc	10/25/2013	25	285.1	26.11	2931	9.54	22.58	52.47	0.1028	61	193.3	24.9	1870	-2.99	40.94	65.68	0.1351	13	52.5	3.09	524	3.3	33.18	12.41	0.341	53	318.9	31.47	3611	-2	61.04	111.13	0.4439
Sanliurfa	Suruc	10/15/2014	17	200.6	12.09	2149	9.55	34.32	41.2	0.0827	10	73.9	8.45	758	4.58	0.16	21.93	0.1208	12	100.6	7.97	1001	7.07	30.62	23.14	0.2036	23	122.1	5.44	1169	3.96	51.54	21.98	0.3617
Sanliurfa	Merkez	11/16/2012	17	124.8	6.41	1216	5.53	8.68	19.91	0.2665	19	52.6	3.57	499	0.7	37.38	16.19	0.306	25	39.5	6.71	375	-2.18	18.93	24.8	0.3254	33	105.9	7.93	1084	0.8	16.49	40.6	0.3448
Diyarbakir	Yenisehir	11/12/2013	18	123.4	6.26	1205	6.03	11.74	17.36	0.1389	61	38.5	10.18	376	-9.62	47.34	37.31	0.2484	29	95.2	14.74	1023	-4.98	28.64	45.15	0.3065	33	208.1	15.45	2448	-0.38	56.34	70.21	0.3646
Gaziantep	Araban	11/6/2012	13	129.0	6.94	1292	8.2	40.56	20.25	0.1648	19	110.2	5.17	1064	4.49	42.32	17.54	0.204	4	14.0	0.54	132	4.96	16.79	2.98	0.2433	89	219.2	23.63	2275	-2.79	152.55	109.34	0.2562
Gaziantep	Araban	11/22/2011	23	36.3	10.76	348	-1.51	35.38	34.16	0.021	17	29.2	4.61	275	0.2	19.1	16.77	0.0989	39	23.5	7.42	228	-3.3	91.3	37.46	0.1767	49	109.2	21.51	1297	-2.72	147.87	112.55	0.3937
Gaziantep	Nurdagi	11/15/2013	39	238.3	21.64	2314	-1.21	19.54	44.7	0.1713	14	66.4	6.33	630	1.62	16.48	16.62	0.2499	28	152.1	11.11	1501	-0.67	24.76	31.14	0.3944	31	248.5	19.93	2719	-0.21	23.62	59.27	0.4805
Gaziantep	Nurdagi	11/7/2014	31	288.7	20.25	2841	4.84	35.42	51.08	0.1768	20	145.1	7.16	1374	2.77	54.76	23.43	0.22	25	80.9	5.56	778	-2.01	112.96	24.17	0.5147	56	361.1	22.24	3947	1.31	208.42	94.49	0.5929
Gaziantep	Nurdagi	11/19/2011	39	82.1	14.95	788	-1.31	97.46	50.21	0.2001	19	16.9	3.16	162	-0.37	77.48	14.78	0.1928	70	92.9	24.65	1081	-4.52	210.78	129.56	0.4198	21	188.7	15.98	2392	3.34	10.98	76.53	0.6194
Gaziantep	Nurdagi	11/10/2012	22	171.3	8.66	1683	4.85	50.91	25.54	0.2041	10	30.2	1.6	287	2.96	81.3	7.38	0.4044	44	71.9	9.7	692	-3.04	128.44	36.47	0.6046	36	159.1	10.7	1693	1.25	129.3	46.08	0.6267
Mardin	Kiziltepe	11/28/2011	29	53.0	12.82	502	-0.71	39.78	45.08	0.1413	17	24.4	4.95	232	0.87	134.7	21.3	0.1824	33	16.0	8.12	163	-5.24	104.82	42.97	0.2236	38	82.9	18.76	971	-2.02	57.91	102.64	0.2947
Mardin	Kiziltepe	12/5/2012	21	56.1	4.35	531	-0.35	44.66	21.95	0.1482	15	15.2	4.62	145	-1.67	33.89	17.16	0.2545	25	50.0	6.36	500	-3.8	50.53	30.46	0.3608	31	110.5	9.69	1197	0.43	42.97	53.45	0.2157
Gaziantep	Oguzeli	11/16/2011	15	43.0	5.67	423	-1.51	39.24	20.43	0.1924	14	16.5	5.83	157	-1.36	28.42	18.28	0.2528	47	48.2	9.46	460	-4.09	147.88	42.04	0.4632	43	32.0	13.59	360	-3.23	121.18	72.09	0.6175
Gaziantep	Oguzeli	12/1/2012	19	46.3	3.5	442	-1.54	79.08	14.44	0.1508	14	33.1	3.34	314	0.04	32.04	11.3	0.2626	31	47.2	6.6	471	-2.97	84.35	30.39	0.3687	40	189.1	15.69	2091	0.34	90.49	76.1	0.4573
Gaziantep	Oguzeli	11/19/2014	22	129.2	5.78	1248	3.9	68.5	21.44	0.1402	17	64.4	3.32	611	2.29	17.76	13.39	0.1766	33	66.9	6.22	652	-3.73	97.66	28.8	0.4378	76	390.9	25.64	4609	-0.62	189.88	142.47	0.5197
Sanliurfa	Akcakale	12/12/2012	18	41.2	4.3	391	-0.58	21.44	19.03	0.2017	19	7.7	4.9	74	-3.14	30.15	21.5	0.2788	23	83.3	4.68	850	1.59	16.42	26.93	0.2827	23	80.1	7.96	881	0.27	44.14	44.77	0.4431
Sanliurfa	Akcakale	11/11/2013	40	193.4	18	1904	-3.75	19.76	51.21	0.1292	22	53.3	9.83	509	1.09	17	26.34	0.1501	36	112.5	14.41	1146	-2.89	24.24	51.48	0.1796	20	163.0	12.76	1818	3.89	45.98	48.52	0.3974
Sanliurfa	Siverek	11/14/2012	13	107.2	5.29	1048	7.91	2.96	15.66	0.143	18	74.2	4.16	706	3.63	24.73	15.39	0.2166	26	44.4	6.1	419	-0.03	22.76	21.66	0.4783	56	155.9	15.04	1642	-2.86	30.99	75.14	0.5046
Sanliurfa	Ceylanpin.	11/10/2014	19	143.1	12.09	1420	4.68	48.46	40.4	0.1284	6	27.5	1.34	264	4.95	0.86	5.13	0.1585	11	76.4	3.88	726	6.28	23.6	15.65	0.356	75	203.6	21.95	2068	-2.98	114.16	101.78	0.5374
Sanliurfa	Ceylanpin.	10/25/2013	25	304.1	28.61	3123	10.07	23.94	60.74	0.1348	25	111.5	11	1084	-2.97	23.14	34.32	0.2153	27	66.6	13.71	635	-1.13	11.92	37.32	0.3801	39	143.9	18.5	1479	-2.62	29.26	65.35	0.5464
Sanliurfa	Merkez	11/19/2011	20	38.6	9.7	376	-2.55	10.46	30.25	0.1853	25	48.2	7.8	456	0.85	42.56	28.46	0.2888	39	28.0	8.38	273	-3.51	73.76	42.76	0.3655	48	108.2	22.04	1278	-1.91	122.44	117.59	0.416
Sanliurfa	Akcakale	11/18/2014	12	58.8	3.48	577	3.92	40.96	13.18	0.1049	12	76.4	3.06	731	3.68	17.06	13.52	0.2093	18	57.0	3.62	541	1.91	18.4	15.28	0.2355	66	157.6	16.94	1641	-3.98	112.22	85.87	0.5015
Sanliurfa	Akcakale	11/1/2011	11	55.5	5.99	570	-0.21	21.3	20.72	0.2203	16	61.2	6.65	606	-2.77	16.8	26.23	0.3607	45	80.7	15.62	770	-1.93	46.82	55.5	0.4055	45	26.8	12.05	280	-4.52	103.26	65.59	0.4503
Sanliurfa	Eyyubiye	10/10/2013	41	515.8	54.27	5448	9.44	22.68	107.53	0.0833	58	173.2	23.31	1673	-3.32	45.26	63.56	0.1804	17	63.8	5.23	636	1.1	34.26	19.18	0.2776	33	172.7	18.5	1890	-2.18	20.1	60.84	0.3962
Sanliurfa	Eyyubiye	11/7/2014	14	124.4	11.49	1247	6.36	6.98	29.31	0.1304	8	33.2	1.86	323	4.42	46.32	7.75	0.1635	9	47.3	2.18	453	3.39	0.92	8.06	0.2193	81	214.4	18.15	2116	-3.48	186.48	85.14	0.4814
Diyarbakir	Sur	1/22/2014	20	24.6	6.68	247	-5.83	22.06	22.04	0.2266	15	72.8	8.44	786	0.89	17.16	24.03	0.258	20	116.9	6.88	1337	2.61	55.9	32.78	0.2894	12	88.1	7.31	1066	-0.46	1.7	30.4	0.4248
Diyarbakir	Bismil	11/16/2011	45	63.2	16.22	609	-2.9	83.96	53.67	0.3225	37	16.2	6.81	155	-5.04	112.64	34.15	0.4573	44	41.9	16.07	482	-2.42	87.02	87.15	0.5297	16	99.1	10.05	1230	2.06	20.42	50.3	0.5684
Diyarbakir	Bismil	11/26/2013	50	32.8	7.69	314	-10.14	35.24	28.35	0.0134	33	54.9	11.83	559	-5.28	26.44	37.11	0.3486	12	65.5	6.28	714	2.26	15.9	20.8	0.4898	17	107.7	6.28	1235	3.14	49.56	29.42	0.6142
Diyarbakir	Bismil	12/8/2012	27	45.0	6.48	422	-1.08	86.1	24.48	0.3534	30	40.5	5.83	401	-6	124.56	27.12	0.3601	25	71.3	6.39	760	0.62	62.34	32.17	0.4242	29	151.6	15.2	1778	0.26	20.33	75.55	0.5181
Sanliurfa	Hilvan	11/27/2011	29	51.3	11.37	485	-2.97	9.2	33.86	0.1497	19	26.7	4.33	254	0.82	11.7	17.92	0.2066	56	37.0	15.22	394	-3.51	135.4	82.77	0.3378	25	115.3	15.39	1402	-1.79	54.6	79.34	0.3863
Sanliurfa	Hilvan	12/9/2014	19	73.3	4.34	691	2.1	36.22	16.89	0.1485	15	31.8	2.18	301	-3.58	61.22	11.98	0.1575	29	71.7	8.75	725	-2.31	39.94	36.26	0.331	41	150.8	13.16	1721	0.04	95.68	69.71	0.3979
		Average:	24.02	126.8	11.44	1274	1.88	35.24	34.00	0.16	22.20	58.6	6.27	572	-0.29	40.41	23.02	0.24	27.44	62.9	7.88	642	-0.78	51.41	33.98	0.35	43.00	157.9	15.49	1758	-0.71	76.80	74.49	0.46
		Std.Dev.:	9.51	106.9	9.35	1118	4.51	26.03	19.12	0.07	12.89	42.5	4.82	412	3.12	35.44	13.12	0.08	13.13	32.9	4.84	349	2.72	46.13	22.84	0.11	20.03	81.6						



Variation of the correlations between the agro-meteorological indices and the phenological stage durations.

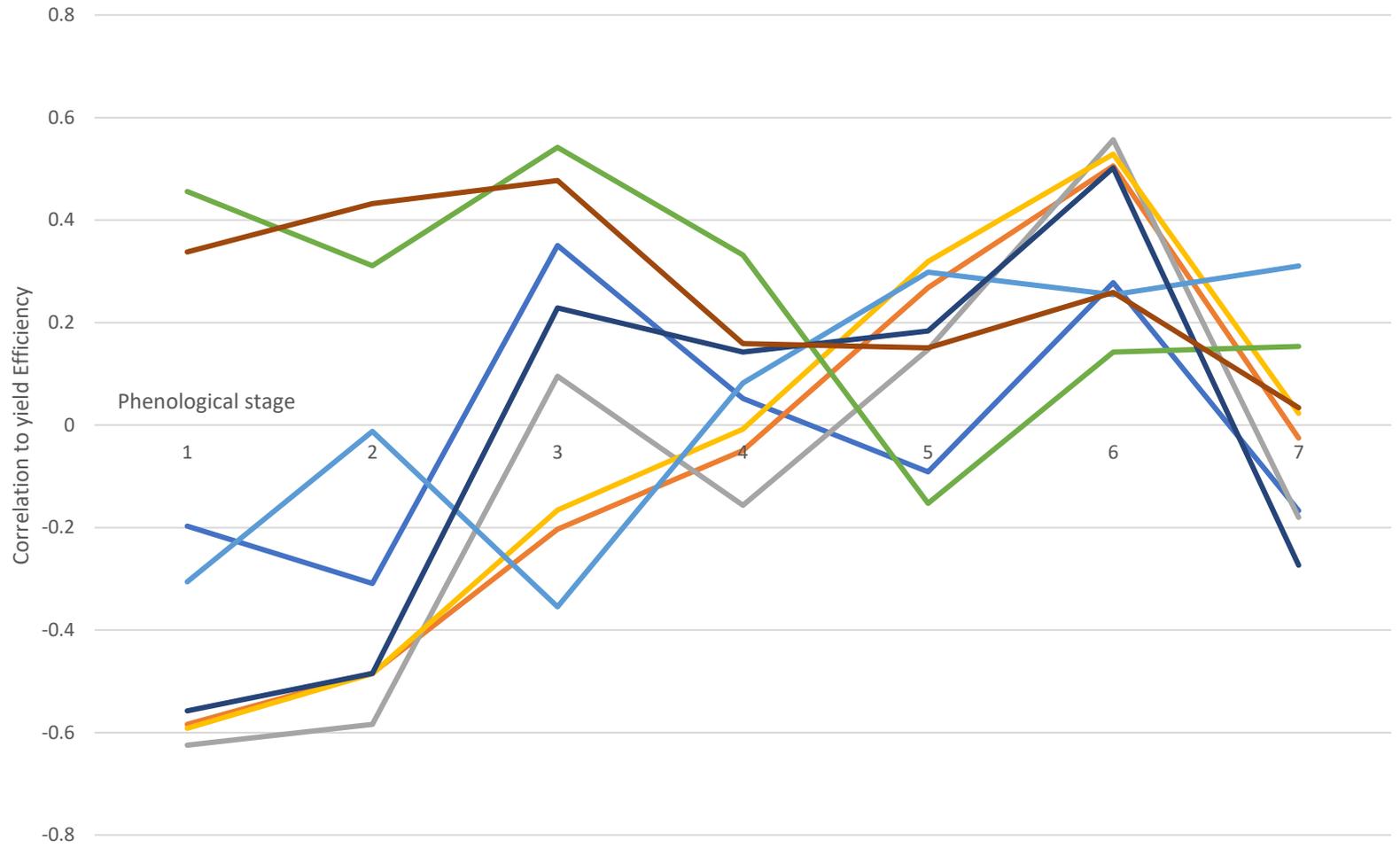
PS stage	Duration (d)	R^2
PS1	$0.8967 \times ET_0 - 0.9984 \times T_{\min} + 7.0258$	0.818
PS2	$2.2669 \times VPD - 2.0421 \times T_{\min} + 7.3934$	0.840
PS3	$1.4718 \times VPD - 1.5048 \times T_{\min} + 0.1308 \times \text{Prec.} + 0.6922$	0.935
PS4	$0.0724 \times GDD + 0.1198 \times \text{Prec.} - 4.7053$	0.915
PS5	$0.1479 \times ET_0 + 0.0209 \times GDD + 1.2620$	0.967
PS6	$0.0035 \times PTU + 0.9313$	0.944
PS7	$0.0221 \times GDD + 0.0699 \times ET_0 + 0.5162$	0.979

How can we choose proper parameters and features in real-time Machine Learning ?

