

RESEARCH TOPICS ON POWER SYSTEMS

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<https://www.pbs.org/video/america-revealed-birth-power-grid/>



<https://www.power-grid.com/executive-insight/power-gen-europe-confidence-index-european-power-report/#gref>

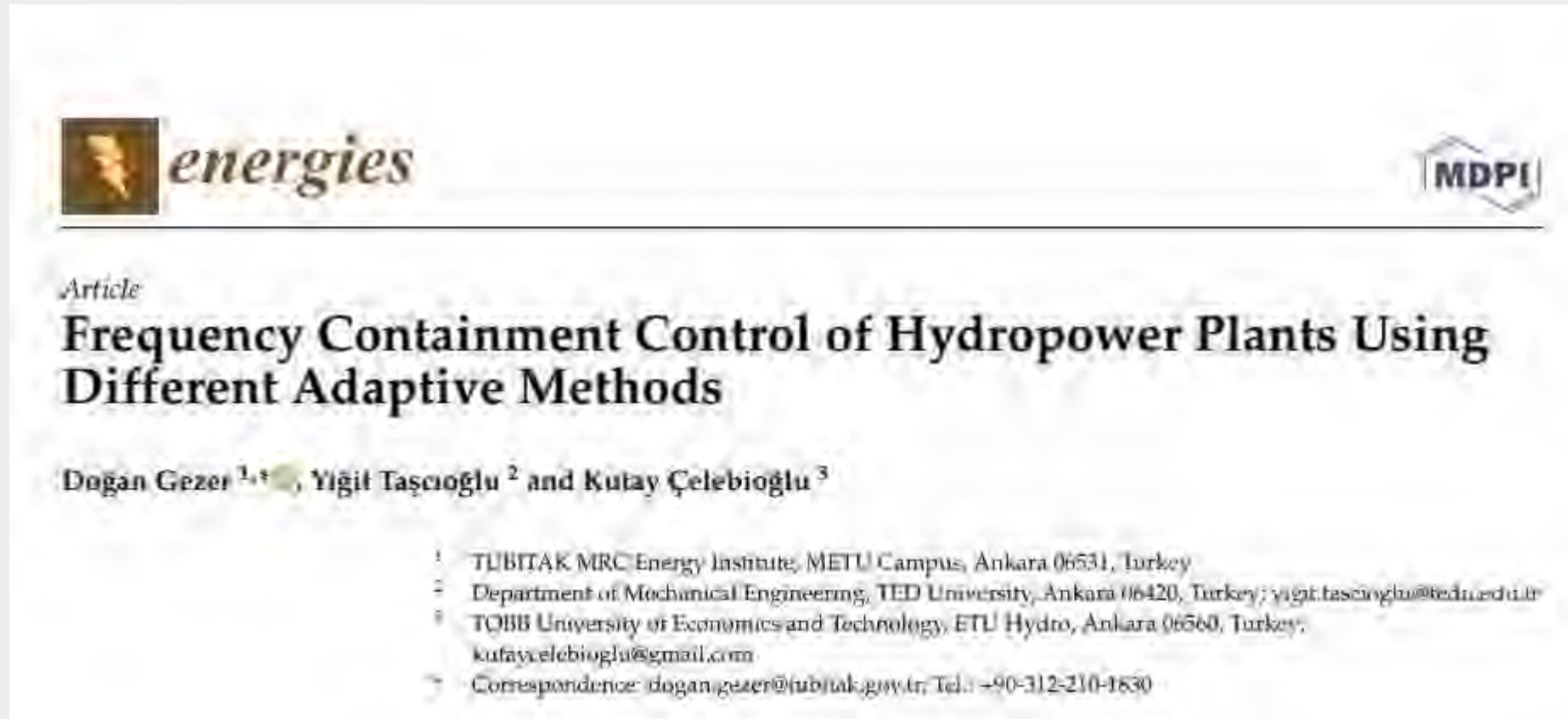
TOPICS THAT WILL BE COVERED

- ✓ Advancements in Generation System
- ✓ Advancements in Transmission system
- ✓ Advancements in Distribution system
- ✓ Machine learning for power systems (Time permits!)

ADVANCEMENTS IN GENERATION SYSTEM

➤ Modeling and Validation of Hydropower Plants

- Physics driven (Based on differential equation)



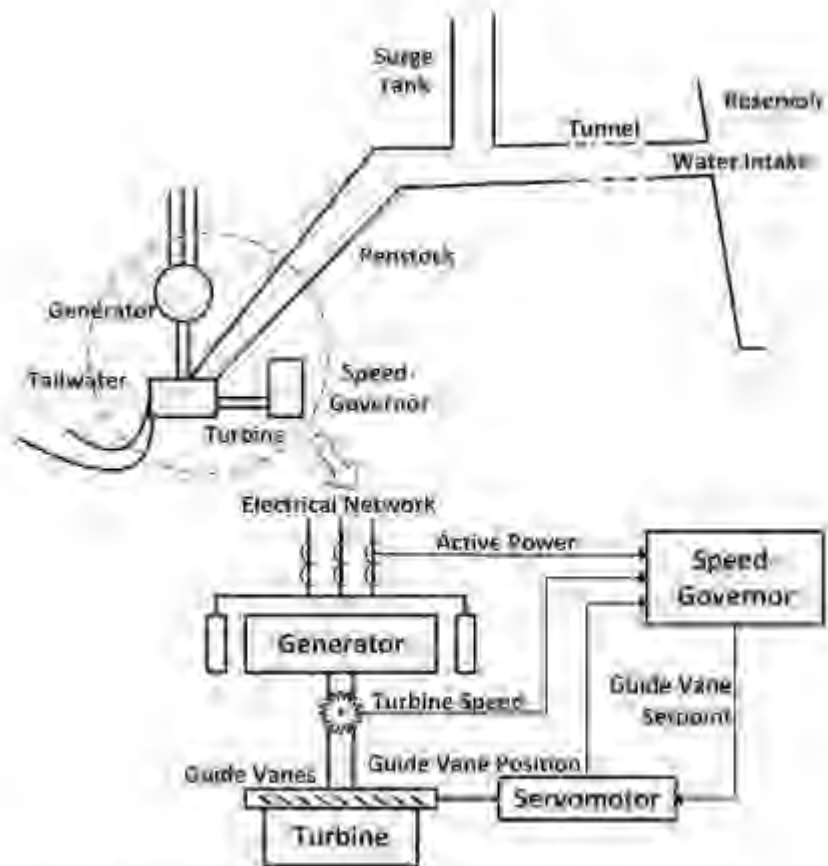


Figure 1. General layout of a hydropower plant (HPP)

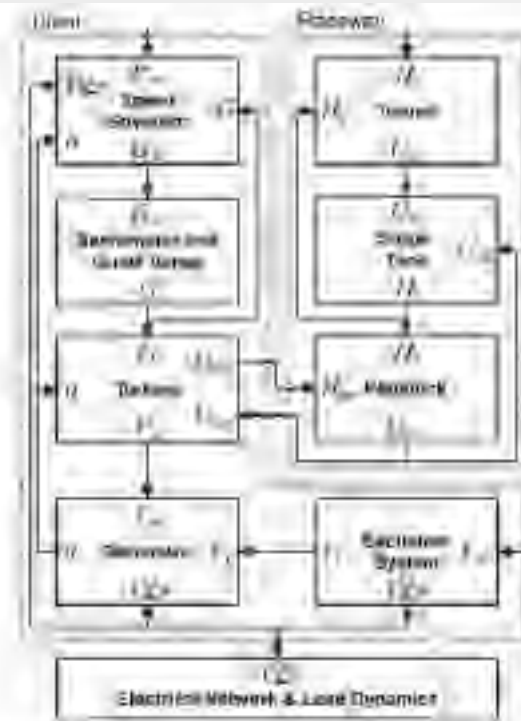


Figure 2. Transient and steady-state models of HPP

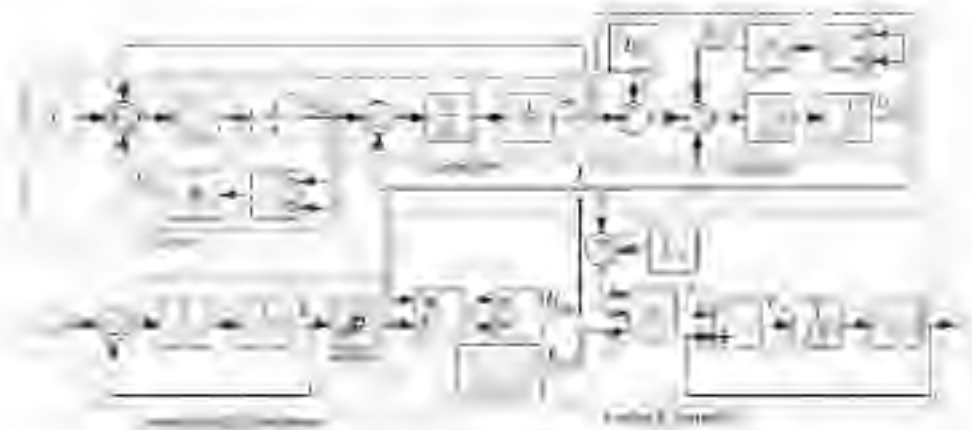


Figure 3. Dynamic model of a hydropower plant

2.3 System Description

In this study, Seyhan I HPP was modeled with the characteristic parameters listed in Table 1. Seyhan I HPP is an aged hydropower plant that started operation in 1997. It has three vertical Francis turbines with 22.5 MVA of installed capacity each.

Table 1. Characteristic parameters of Seyhan I HPP¹

Parameter	Value	Unit
Nominal net head	72	m
Nominal flow rate	77	m ³ /s
Surge tank storage capacity	124.8	s
Surge tank cross-sectional area	380	m ²
Penstock length	417	m
Penstock cross-sectional area	21.2	m ²
Penstock friction factor	0.05	
Gate valve opening at no-load	11	%
Gate valve opening at full load	65	%
Surge tank time constant	1	s
Gate valve delay time	2	s
Flywheel effect of the generator	3320	tonm ²
Turbine nominal speed	125	rpm
Apparent power of the generator	22.5	MVA
Inertia constant of the generator	3.14	s

- Software: Open-Modelica, Simulink..

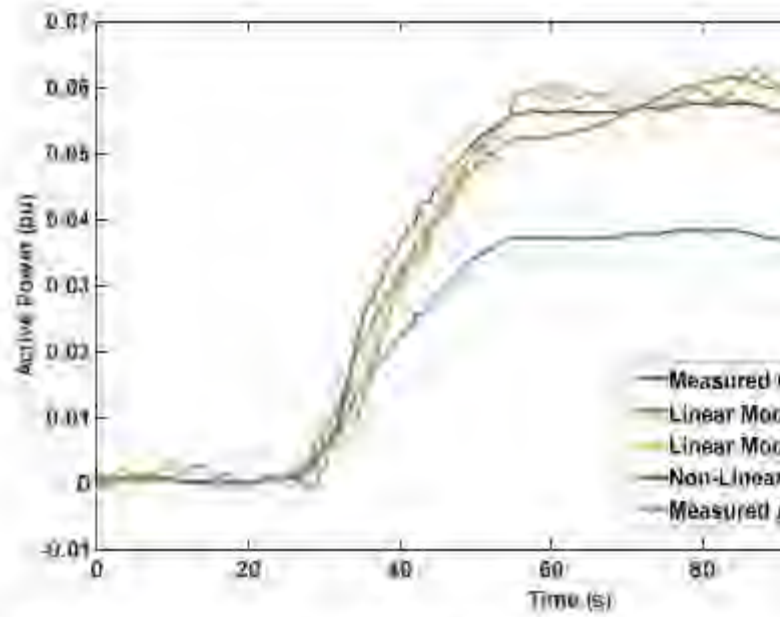


Figure 3. Comparison of various models with site measurements.