How can we avoid overfitting or underfitting problems?

PHENOLOGICAL STAGE SEGMENTED PARAMETERS' CORRELATION TO WHEAT YIELD EFFICIENCY

	Stage1	Stage2	Stage3	Stage4	Stage5	Stage6	Stage7
St.D.	-0.1970	-0.3093	0.3505	0.0517	-0.0913	0.2778	-0.1670
GDD	-0.5842	-0.4844	-0.2035	-0.0498	0.2678	0.5059	-0.0253
VPD	-0.6244	-0.5843	0.0952	-0.1566	0.1474	0.5566	-0.1802
PTU	-0.5919	-0.4851	-0.1661	-0.0081	0.3192	0.5288	0.0230
Tmin	-0.3057	-0.0125	-0.3547	0.0821	0.2985	0.2543	0.3102
RF	0.4559	0.3109	0.5418	0.3317	-0.1529	0.1423	0.1534
ET0	-0.5575	-0.4849	0.2285	0.1425	0.1834	0.5019	-0.2733
NDVI	0.3377	0.4322	0.4771	0.1588	0.1506	0.2588	0.0339



stage dur. GDD VPD PTU Tmin RF ETO NDVI

	VPD6	VPD12	ARF1234	PTU56	Tmin7	Yield Eff. (kg/da)
	6.07	51.01	157.74	3621.39	10.03	38
	7.94	20.54	116.64	2690.08	2.51	171
	13.03	9.98	81.48	5798.1	9.59	144
	20.32	16.44	144.06	5929.16	9.13	271
	29.73	12.11	252.22	8066.38	11.39	284
	22.66	15.37	293.65	9328.96	13.42	256
	13.29	27.97	84.4	8049.57	12.61	136
	13.7	27.41	411.56	9282.93	13.77	303
	15.7	18.11	396.7	6723.64	12.62	270
	14.18	10.26	389.95	7325.63	13.17	349
	12.69	17.77	337.21	6396.51	11.7	226
	22.76	8.97	172.05	7732.59	10.17	292
	13.86	11.5	336.72	5759.22	12.1	268
	24.28	6.84	285.96	7509.15	11.66	270
	16.42	9.1	373.8	6729.33	12.69	310
	14.83	9.2	112.15	5331.35	11.13	203
	11.31	27.83	106.98	5140.28	10.82	38
	26.43	9.45	81.34	9651.8	12.26	230
	14.64	13.43	187.08	5723.36	10.49	160
	17.47	39.61	88.26	7665	11.57	89
	21.38	17.5	249.22	5737.32	12.06	223
	11.27	6.54	188.64	5545.36	11.64	160
	20.83	12.64	188.18	5772.36	11.72	254
	10.75	77.58	122.3	4583.42	10.23	38
	6.47	13.35	240.7	7819.03	11.56	163
	19.07	15.12	96.82	7315.67	8.73	233
	20.58	23.03	304.04	8779.33	12.76	248
	19.76	19.52	127.14	8426.76	8.82	232
	21.95	12.31	293.33	8384.5	12.02	317
	13.03	15.7	210.9	7231.78	12.06	247
	19.13	6.52	233.06	6245.11	11.37	189
Average:	16.630	18.797	214.977	6783.712	11.155	213.290
St.Dev.:	5.638	14.405	103.087	1625.187	2.028	81.659
R:	0.557	-0.652	0.646	0.521	0.310	1.000

$$YE = 119.52 + 4.418 \cdot VPD6 - 1.908 \cdot VPD12 + 0.501 \cdot ARF1234 + 0.0123 \cdot PTU56 - 12.116 \cdot Tmin$$

	VPD6	VPD12	ARF 1234567	Actual Yield Eff.	Predicted Yield Eff.	Error
	6.07	51.01	180.26	38	40.59	-2.59
	7.94	20.54	374.28	171	191.99	-20.99
	13.03	9.98	180.3	144	156.25	-12.25
	20.32	16.44	202.06	271	201.13	69.87
	29.73	12.11	329.08	284	325.21	-41.21
	22.66	15.37	335.54	256	277.41	-21.41
	13.29	27.97	190.5	136	131.18	4.82
	13.7	27.41	495.12	303	270.60	32.40
	15.7	18.11	444.68	270	277.01	-7.01
	14.18	10.26	496.85	349	304.24	44.76
	12.69	17.77	359.69	226	220.54	5.46
	22.76	8.97	270.85	292	260.33	31.67
	13.86	11.5	436.44	268	273.11	-5.11
	24.28	6.84	351.35	270	309.61	-39.61
	16.42	9.1	402.06	310	278.25	31.75
	14.83	9.2	196.01	203	176.07	26.93
	11.31	27.83	139.62	38	96.12	-58.12
	26.43	9.45	160.76	230	233.77	-3.77
	14.64	13.43	278.72	160	204.39	-44.39
	17.47	39.61	170.92	89	128.82	-39.82
	21.38	17.5	284.38	223	242.75	-19.75
	11.27	6.54	274.16	160	192.88	-32.88
	20.83	12.64	280.58	254	246.00	8.00
	10.75	77.58	193.68	38	30.17	7.83
	6.47	13.35	358.5	163	188.09	-25.09
	19.07	15.12	162.48	233	177.82	55.18
	20.58	23.03	343.9	248	254.58	-6.58
	19.76	19.52	190.5	232	187.06	44.94
	21.95	12.31	386.41	317	300.90	16.10
	13.03	15.7	242.06	247	173.84	73.16
	19.13	6.52	315.28	189	261.29	-72.29
Average:	16.630	18.797	291.194	213.290	213.290	0.000
St.Dev.:	5.638	14.405	101.956	81.659	73.281	36.029
R:	0.557	-0.652	0.648	1.000	0.897	

$$YE = 10.24 + 6.367 \cdot VPD6 - 1.739 \cdot VPD12 + 0.446 \cdot ARF1234567$$

Minimum (COV/R²) → Maximum (Estimation Confidence Factor)

$$YE = 119.52 + 4.418 \cdot VPD6 - 1.908 \cdot VPD12 + 0.501 \cdot ARF1234 + 0.0123 \cdot PTU56 - 12.116 \cdot Tmin$$

$$\mathbf{R^2 = 0.83}$$

$$YE = 10.24 + 6.367 \cdot VPD6 - 1.739 \cdot VPD12 + 0.446 \cdot ARF1234567$$

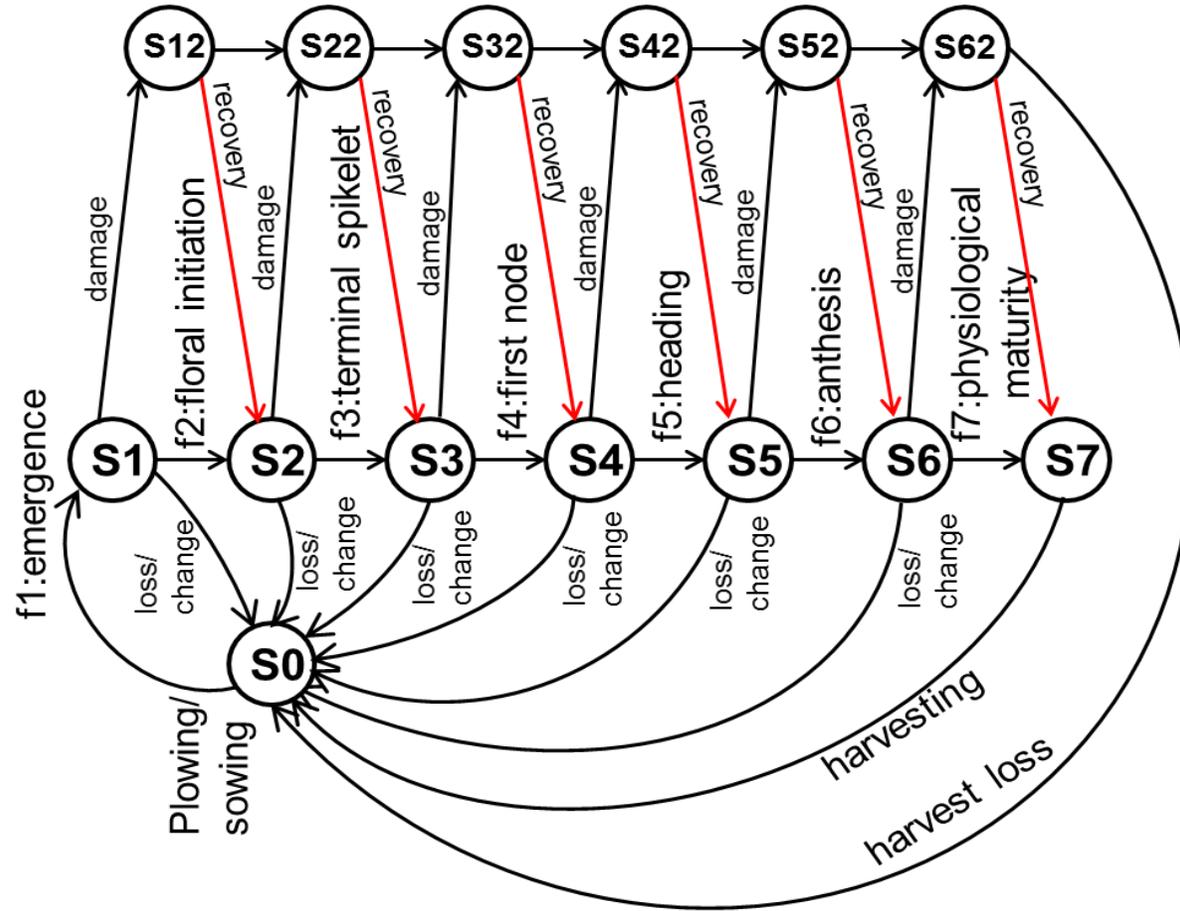
$$\mathbf{R^2 = 0.81}$$

$$YE = -34.163 + 5.187 \cdot VPD6 - 2.088 \cdot VPD12 + 0.392 \cdot ARF1234567 + 31.58 \cdot NDVI1234567$$

$$\mathbf{R^2 = 0.83}$$

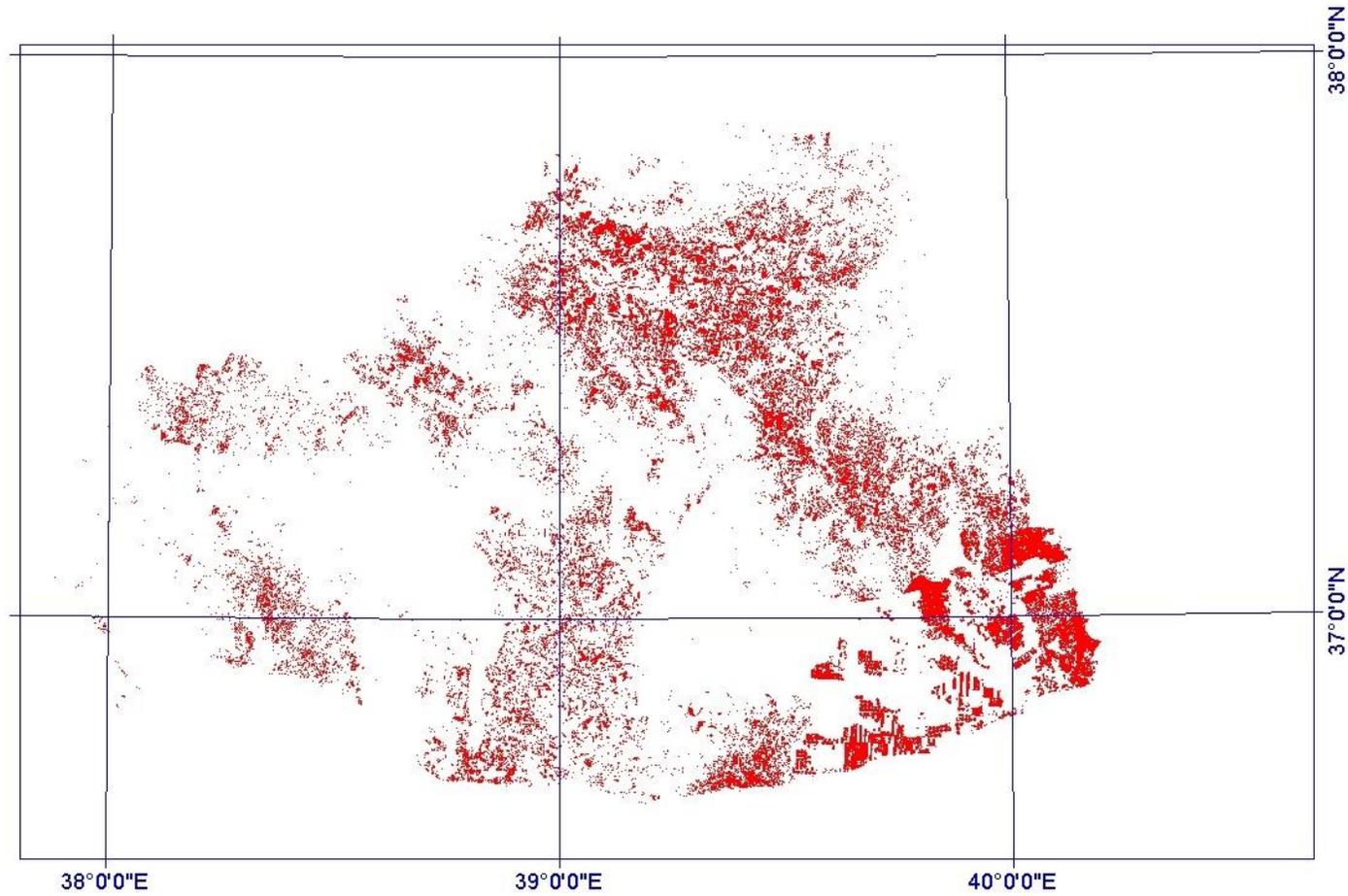
$$YE = -8.046 + 5.559 \cdot VPD6 - 1.896 \cdot VPD12 + 0.386 \cdot ARF1234567 + 147.76 \cdot NDVI3$$

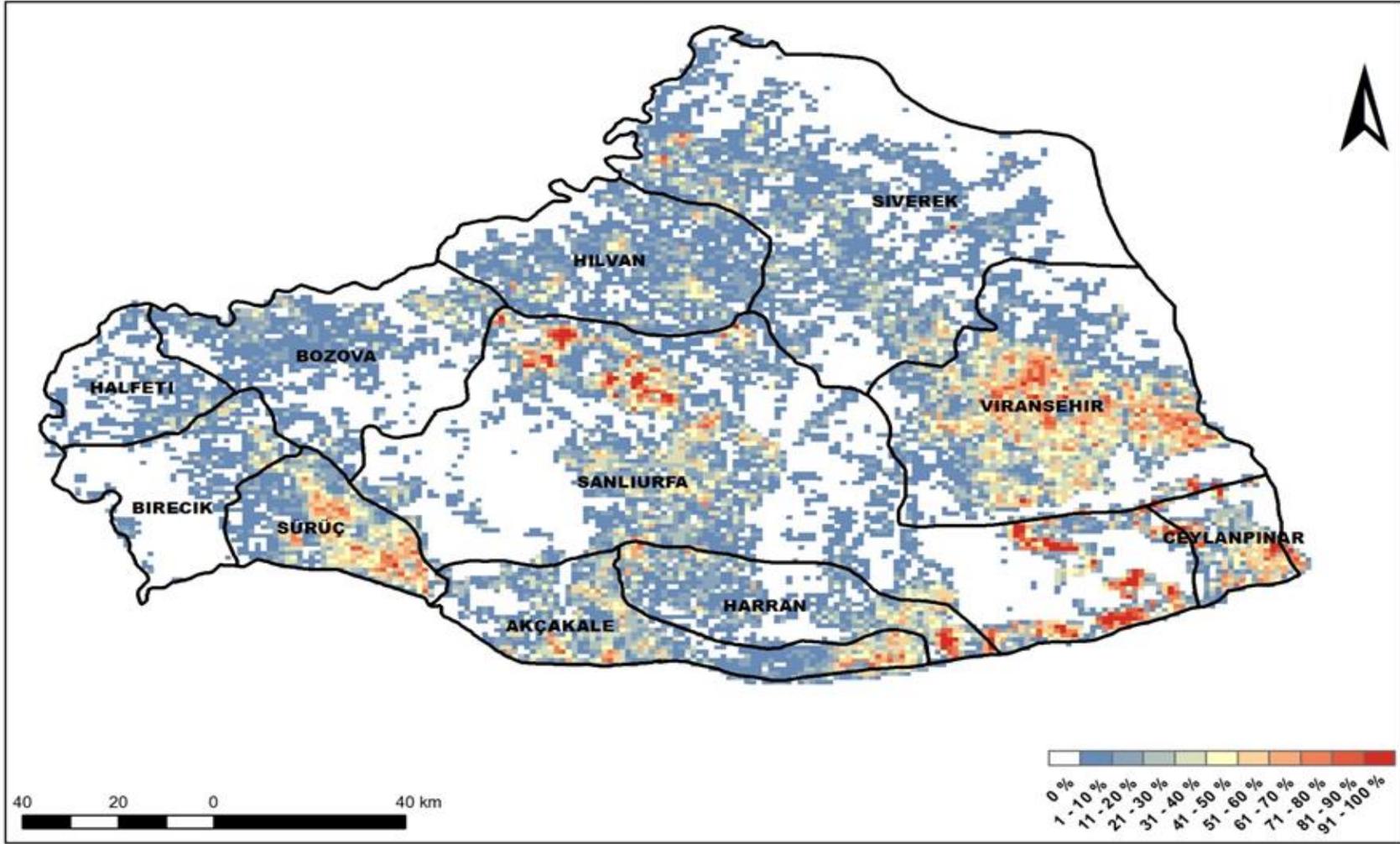
$$\mathbf{R^2 = 0.84}$$



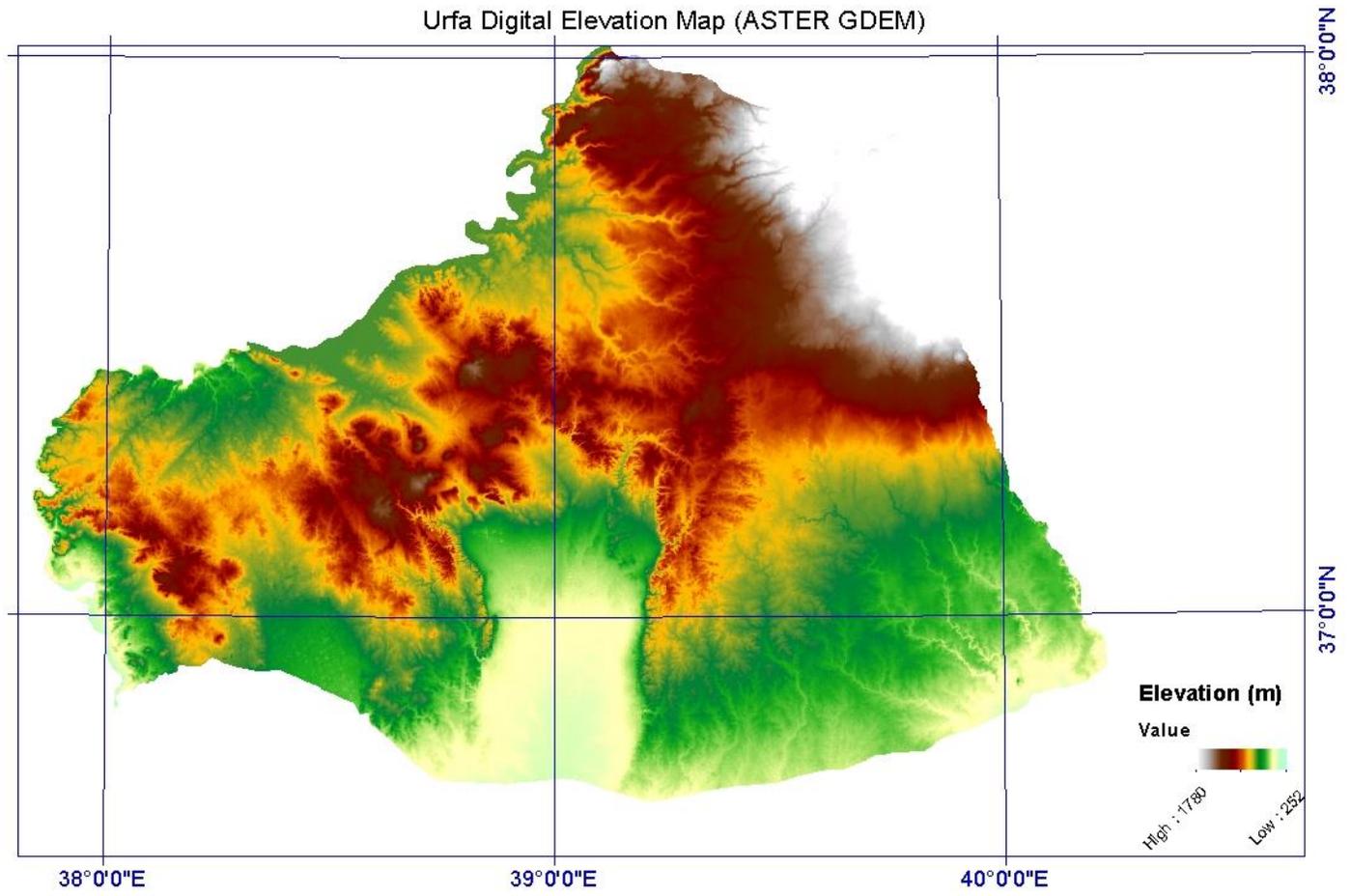
State machine for more complex roadmaps to possible permanent damage

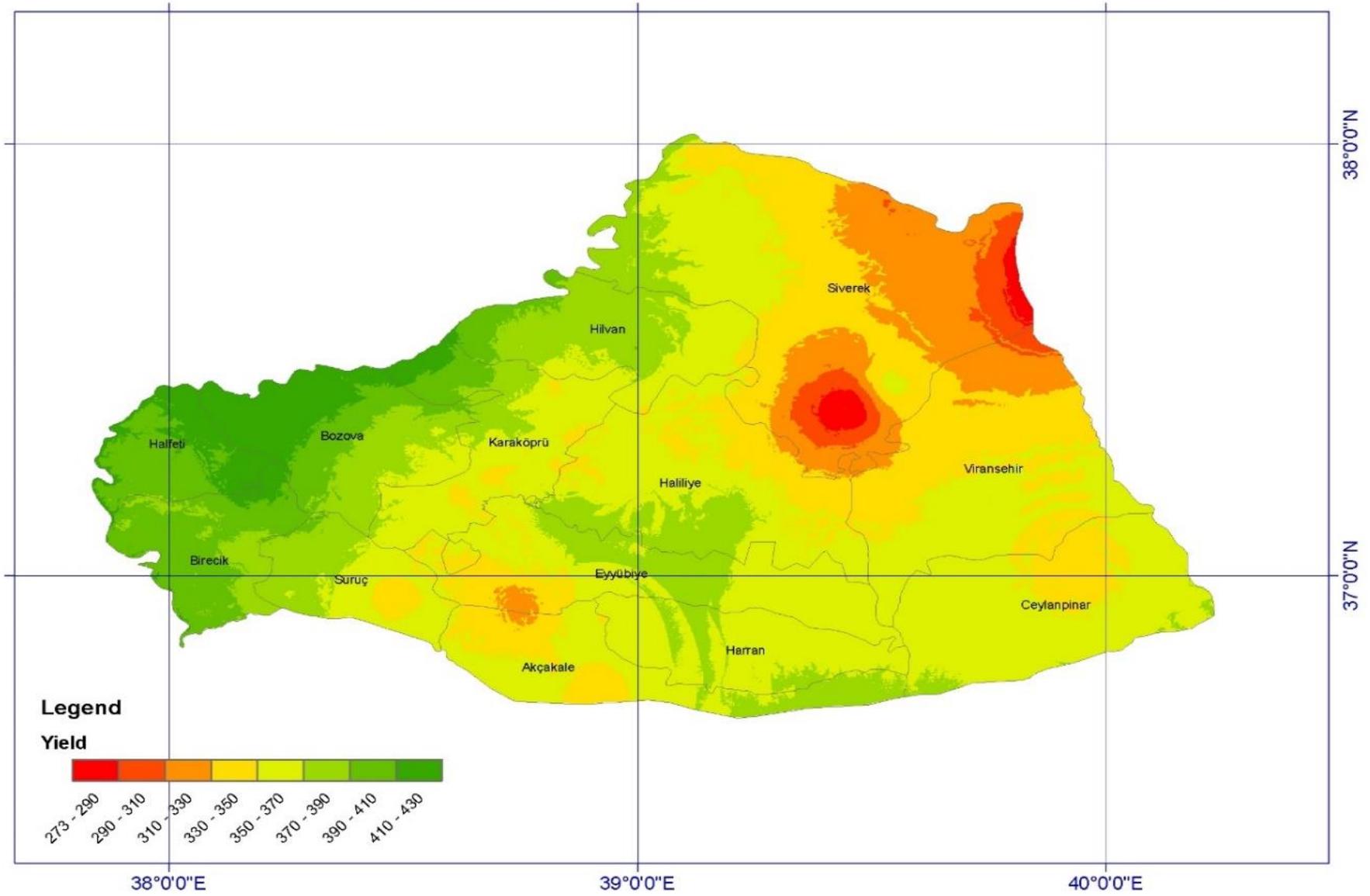
Winter wheat detected area in Şanlıurfa province (2015)