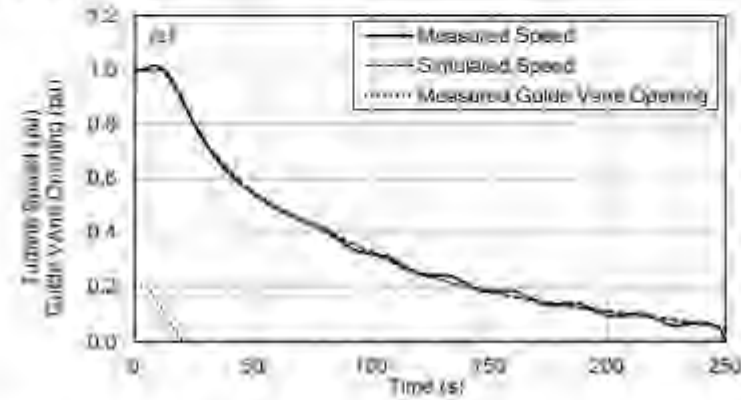
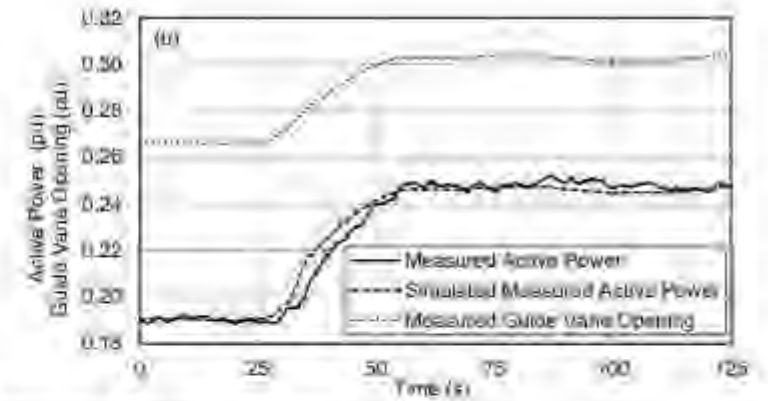
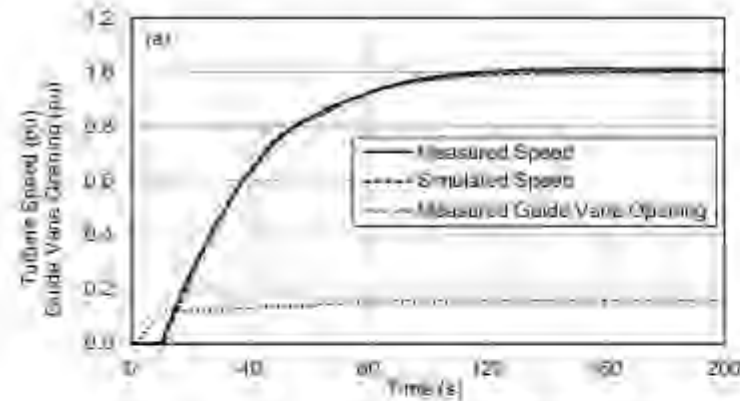


Figure 4. Model validation of Seyhan I HPP with site measurements during (a) start-up, (b) loading, and (c) shutdown.

- Can be further extended to data-driven models

➤ Condition Monitoring for Hydropower plants

- Sensors, Signal Processing, Internet of Things (IOT), AI..



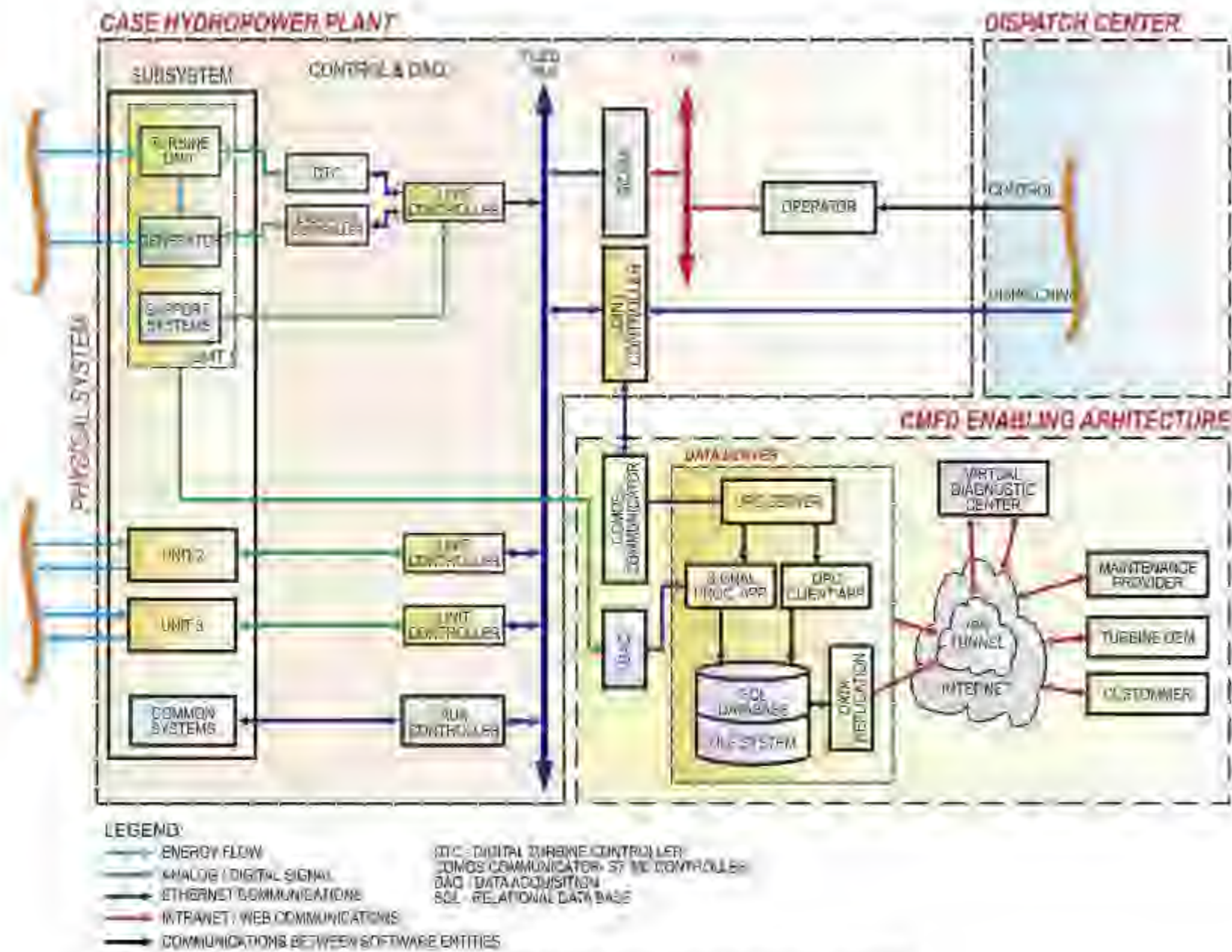


Fig. 7. Implementation of CMFD architecture for the case study of the HPP

CMFD: Condition Monitoring and Fault Diagnostics

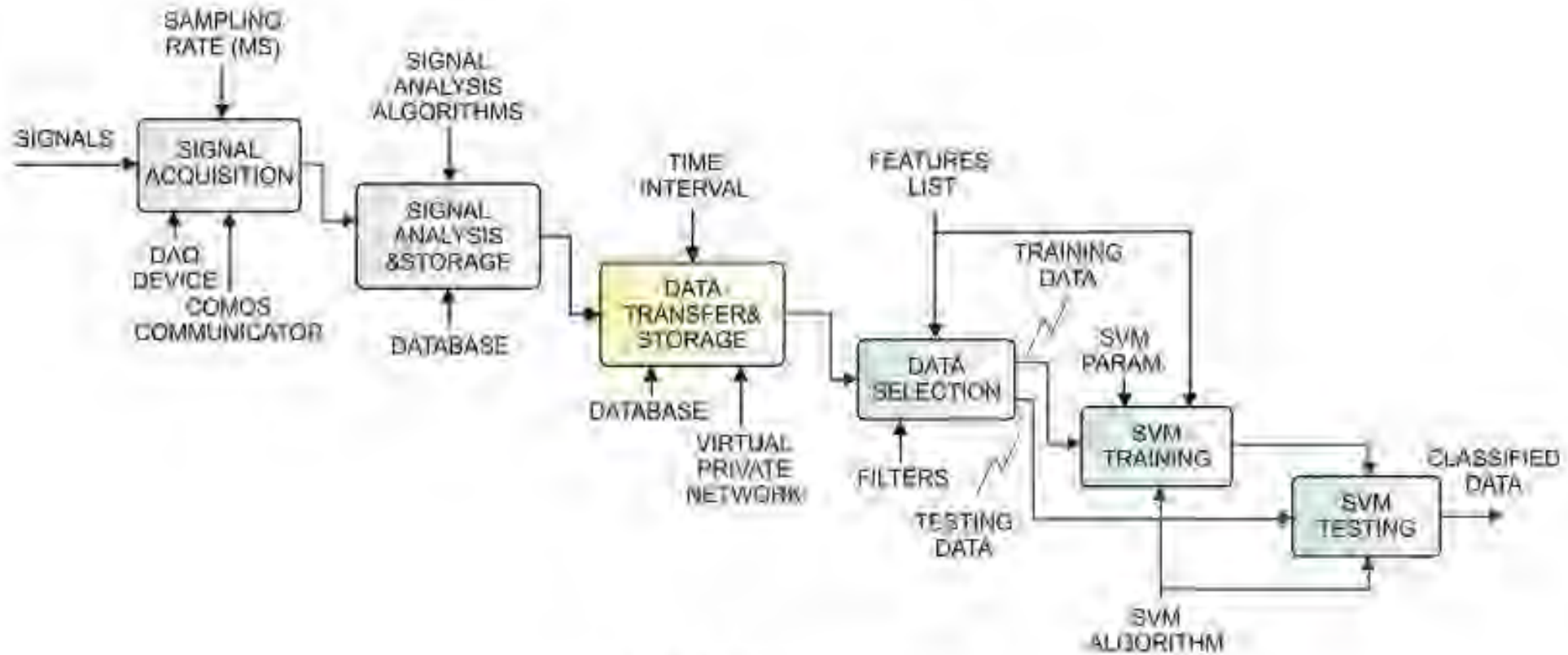


Fig. 4. The CMFD functional model.