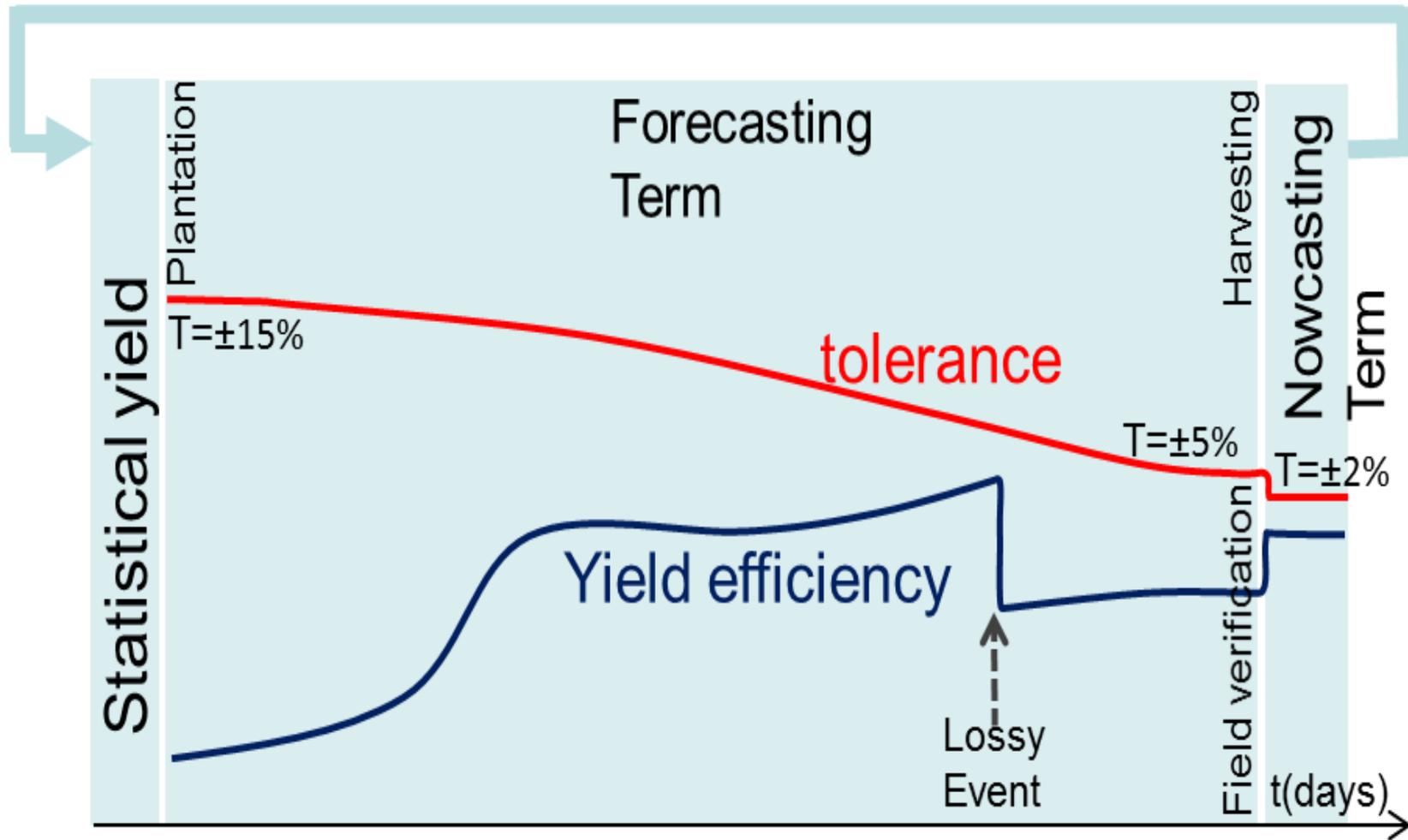




Urfa Digital Elevation Map (ASTER GDEM)



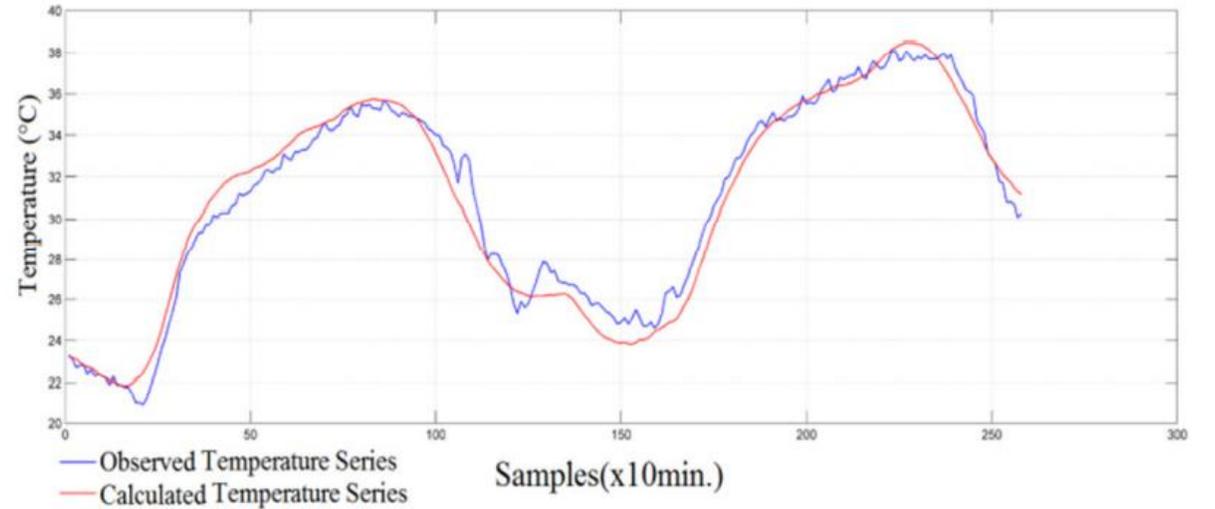
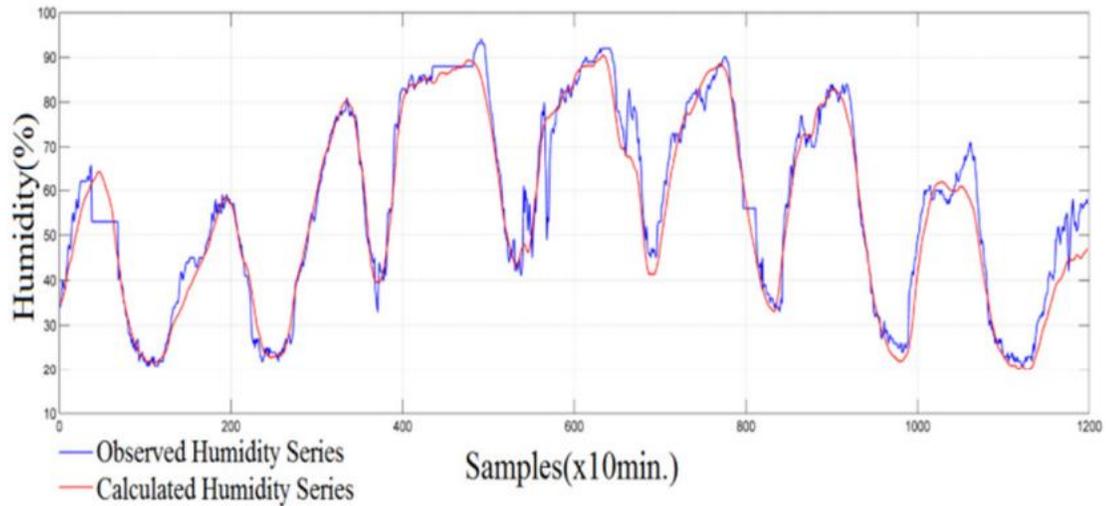
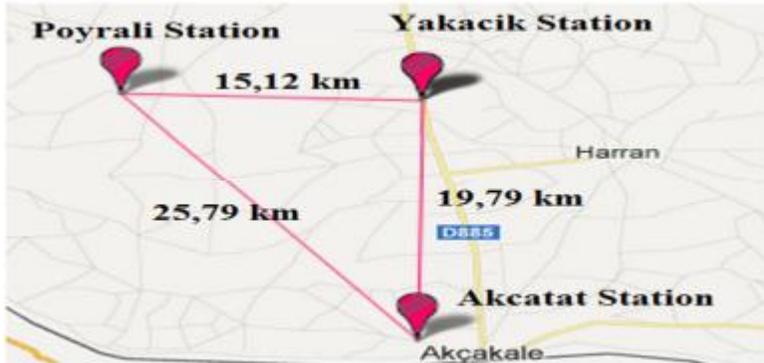




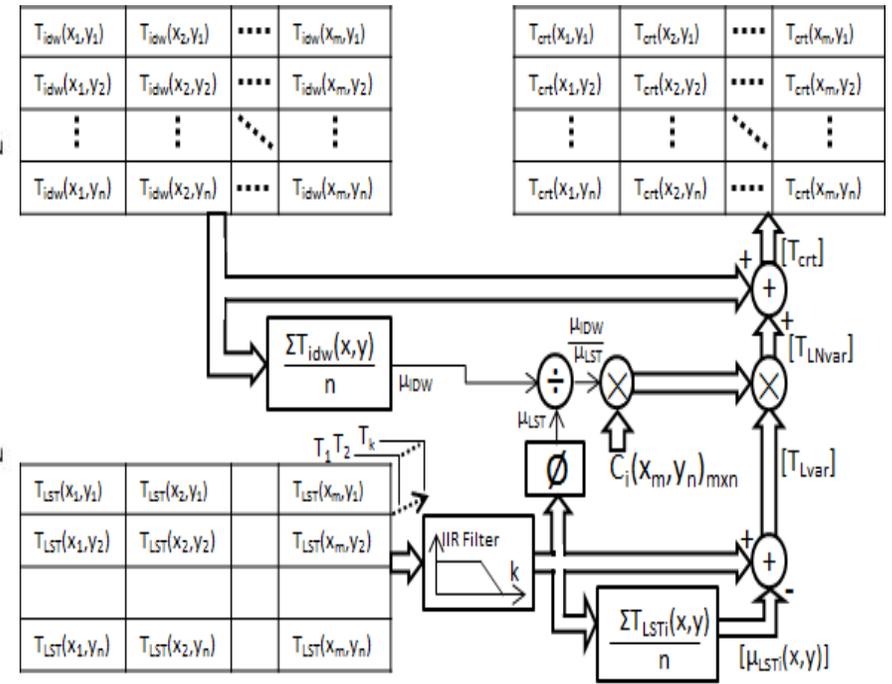
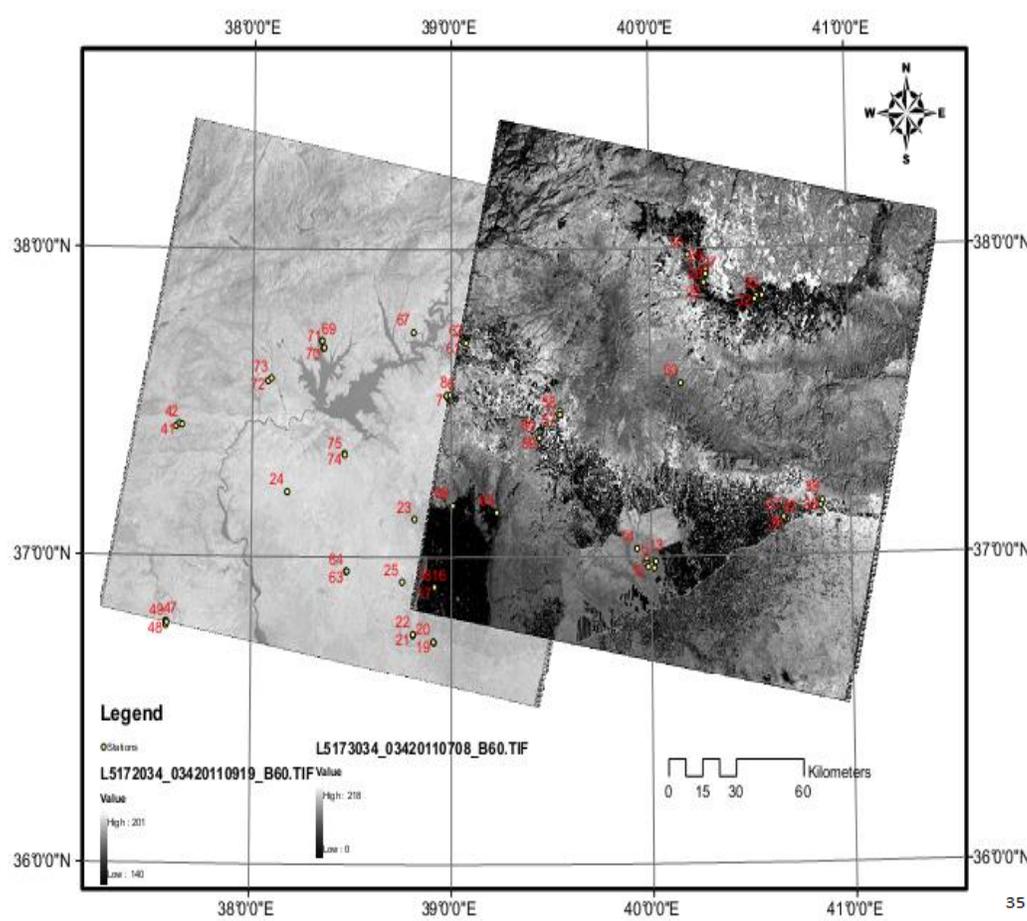
$$YE = \mu_{YE} \mp Zc \cdot \frac{\sigma_{YE}}{\sqrt{n}}$$

QoS Management in Twinning Process

a) Real-time missing data reconstruction example:

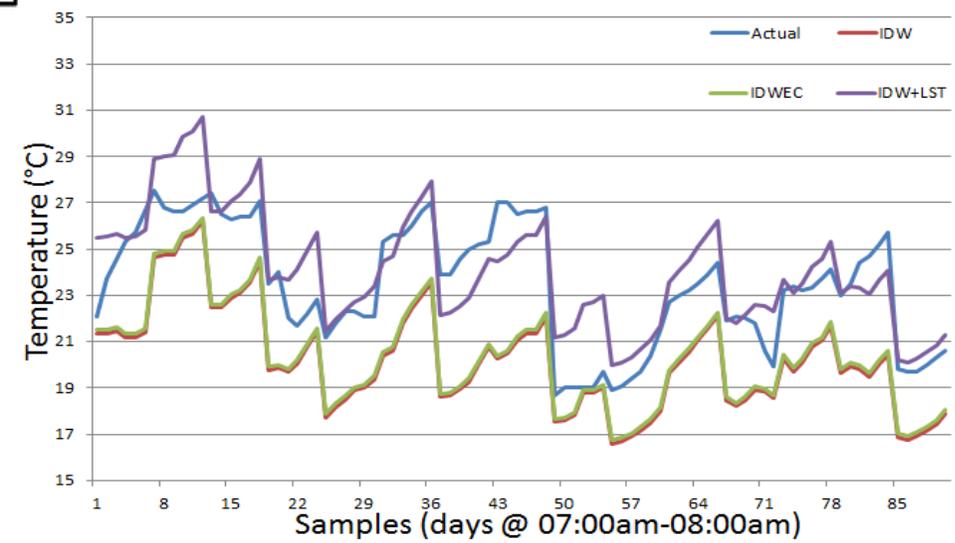


b) Accuracy & Tolerance Management

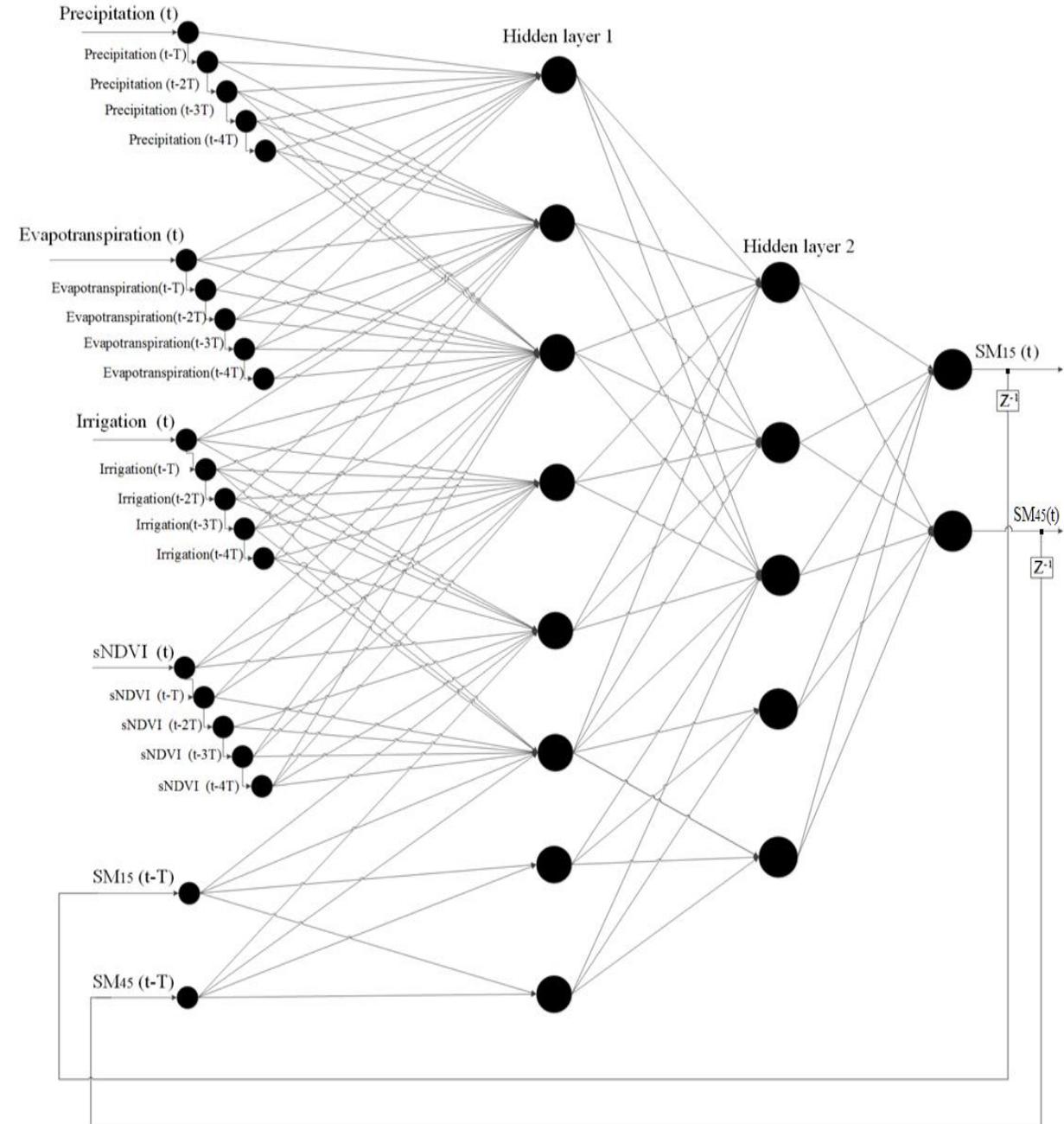
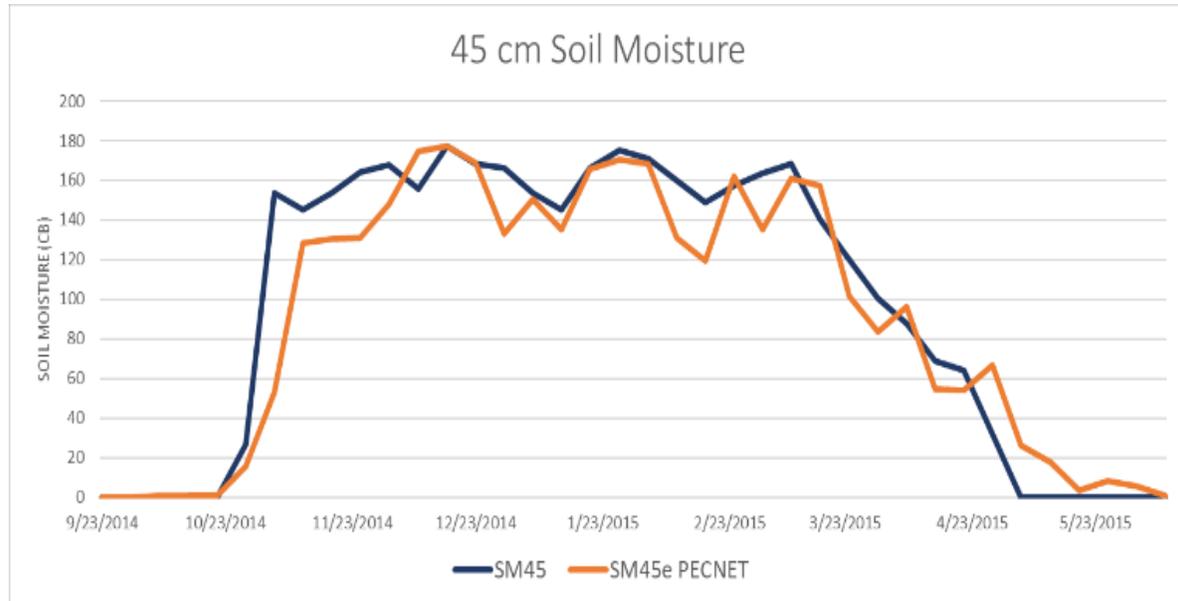
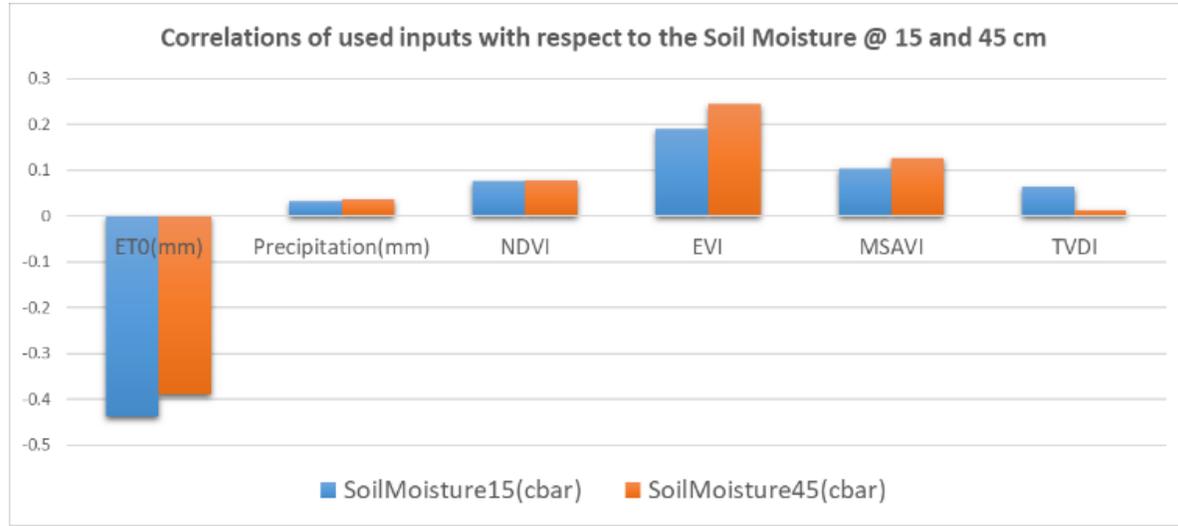


Adaptive model based quality improvement of spatio-temporal fusion data

Example: 250m fused temperature Data interpolation MSE is less than 1°C



Performance improvement with small data in ML? → PECNET

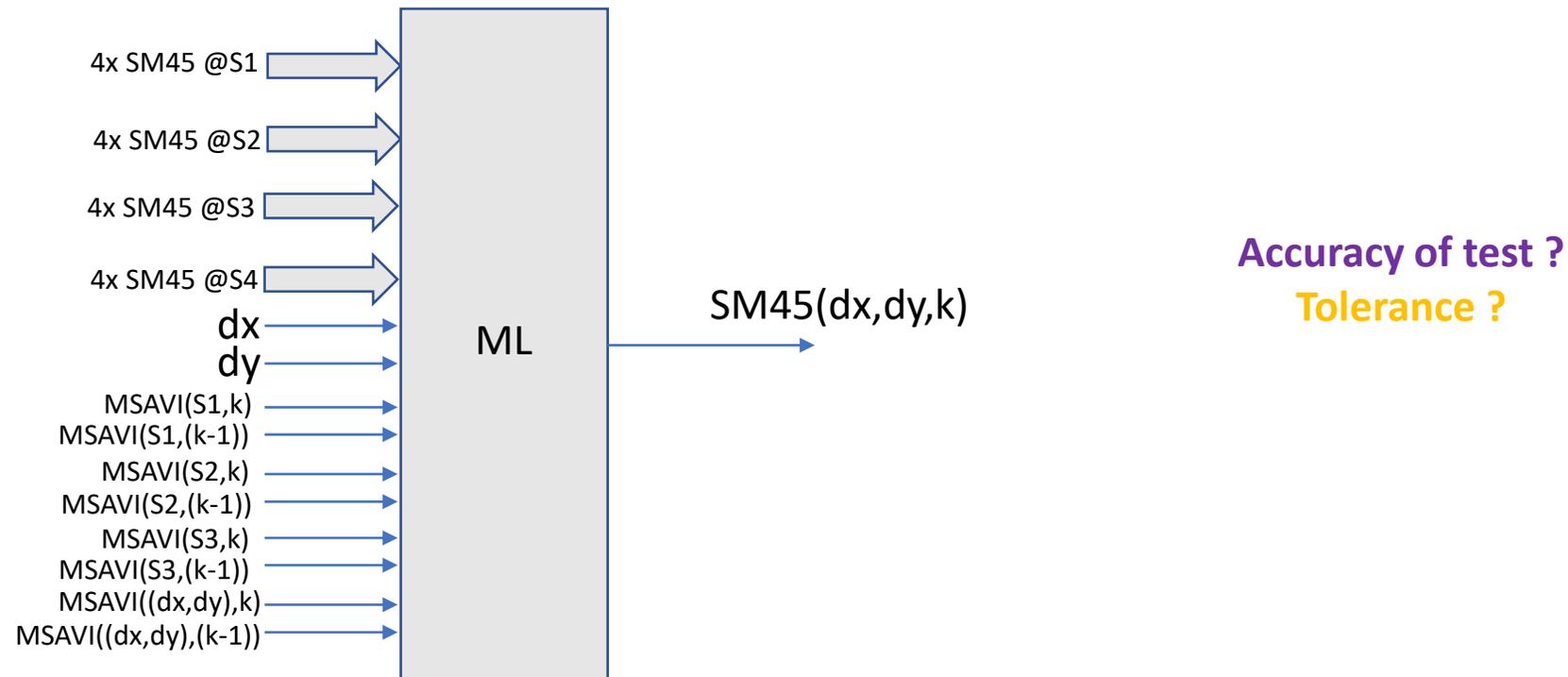


Problem in Spatio Temporal Data Fusion:

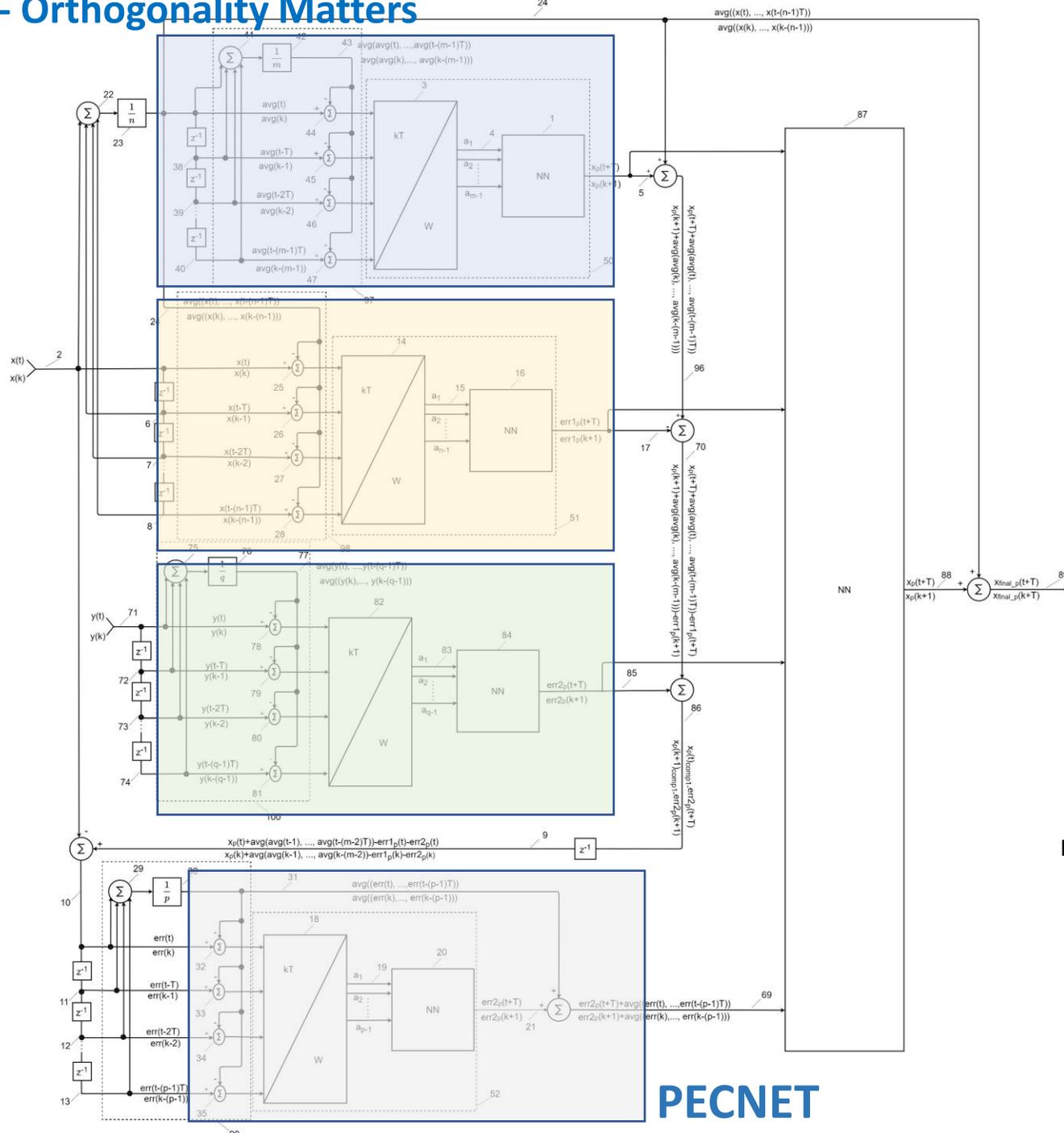
Over Fitting (How to choose the proper time window and the data type)