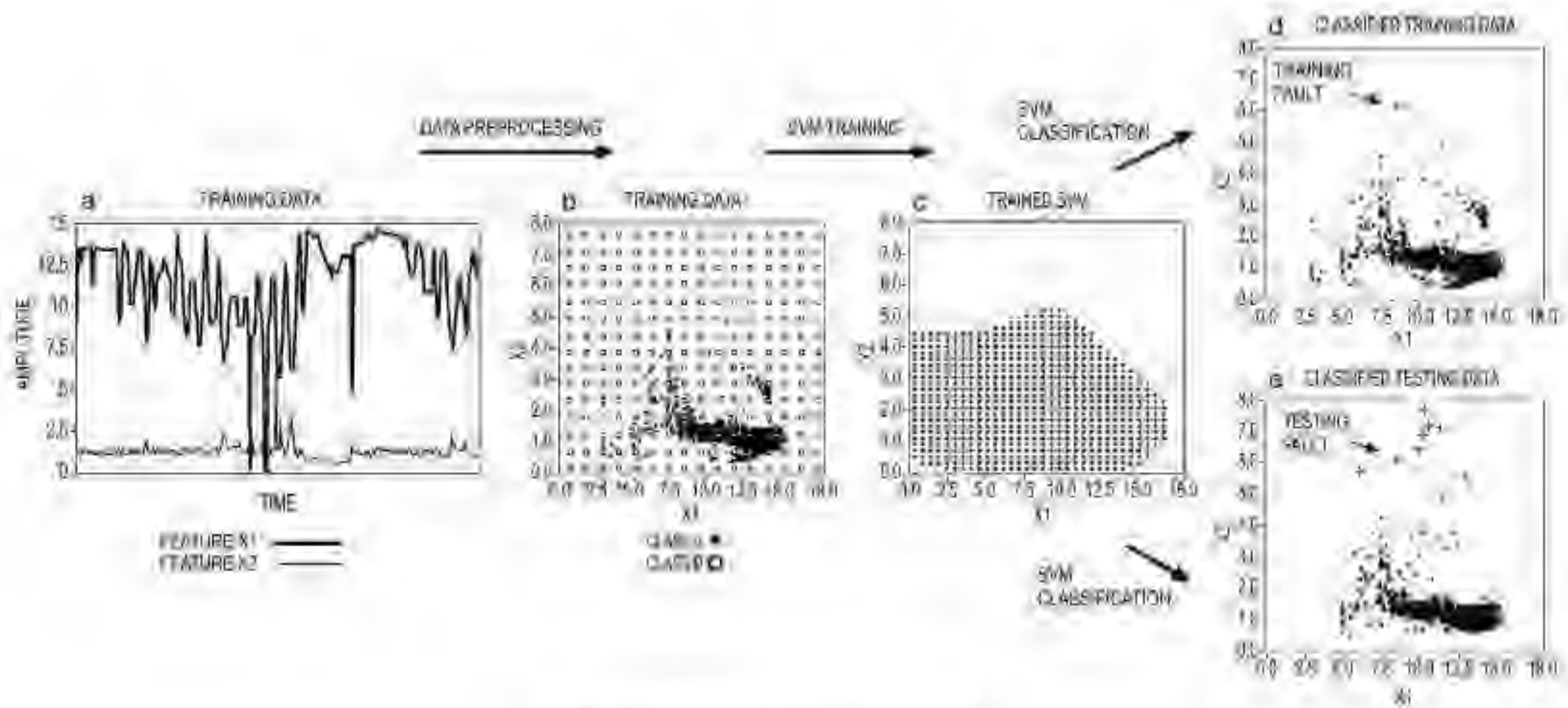
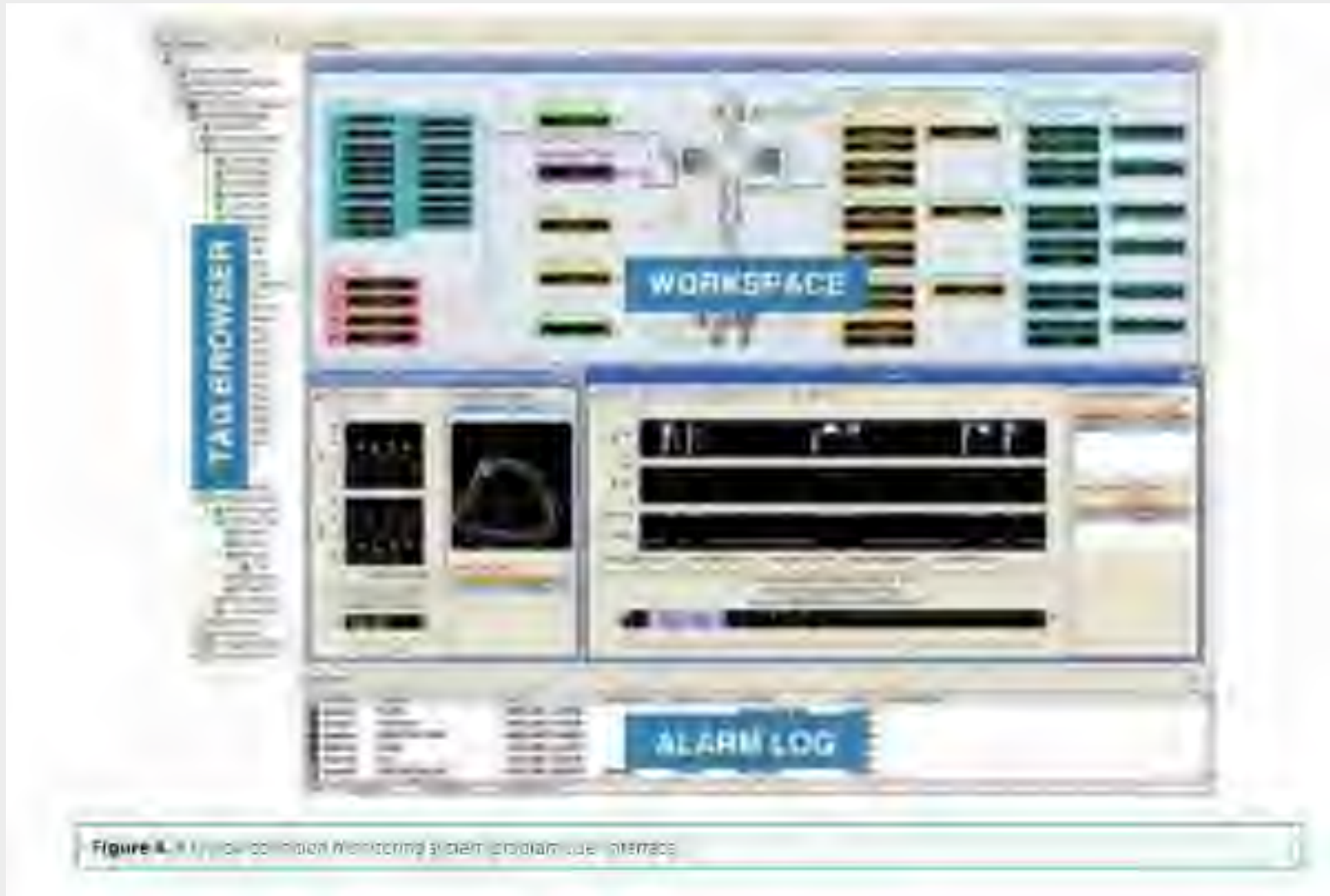


Fig. 6. Fault classification of one causality.

Table 1

Causalities of faults and indicators for thrust bearing.

Causality ID	Fault description	Feature					
		Output power	Rotation frequency	Bearing babbit temperature	Oil level	Oil temperature	R.M.S. velocity
9	Overheating of bearing	X1		X2			
10	Bearing lubrication oil consumption		X1		X2		
11	insufficient heat transfer from the bearing cooling system	X1				X2	
12	Bearing performance degradation	X1					X2
30	Excessive bearing babbit temperature			X1			
31	Bearing lubrication oil consumption				X1		
32	Overheating of bearing					X1	
33	Bearing performance degradation						X1



Alagöz, İzzet, Mehmet Bulut, Veysel Geylani, and Arif Yıldırım. "Importance of real-time hydro power plant condition monitoring systems and contribution to electricity production." *Turk. J. Electr. Power Energy Syst* 1, no. 1 (2021): 1-11.

SOLAR PHOTOVOLTAIC GENERATION IN NEPAL

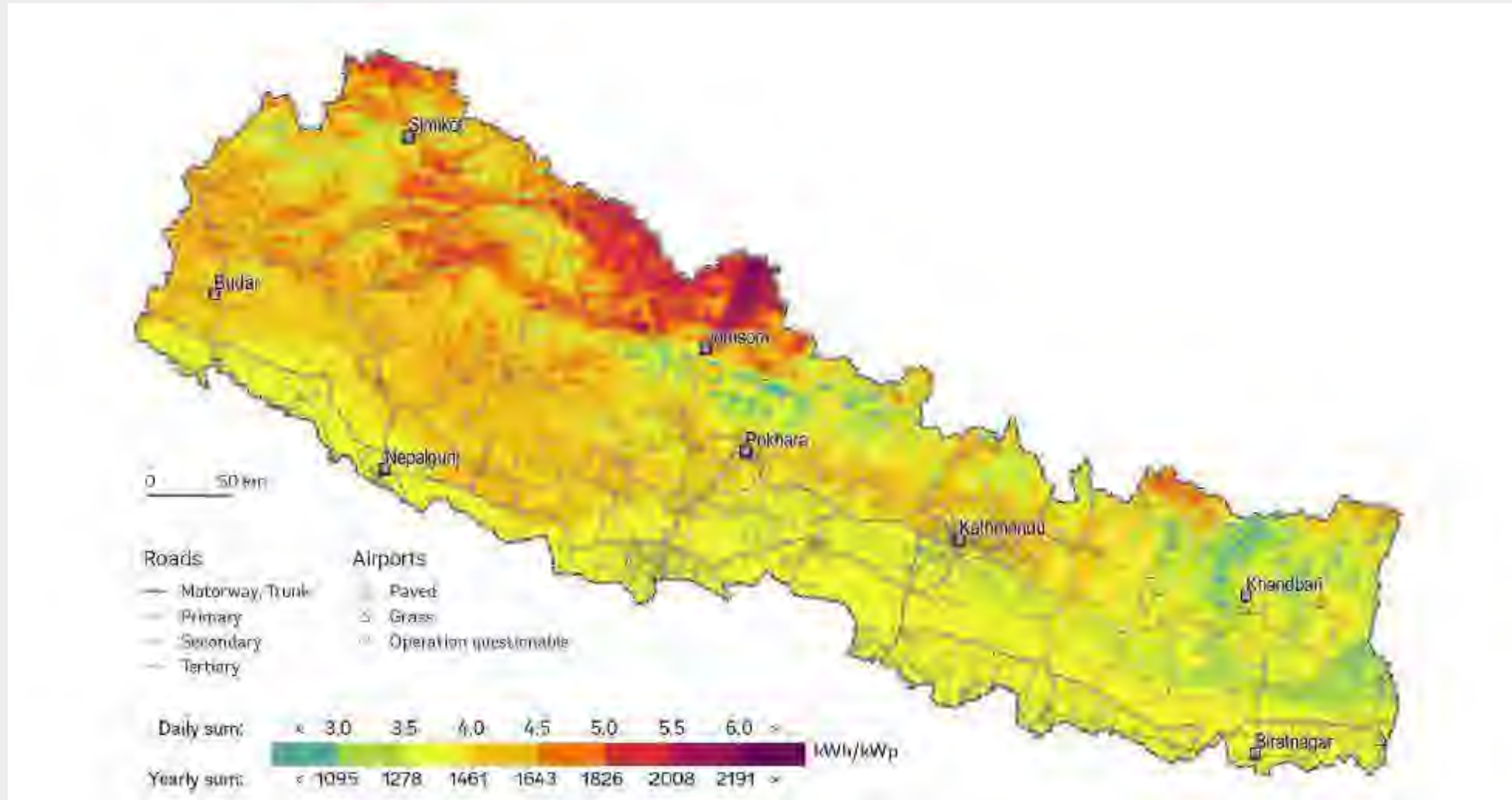


Fig. Solar PV potential in Nepal

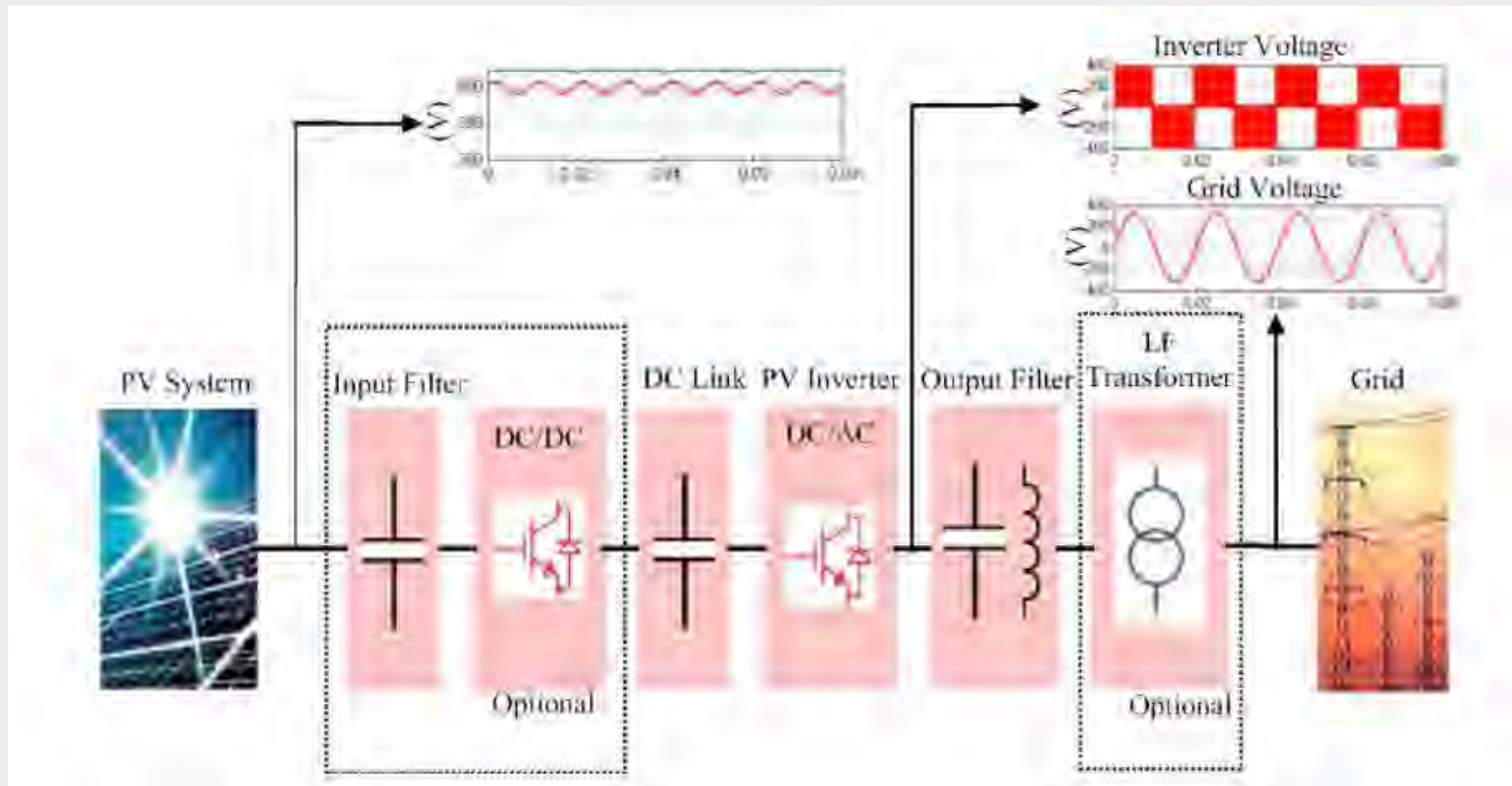


Fig. PV schematic diagram

➤ Controller Interaction

- Parallel operation of **30 inverters** a 500 kW solar plant connected to a 44kV feeder via 600V/44kV

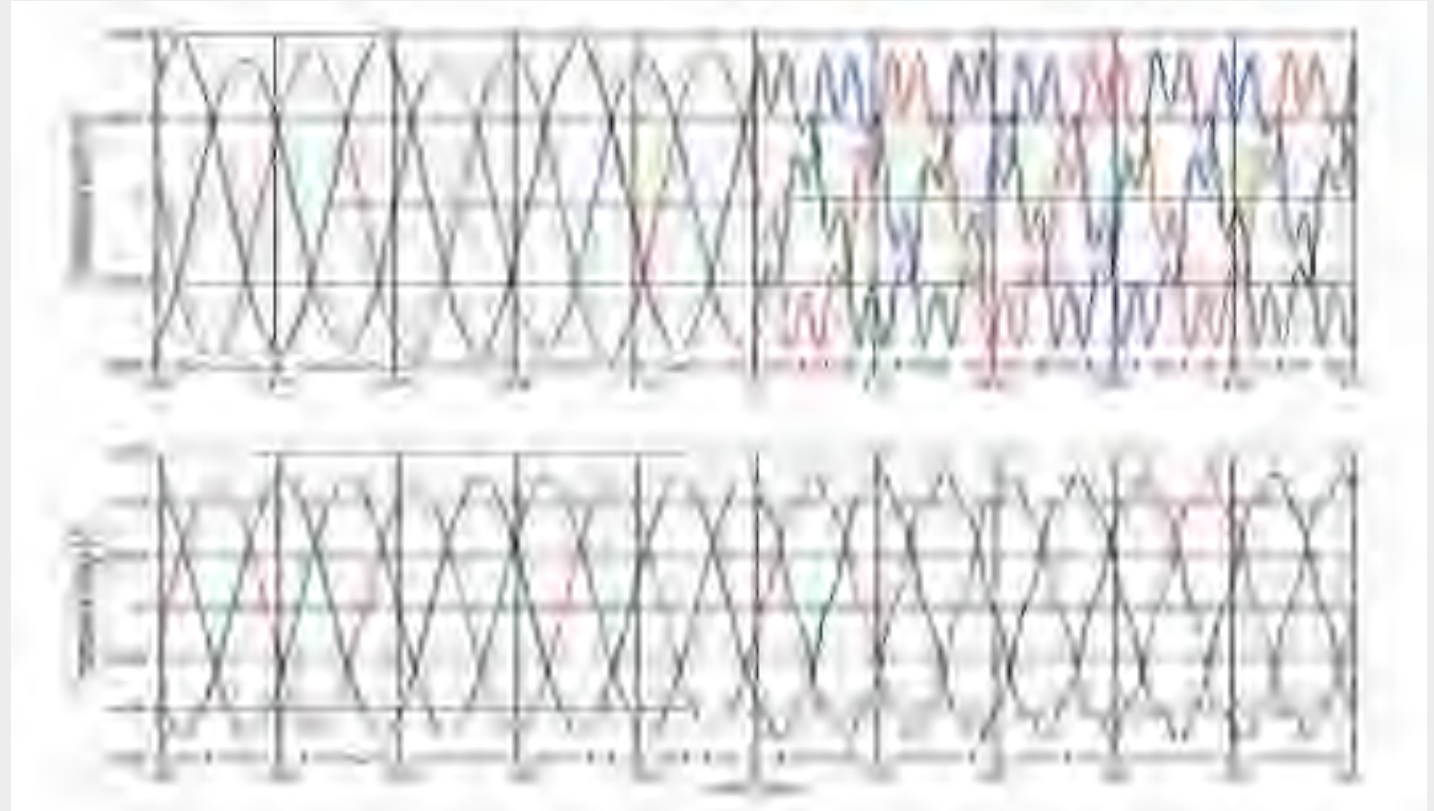
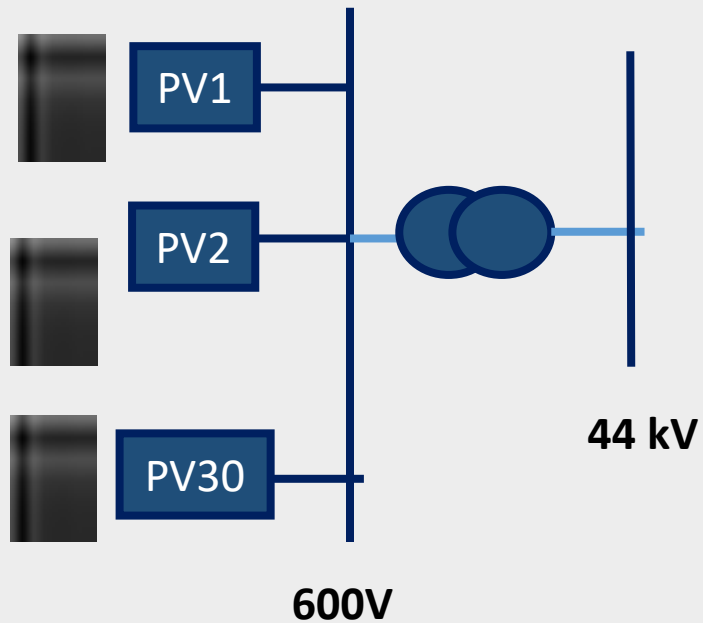
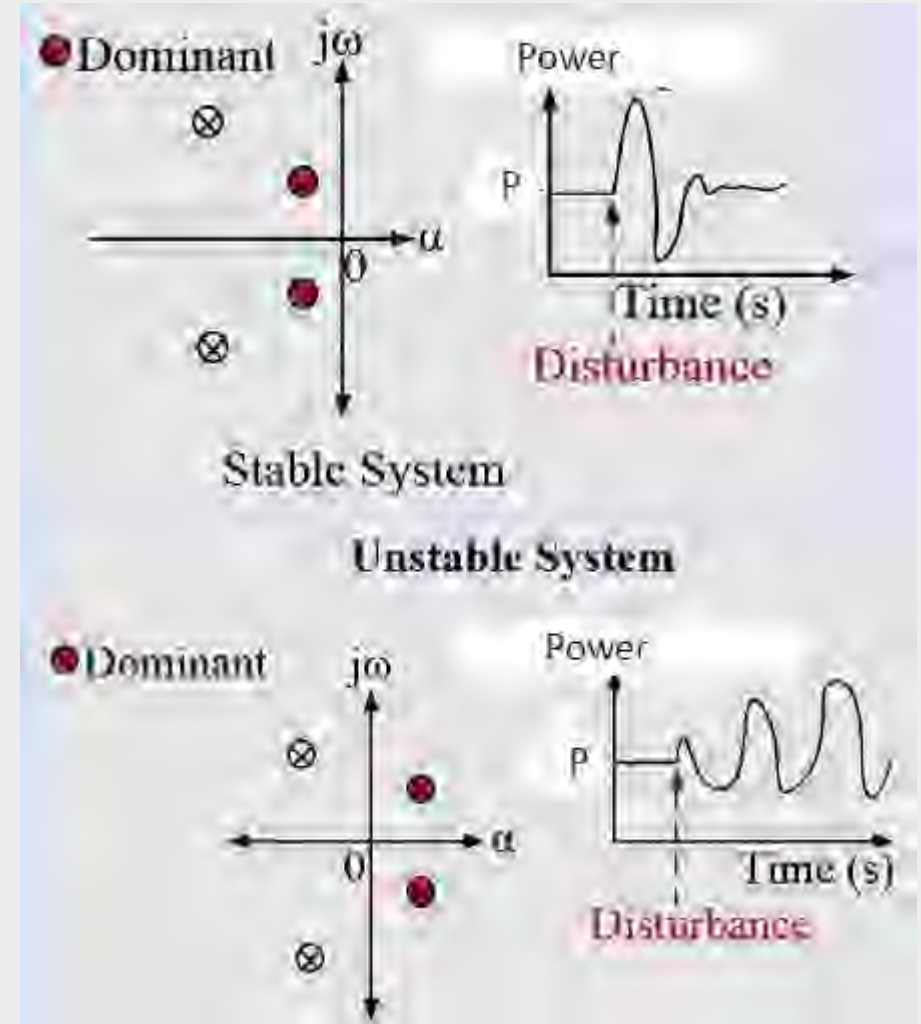
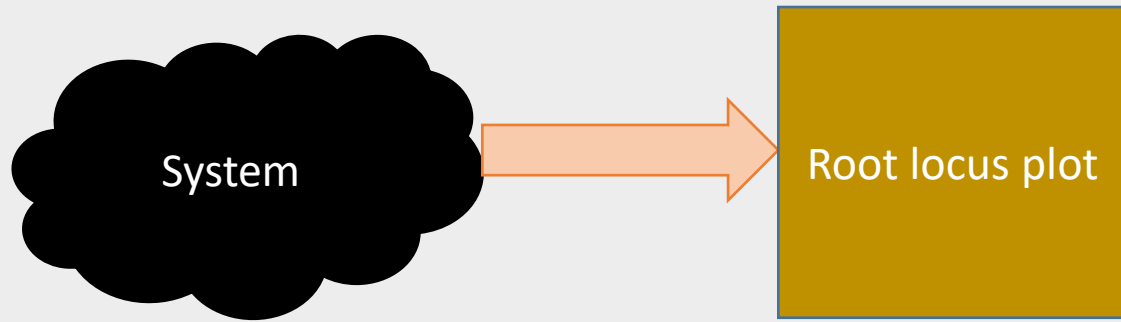
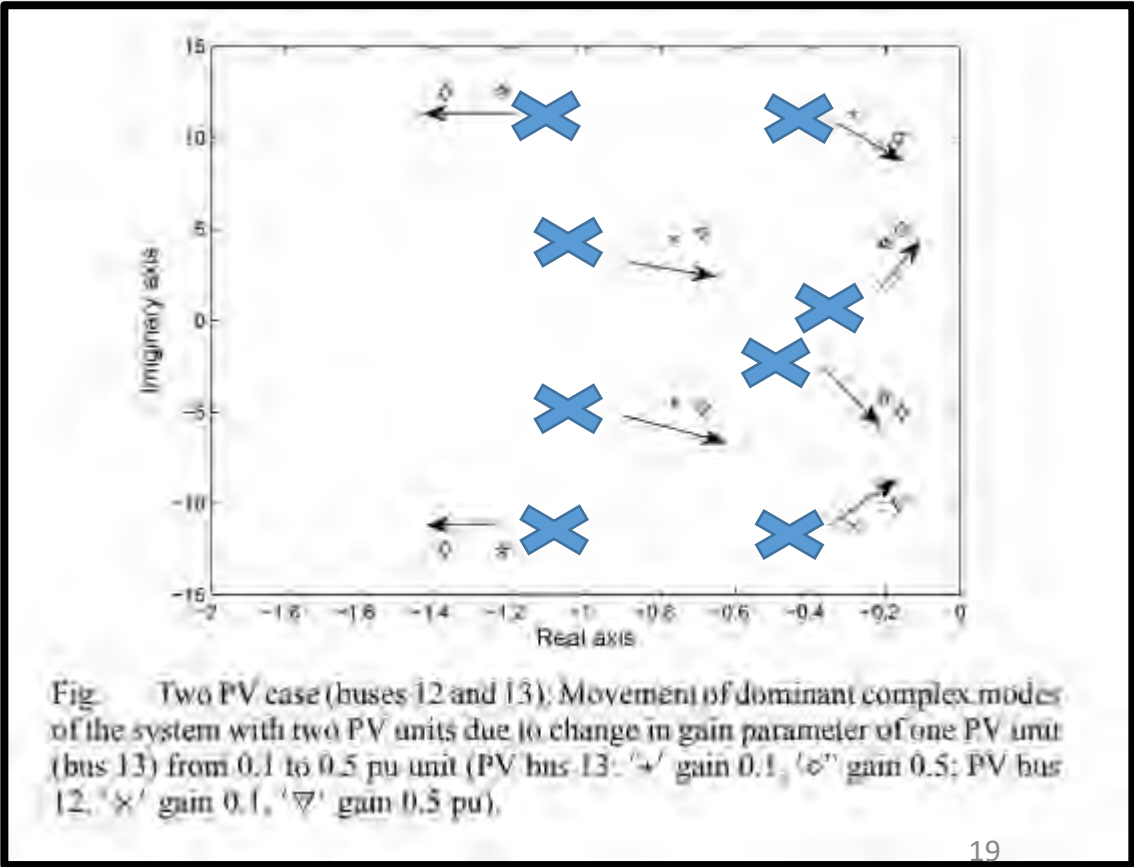
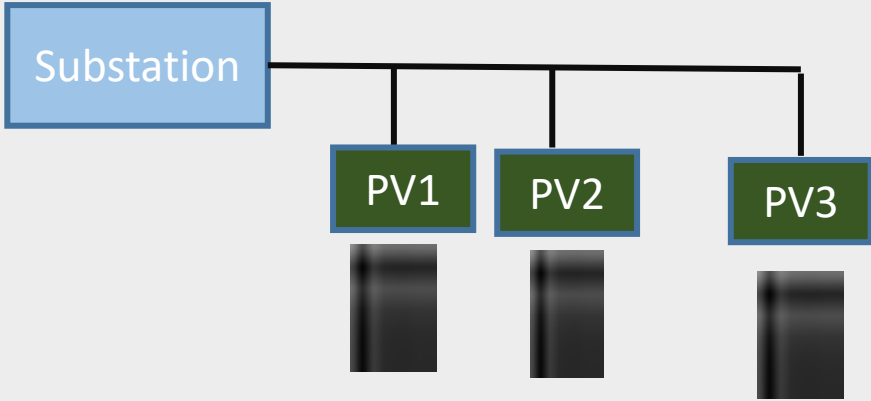