Over Fitting (How to process, ML type and Hyperparameter selection)

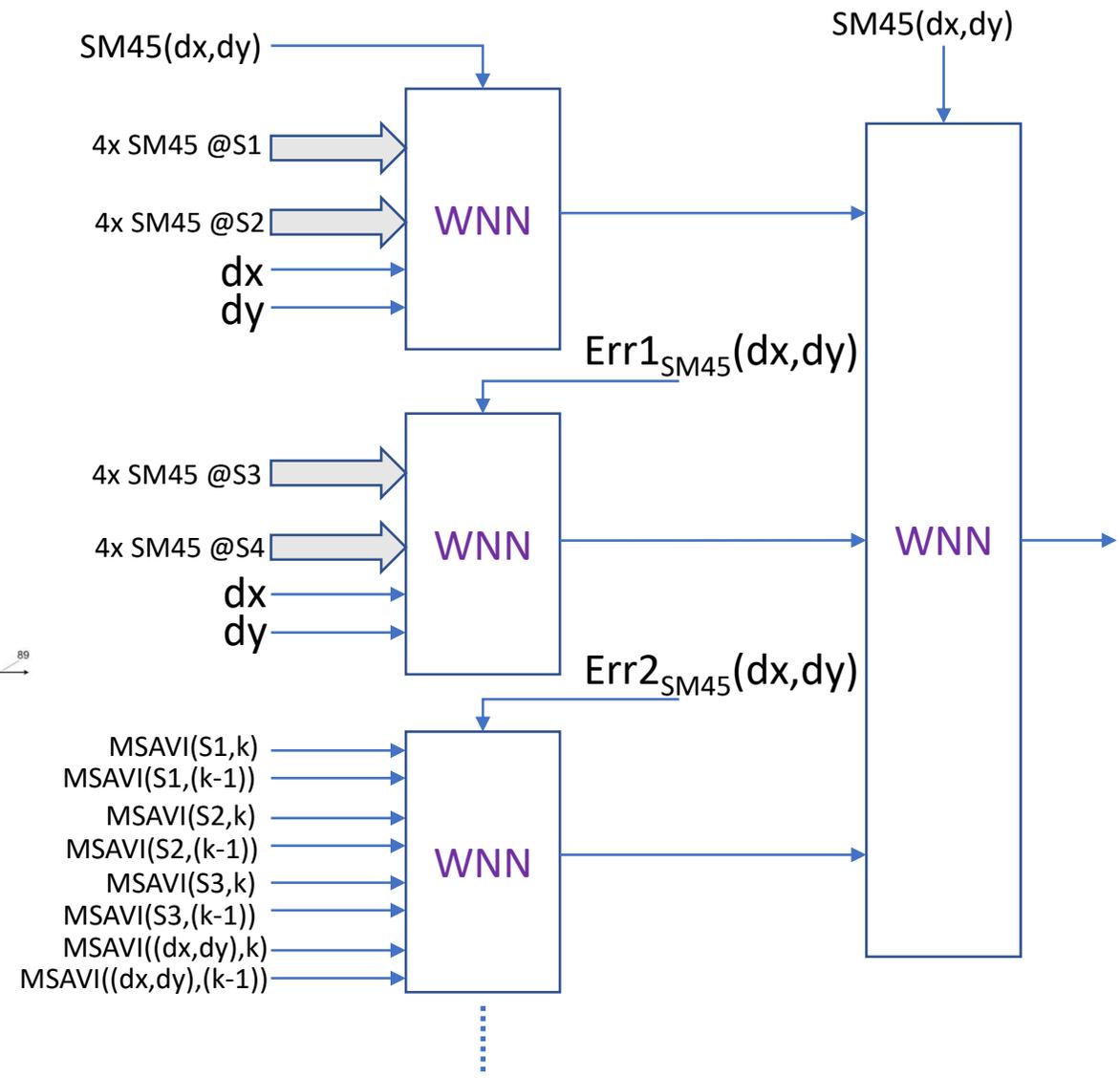
Over Fitting (How to struggle against non-linearity, uncertainty under and restricted labeled data)



3- Orthogonality Matters



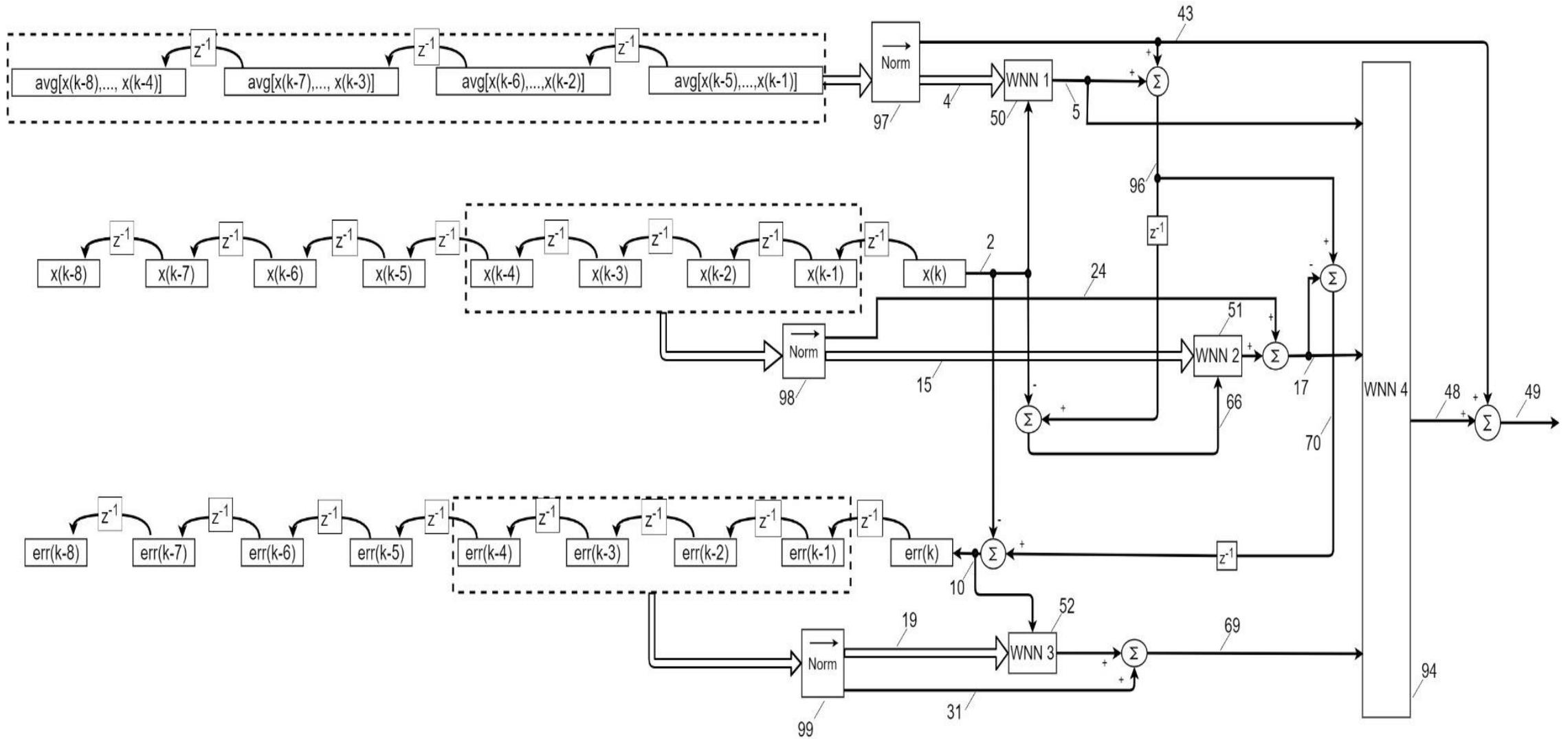
PECNET

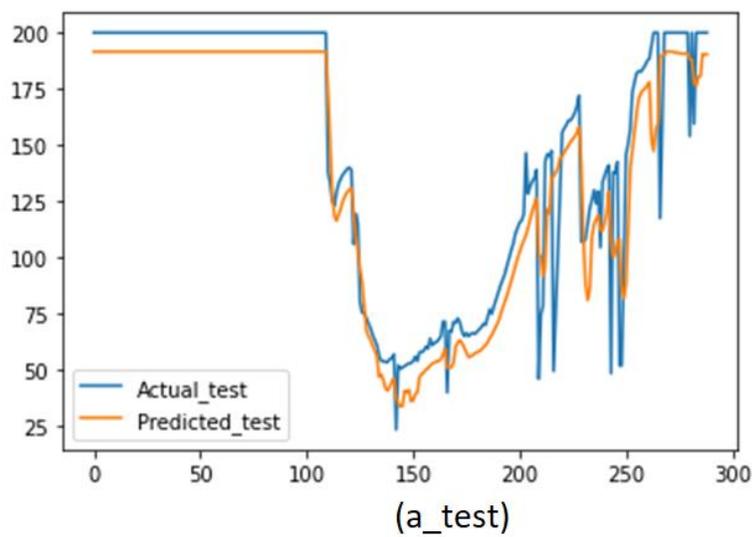
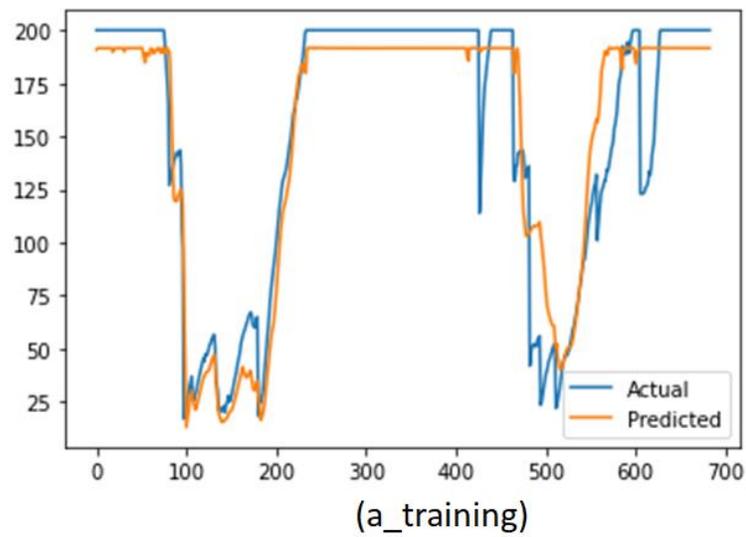
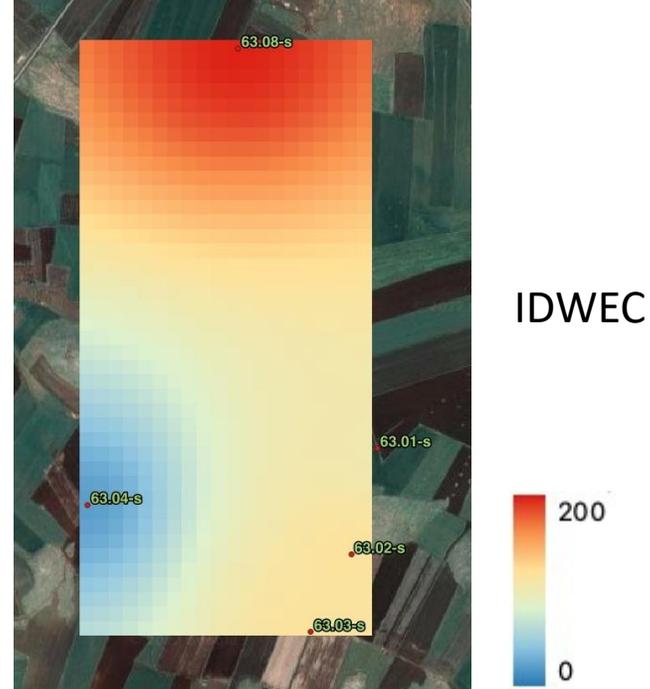
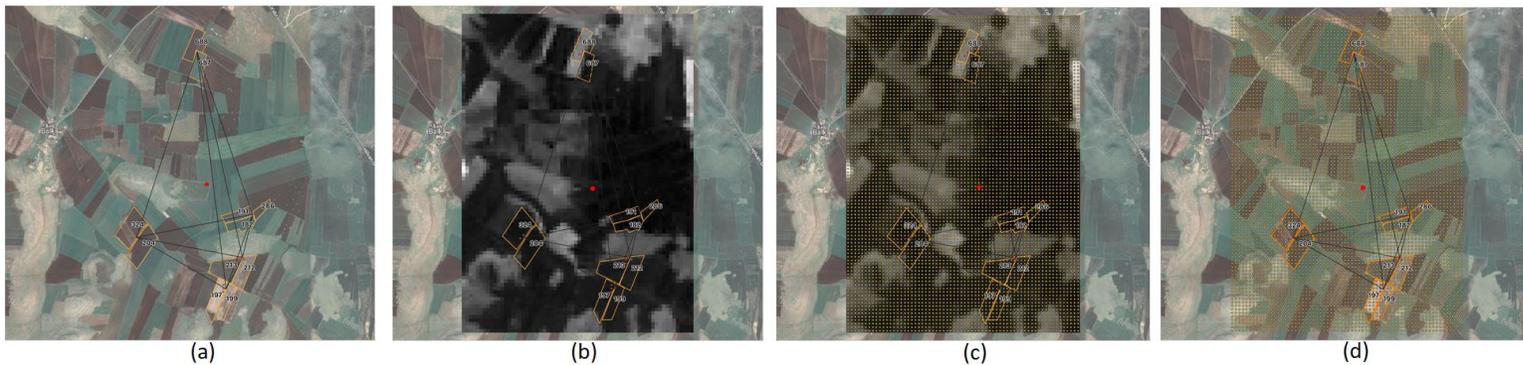
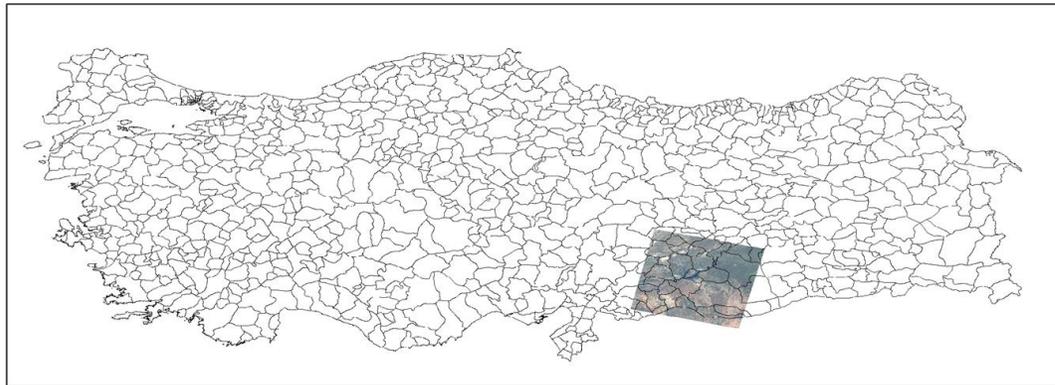