Fig. Current and voltage waveforms under normal and unstable operations



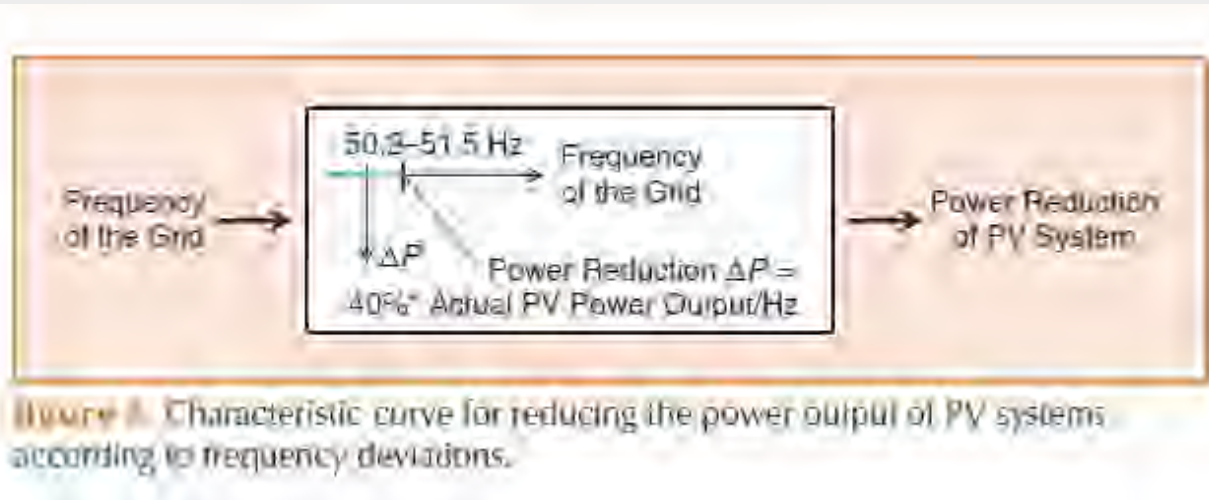


OTHER REQUIREMENTS OF SOLAR INVERTER

- Active power curtailment
 - Reactive power support
 - Inertia support
 - Low Harmonic content....
- } **Ancillary services**

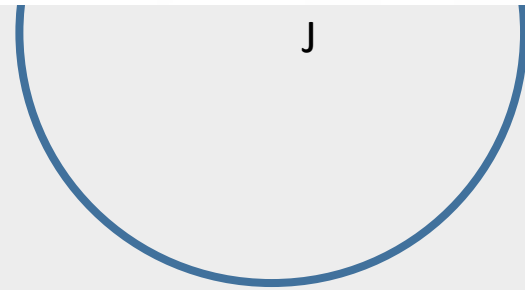
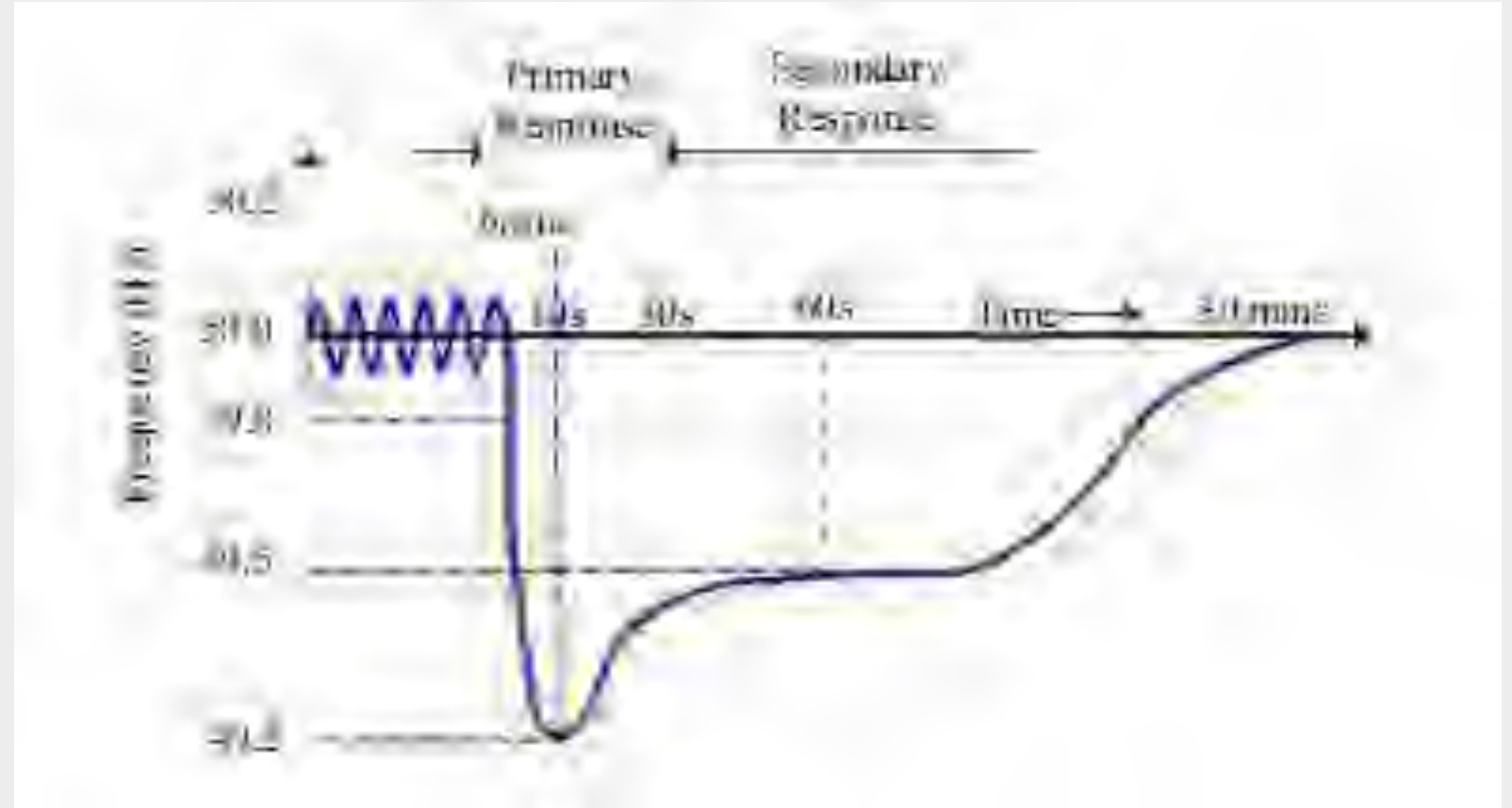
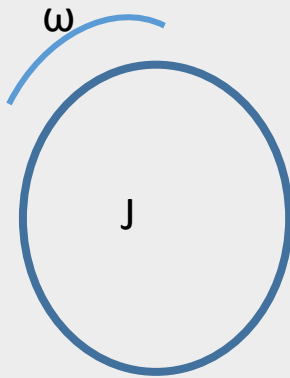
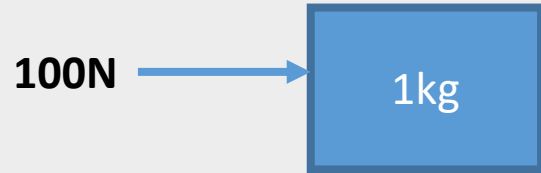
➤ Solar converter must be able to be able to **fulfill all the requirements** specified by the country grid code

Retrofit controllers for Active Power Curtailment



Retrofit controllers

Problem of Inertia



- SG provides inertia wh

$$\frac{df}{dt} = \frac{f_0}{2H_{sys} S_B} (P_m - P_e)$$

VIRTUAL SYNCHRONOUS MACHINE (VSM)

- Allows renewable converters to behave like conventional generators -> **(Synthetic inertia)**

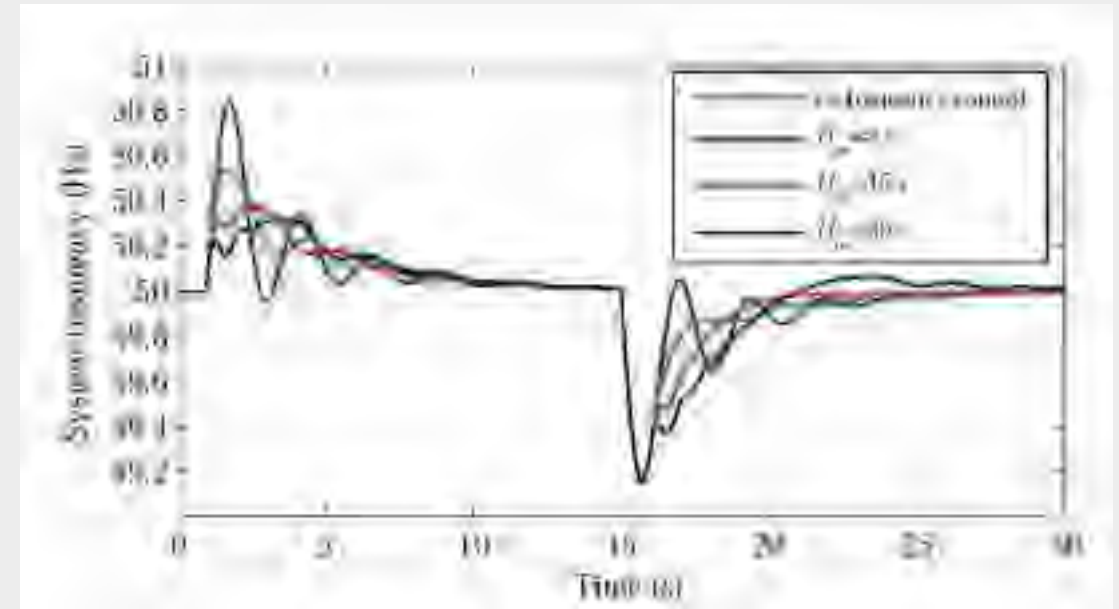
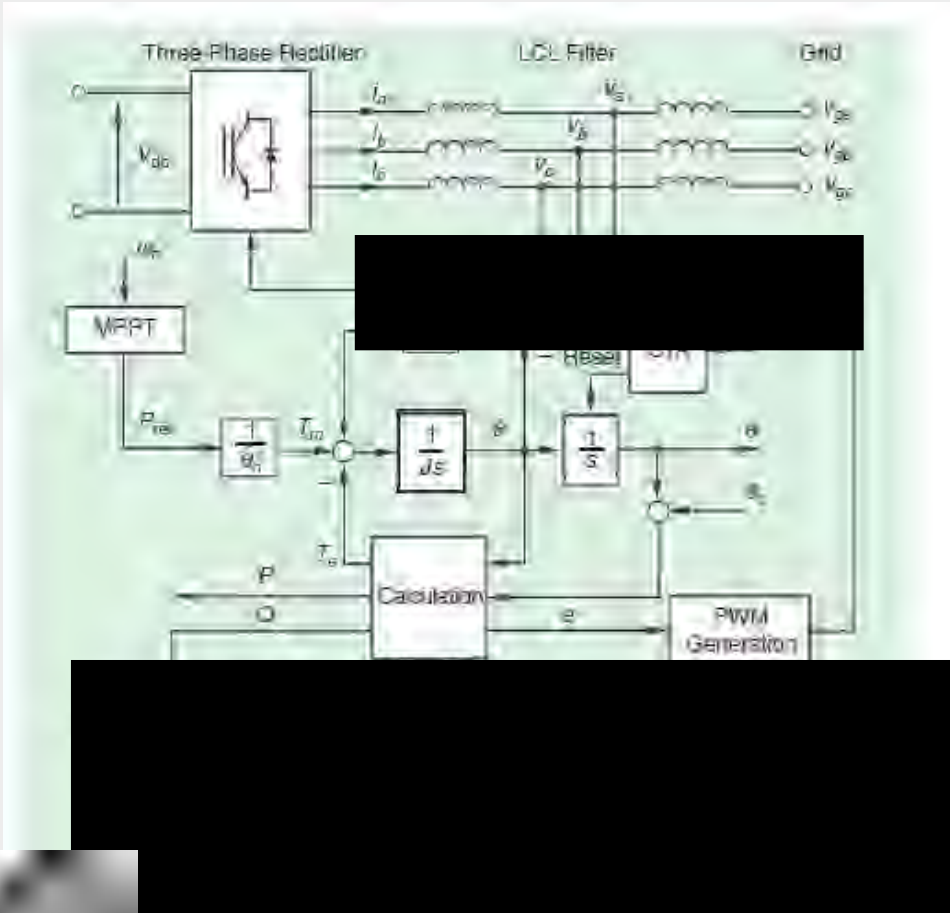


Fig. Virtual synchronous machine

Fig. System Frequency



**Problem Assessment Project
Mid Term Progress**

FREQUENCY STABILITY ASSESSMENT OF POWER SYSTEM NETWORK OF BAGMATI AND GANDAKI PROVINCE



Presenter:

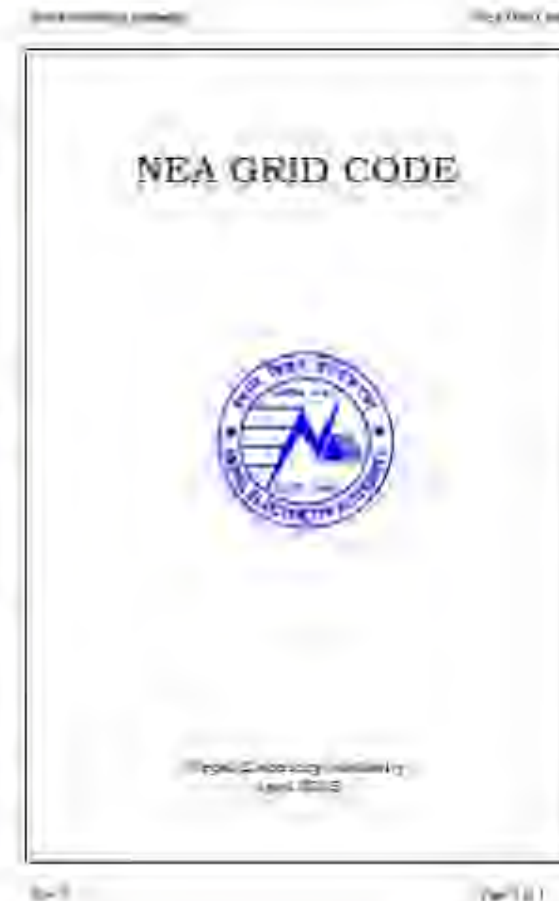
Amrit Parajuli
Graduate Student
M.E., Electrical Power Engineering

Supervisors:

Asst. Prof. Dr. Samundra Guring
Lecturer Anil Lamichhane

Frequency Regulation standards in Nepal

- ❑ Regulations set by Nepal Electricity Authority (NEA).
- ❑ The fundamental frequency of the system should be maintained between 48.75 Hz and 51.25 Hz i.e., $\pm 2.5\%$ of 50 Hz.
- ❑ The grid operating states is classified into 3 operating states from the perspective of frequency regulation as:
 1. Normal State: The System Frequency is within the limits of 49.5 Hz and 50.5 Hz.
 2. Alert State: The System Frequency is outside the limits of 49.5 and 50.5 Hz but within the limits of 48.75 Hz and 51.25 Hz.
 3. Emergency State: The System Frequency is outside the limits of 48.75 Hz and 51.25 Hz.