RMSE (cb): 0..200cb

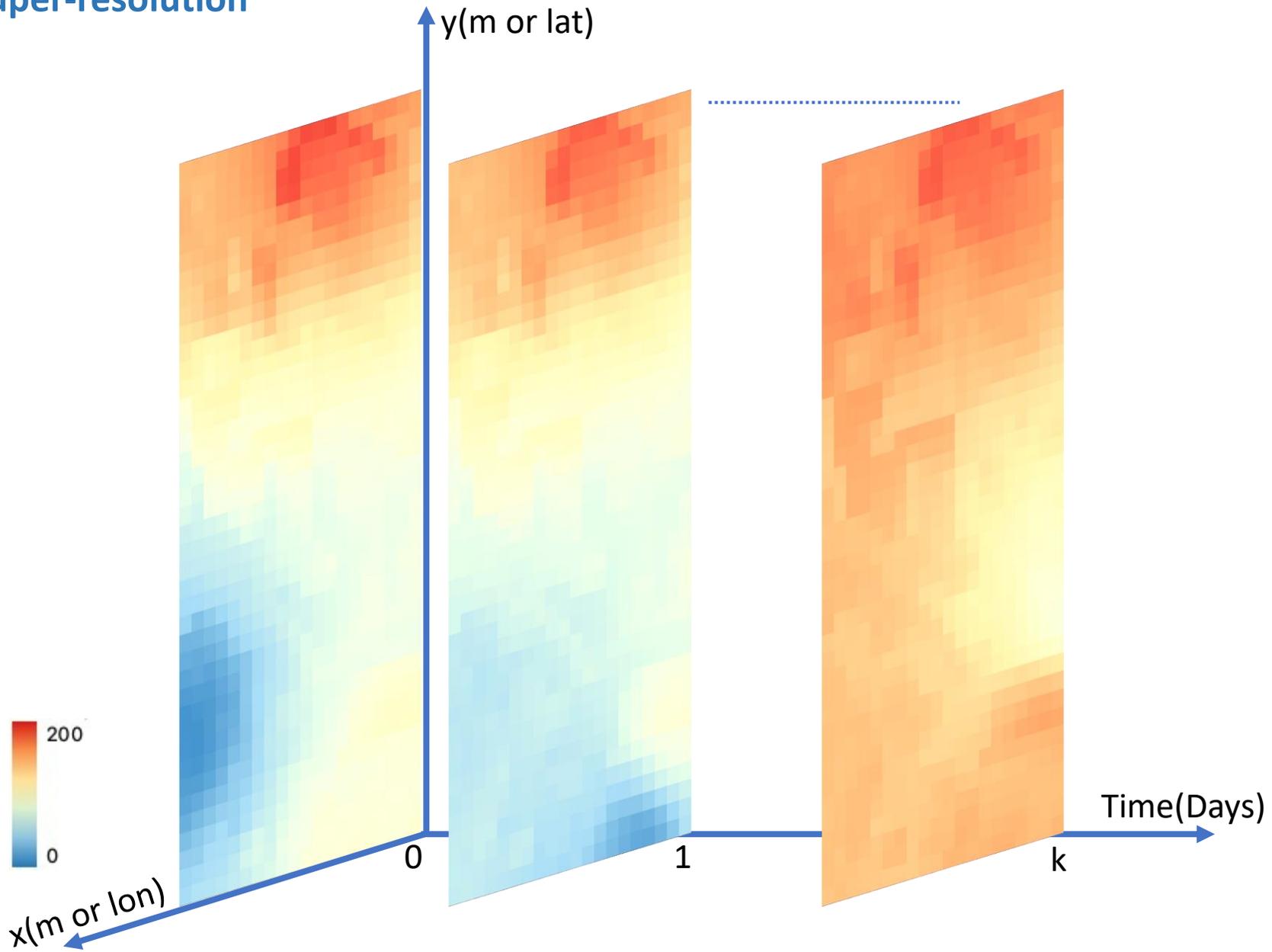
SM45	Regression	LSTM	CNN	ANN	PECNET
SM	39	22	23	26	22
SM, MSAVI	42	22	24	28	17
(RF-ETO), MSAVI	46	23	27	31	18

4- Continual Learning

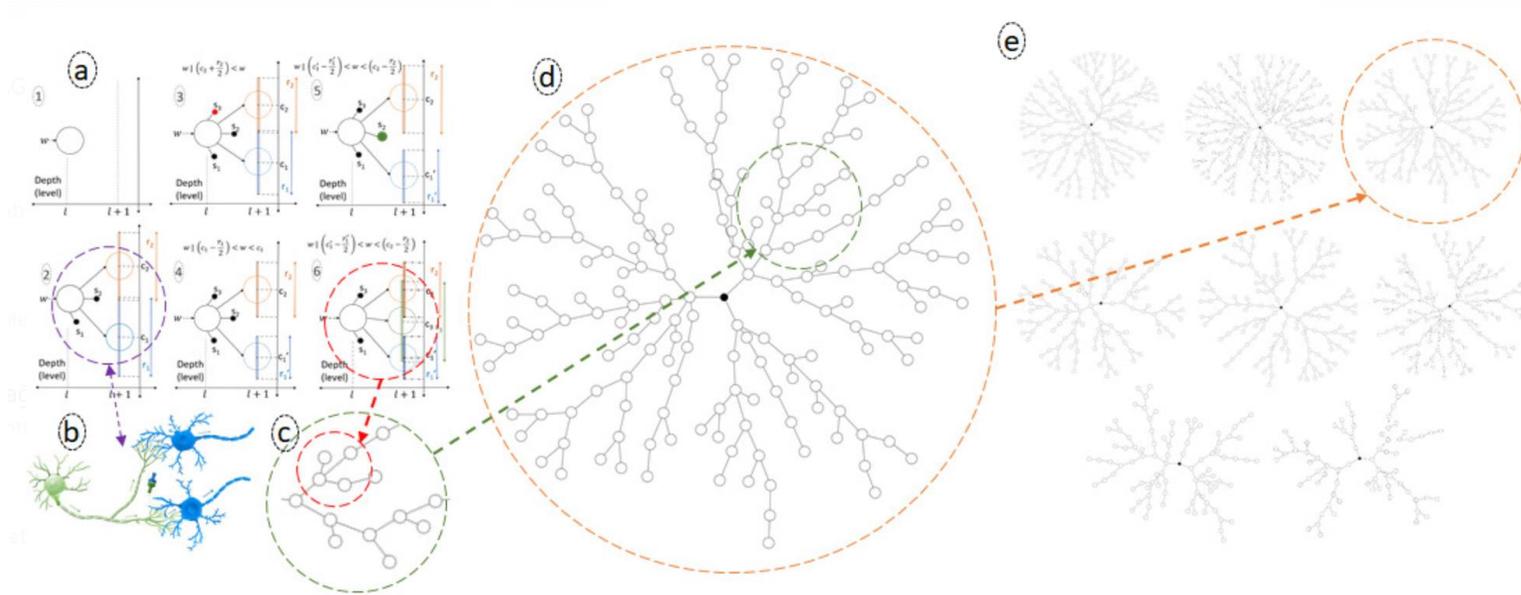
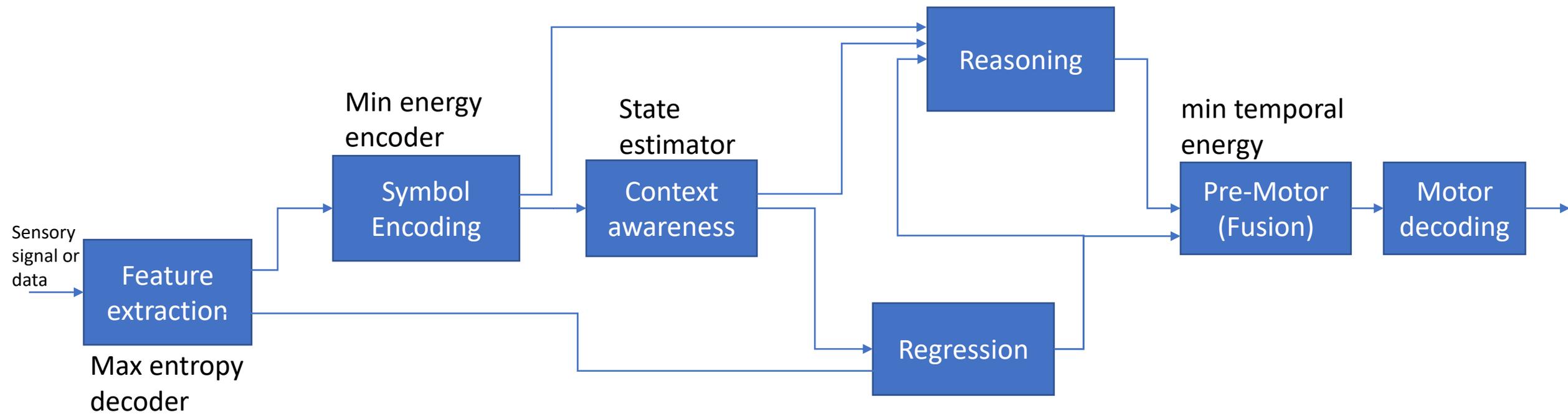


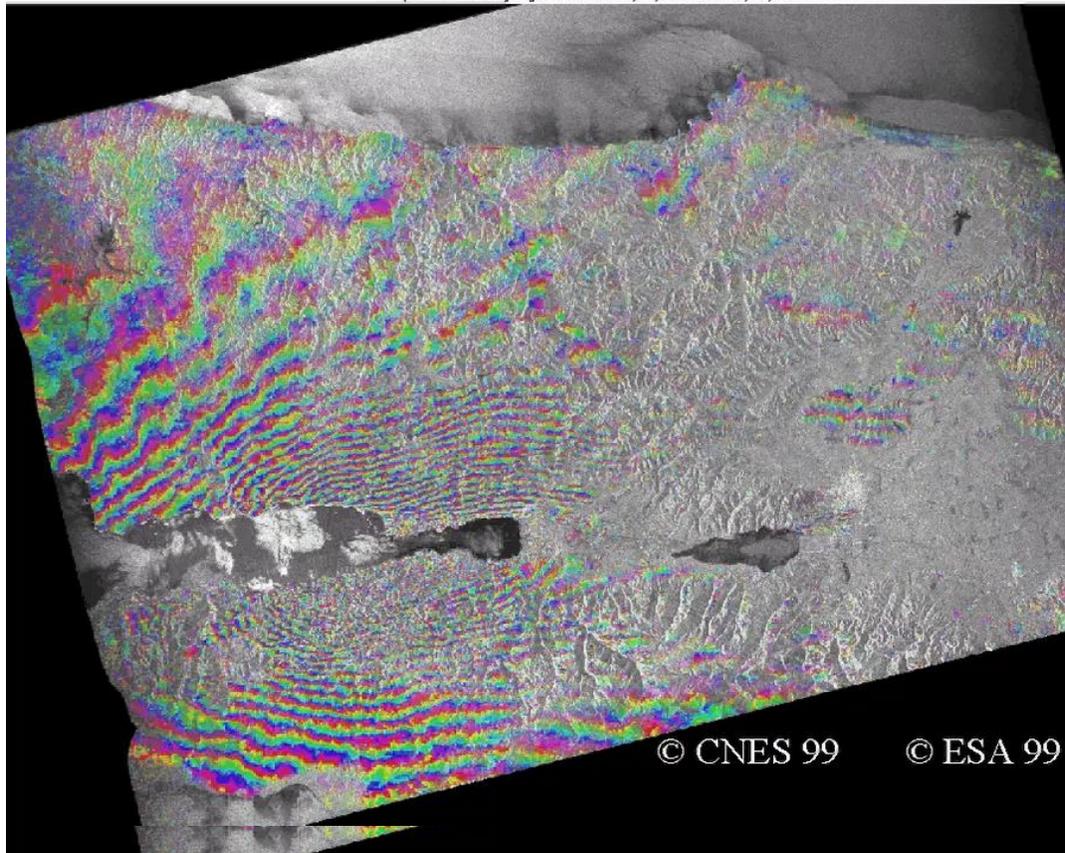


Spatio-temporal super-resolution fusion for casting



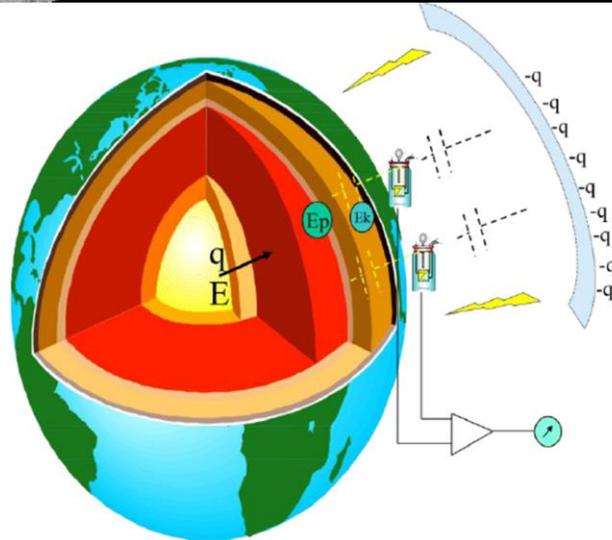
5- Analogy to Decision Making in Brain Connectome (Generalization of ML Roadmap):