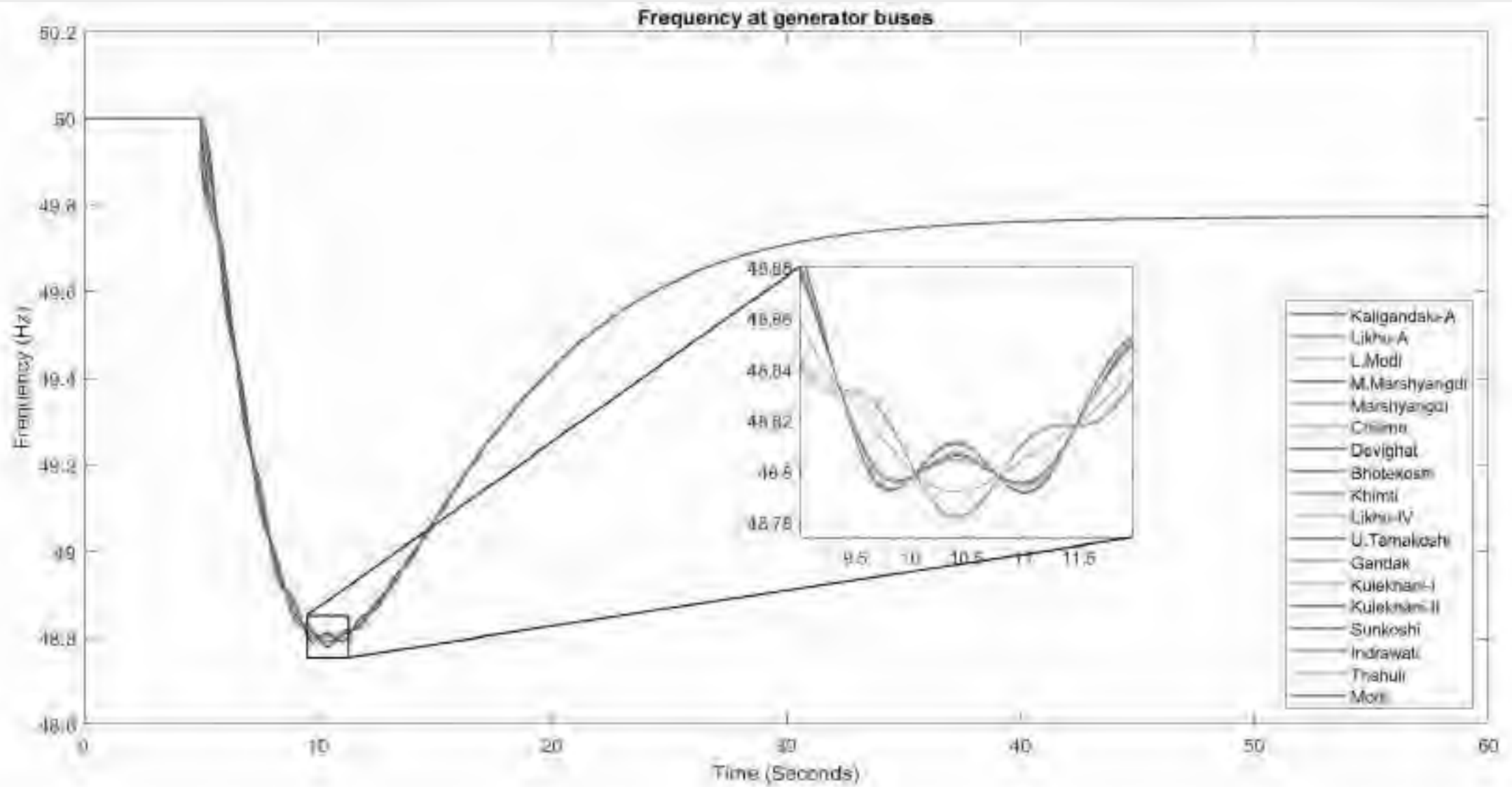


Fig 7: System Frequency Response with IEEE3 Governor

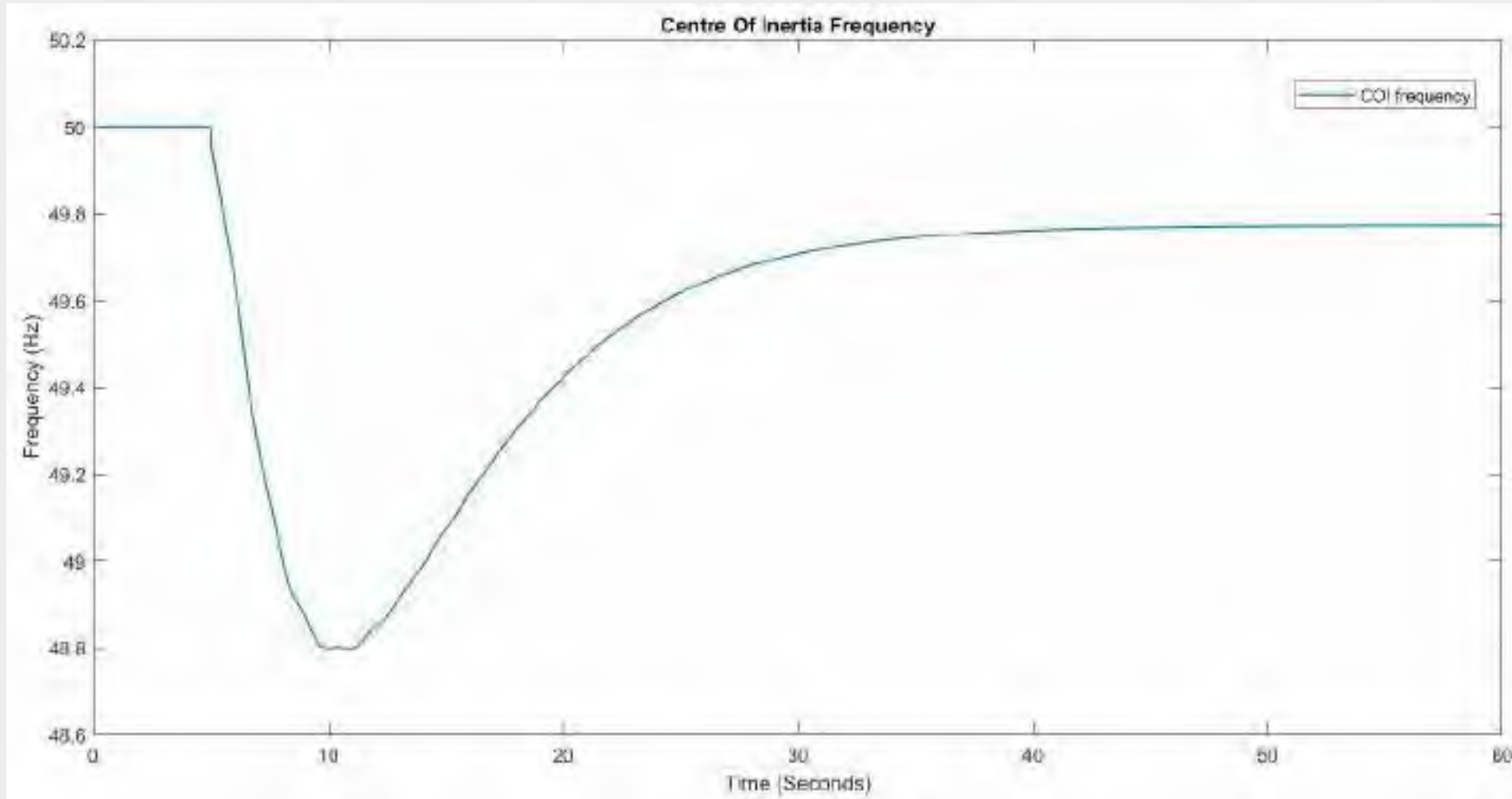


Fig 7: Centre of Inertia frequency (f_{COI}) of the system

Table 2: Frequency nadir and RoCoF of generating buses

Bus Name	f_{nadir} (Hz)	RoCoF (Hz/s)
Kaligandaki – A	48.79227	0.2013
Likhu - A	48.78238	0.2255
Lower Modi	48.79182	0.2014
Middle Marshyangdi	48.79146	0.2011
Marshyangdi	48.79254	0.2009
Chilime	48.79344	0.2493
Devighat	48.79556	0.2014
Bhotekoshi	48.79212	0.2241
Khimti	48.78258	0.2254
Likhu – IV	48.78167	0.2256
Upper Tamakoshi	48.78077	0.2258
Gandak	48.79213	0.2013
Kulekhani – I	48.79518	0.2018
Kulekhani – II	48.79487	0.2015
Sunkoshi	48.79603	0.2020
Indrawati	48.79618	0.2020
Trishuli	48.79482	0.2009
Modi	48.79185	0.2010

Table 3 : Frequency stability indices for selected system

Frequency Stability Indices	Measured point	Value
System lowest Nadir	Upper Tamakoshi	48.7077 Hz
System highest RoCoF	Chilime	0.2493 Hz/s
System frequency Nadir	COI	48.7976 Hz
System RoCoF	COI	0.2038 Hz/s

Research Opportunities

- Retrofit of Grid Connected Solar Farm Controller for frequency response support (Curtailment)
- Impact of VSM in Frequency Response “Inertial response Frame”
- Optimal loadshedding strategy
- Energy storage as an ancillary service (Pricing)

TRANSMISSION TECHNOLOGIES

WIDE AREA MONITORING SYSTEM

- PMUs can measure from 20 to 150 samples per second
- Typically it measures the **voltage and angle of the bus** where it is connected

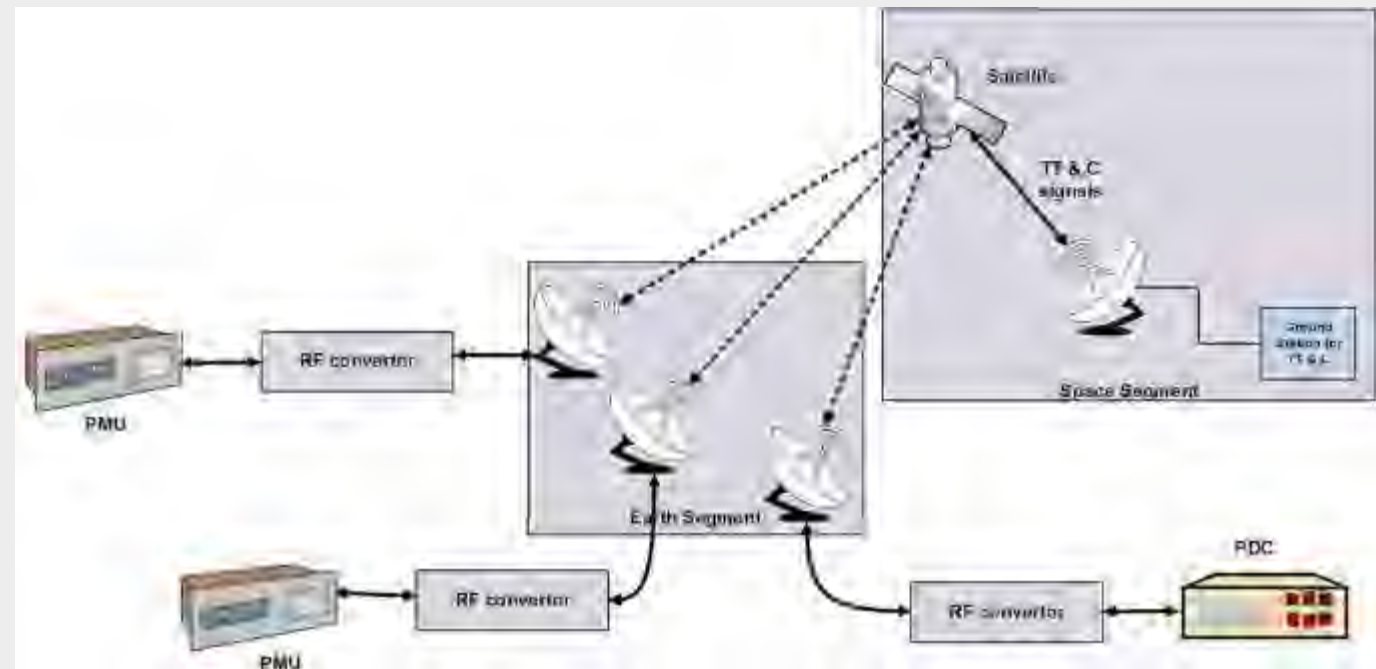
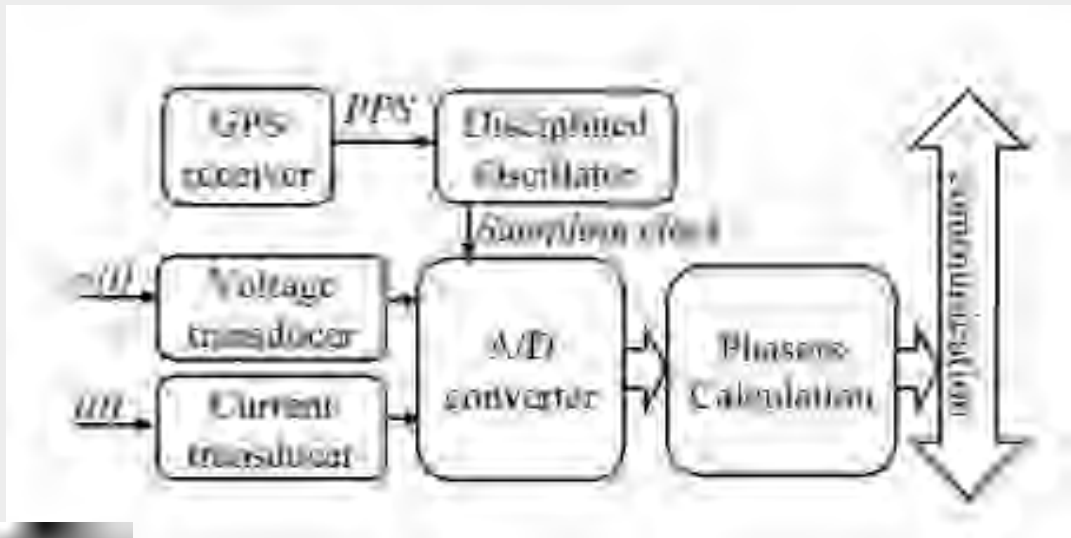




Fig. PMUs in North-American grid

- Currently, NEA is also planning to install PMU at EHV substations
- PMU data can be used for many purpose including **faster estimation/ prevention of stability issues (STVS)**

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IEEE TRANSACTIONS ON POWER SYSTEMS, VOL. 28, NO. 4, NOVEMBER 2013

Real-Time Monitoring of Short-Term Voltage Stability Using PMU Data

Sambarta Dasgupta, Magesh Paramasivam, Umesh Vaidya, *Member, IEEE*, and Venkataramana Ajjarapu, *Fellow, IEEE*