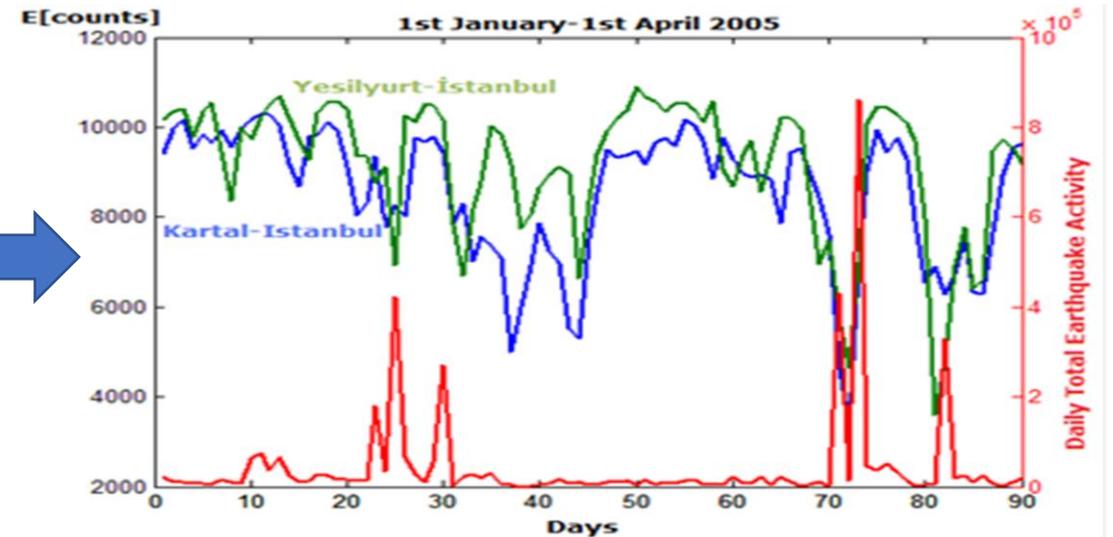


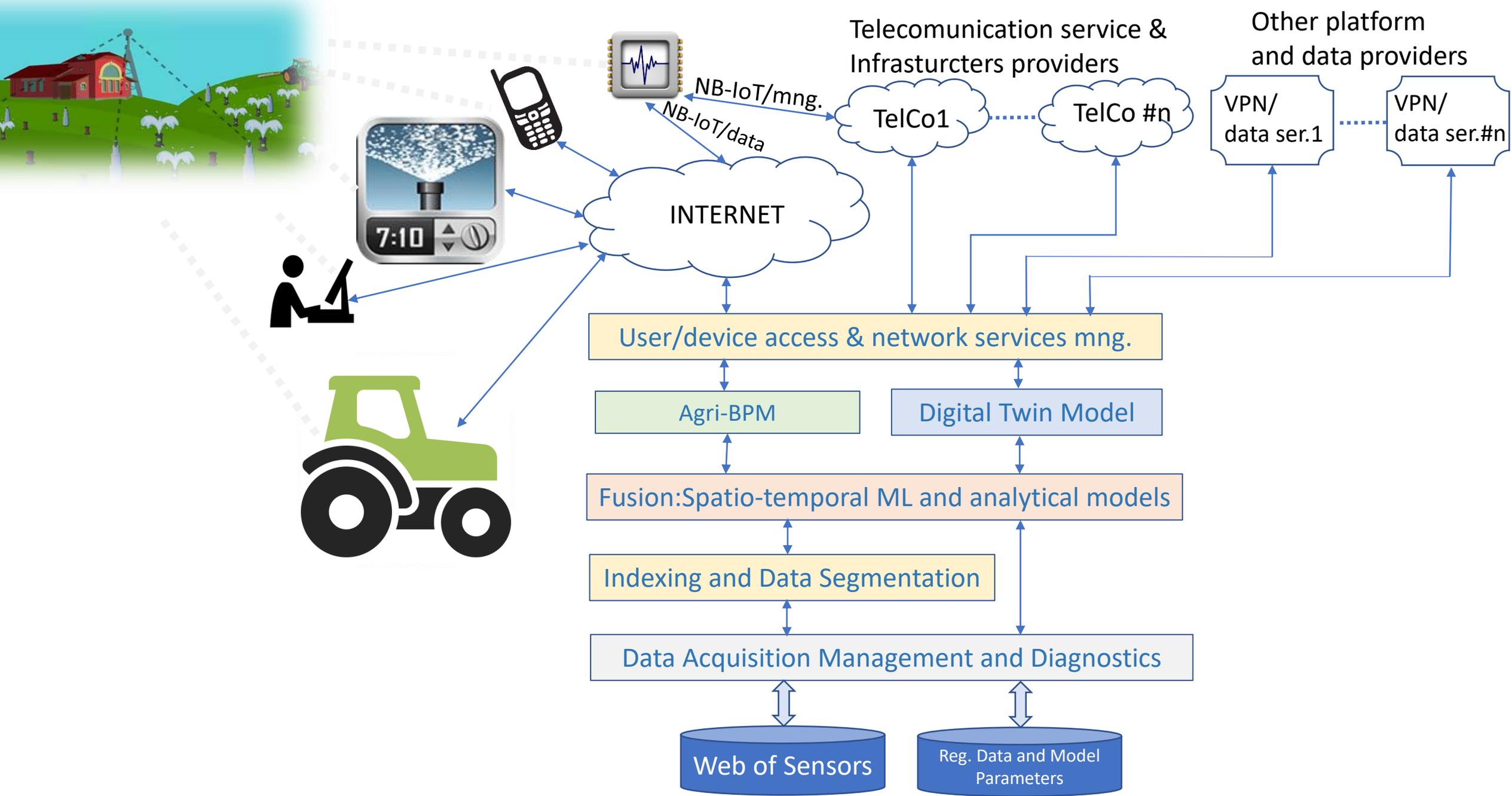
Fusion outcomes base on Artificial General Intelligence Roadmap:

- Risk management
- GDP estimation
- Energy production
- Analytics for mining
- Economic growthrate estimation
- Urban management
- Smart cities
- Resource management
-
- unlimited



AI Fusion





How to convert AI fusion based Digital Twinning into Services: PaaS , DaaS

Subsidy - Crop Pattern Management Panel

TOTAL SUBSIDY



TOTAL INCOME



Wheat

Cotton

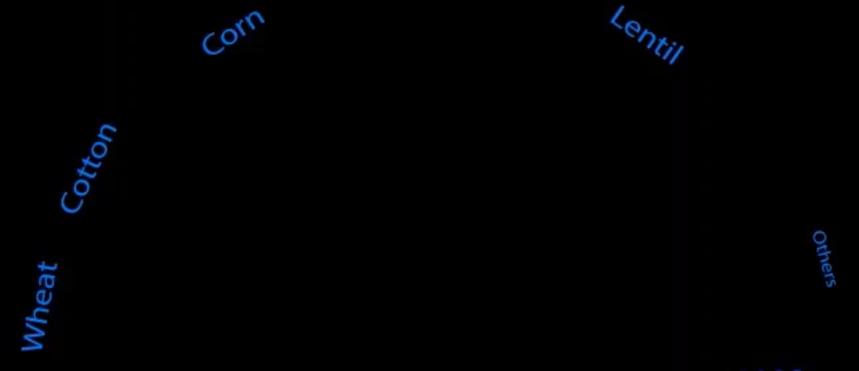
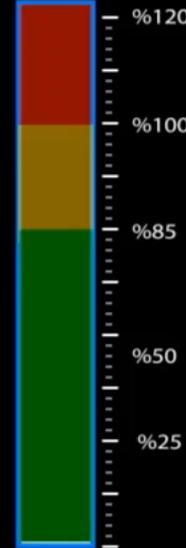
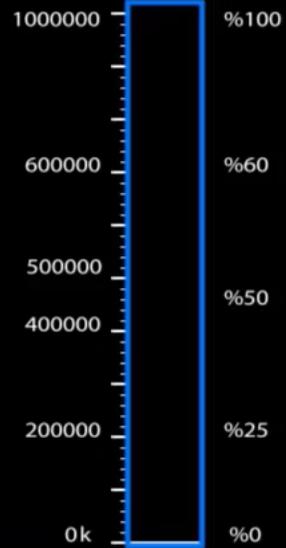
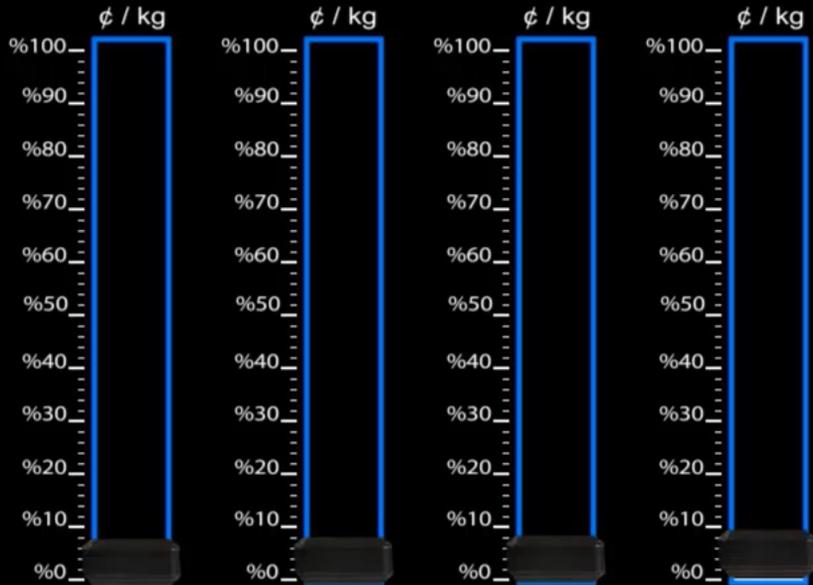
Corn

Lentil

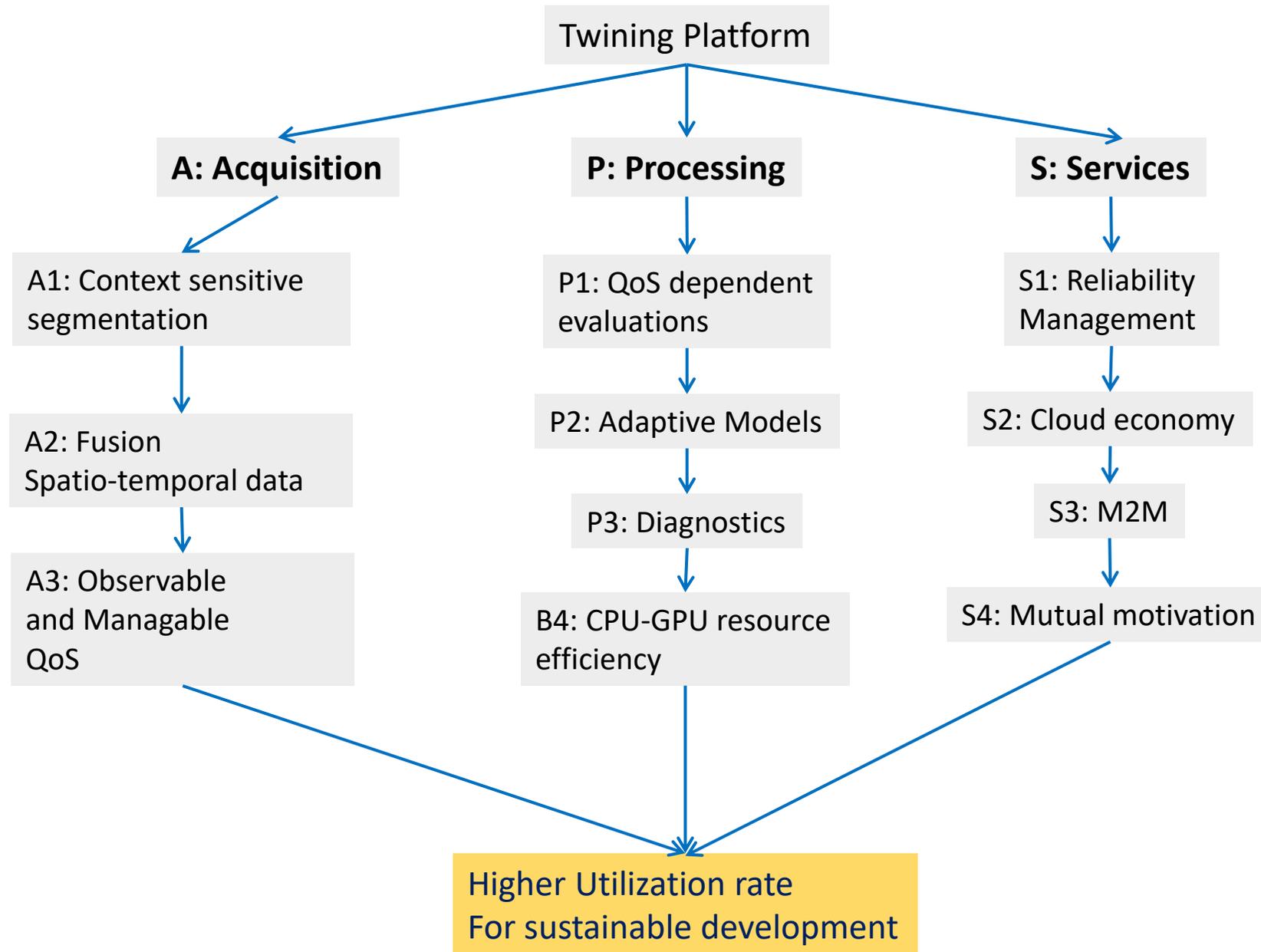
Utilized Basin Area (ha)

Annual Basin Water Resource Capacity

Estimated Cultivation Area Per Crop



Platform in Digital Twinning ?





Conclusions:

Overfitting management is the primary strategy in big RS data environment

“Accuracy as a state” requires near-real-time accuracy and Tolerance computation

Reference Network is necessary in platform scenarios besides individual sets

PECNET minimizes overfitting in Artificial Intelligence on RS data

Context Awareness in ML improves performance against non-linearity

PaaS

Conintual Learning must be supported against time variance of the system

Orthogonal feature extraction can be optimized with PECNET

Stochastic features can be decoded through cortical coding networks

Artificial General Intelligence and Neuromorphism must be rec. for Singularity

DaaS must take part in govermentlal subsidies at economic growth fields

Digital Twinning is the feasible convergence roadmap for nowcasting and forecasting



İTÜ



International Society of Agromatics (ISAM)

www.agromatics.org

TEŞEKKÜRLER

谢谢

THANK YOU

شكر

Prof.Dr. Berk USTUNDAG

bustundag@itu.edu.tr