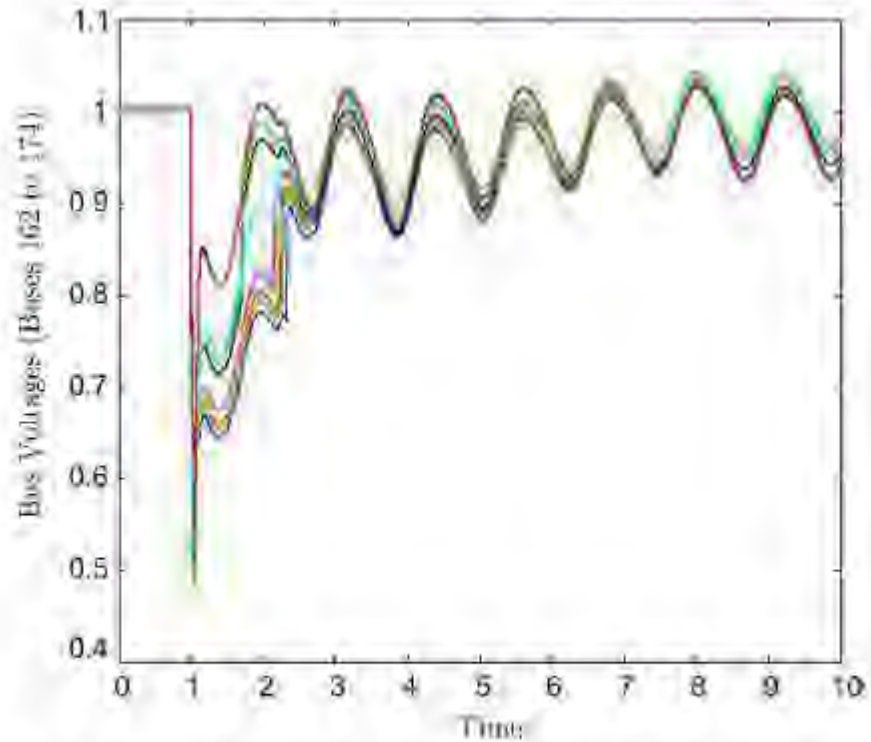


Fig. 2. Evolution of bus voltages (for buses 162–174) for clearing time $t_{cl} = 1.080$ s for IEEE 162-bus system.

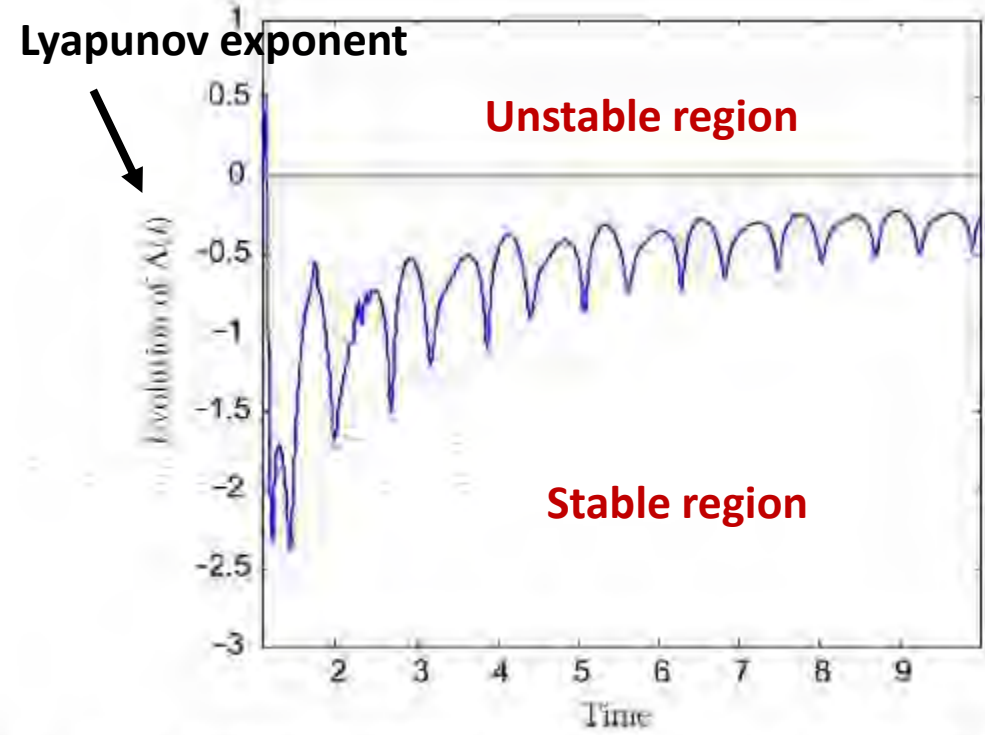


Fig. 3. Evolution of exponent for clearing time $t_{cl} = 1.080$ s for IEEE 162-bus system.

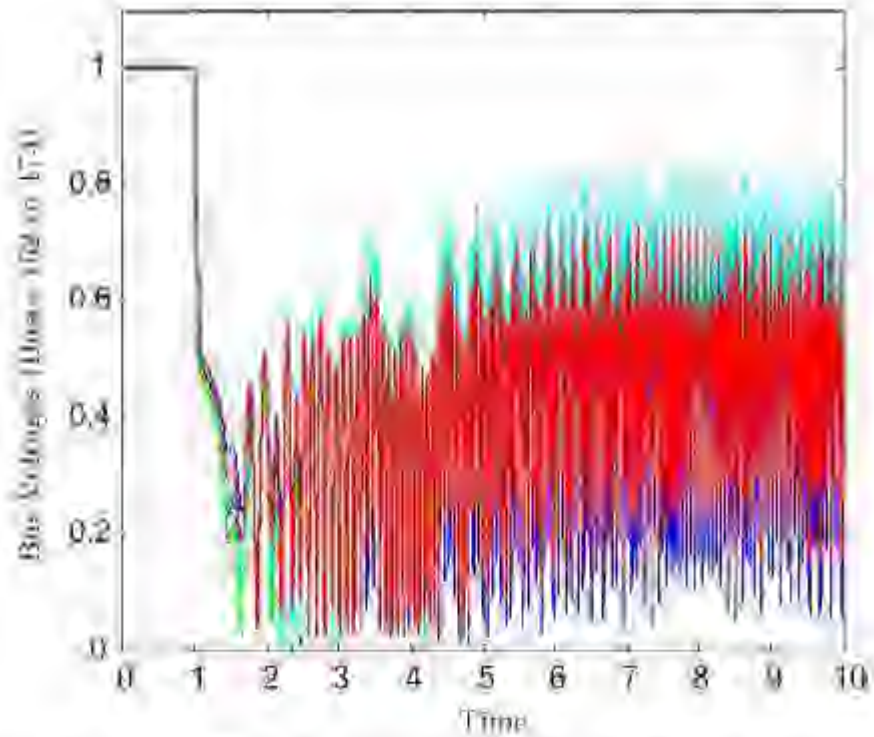


Fig. 4. Evolution of bus voltages (for buses 162-174) for clearing time $t_{c0} = 1.38$ s for IEEE 162-bus system.

Lyapunov exponent

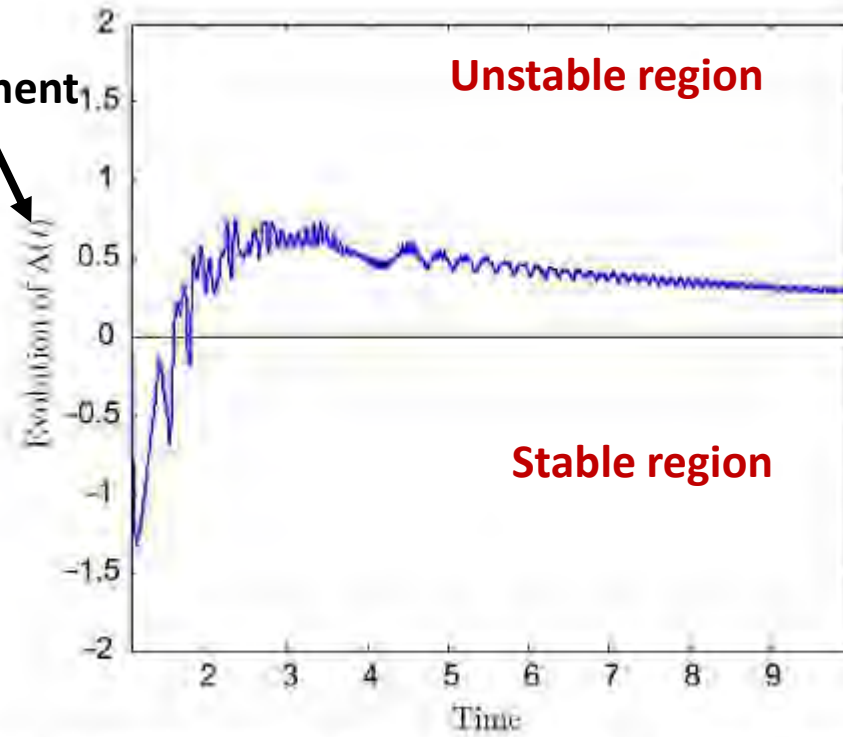


Fig. 5. Evolution of exponent for clearing time $t_{c0} = 1.38$ s for IEEE 162-bus system.

TABLE II
CRITICAL CLEARING TIME COMPUTED FOR IEEE 162-BUS SYSTEM

Fault Case	Fault Location	Trip Line	CCT (in sec)
ν	1	1-2	0.126
α	5	5-1	0.154
σ	26	26-25	0.104
π	120	120-5	0.175
ψ	120	120-112	0.140
ω	129	129-5	0.207

Based on Data Length Considered

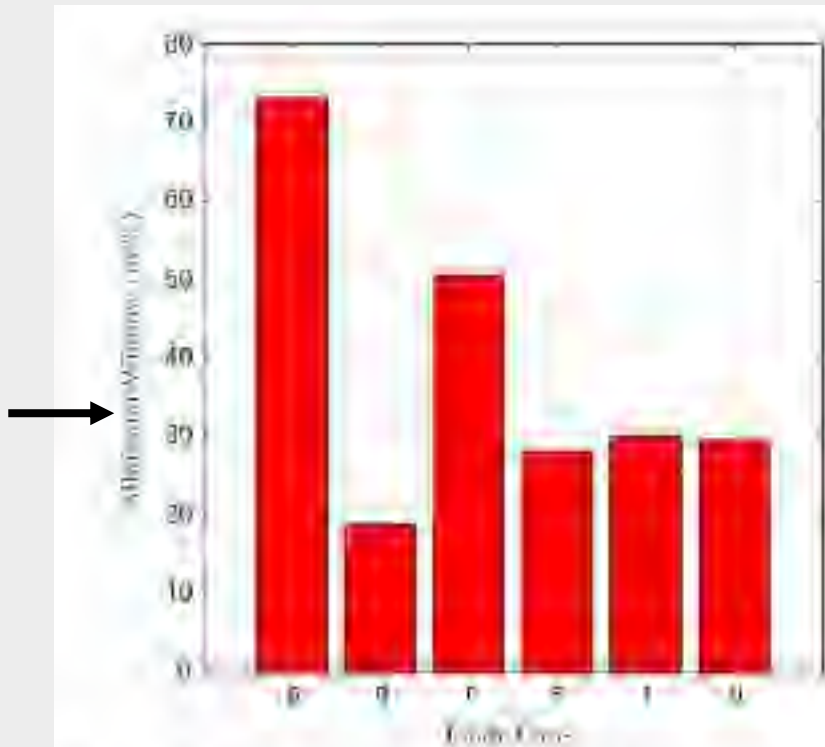


Fig. 1. Maximum voltage levels for different fault locations for IEEE 162-bus system (Critical fault location corresponding to bus-cases listed in Table II.)

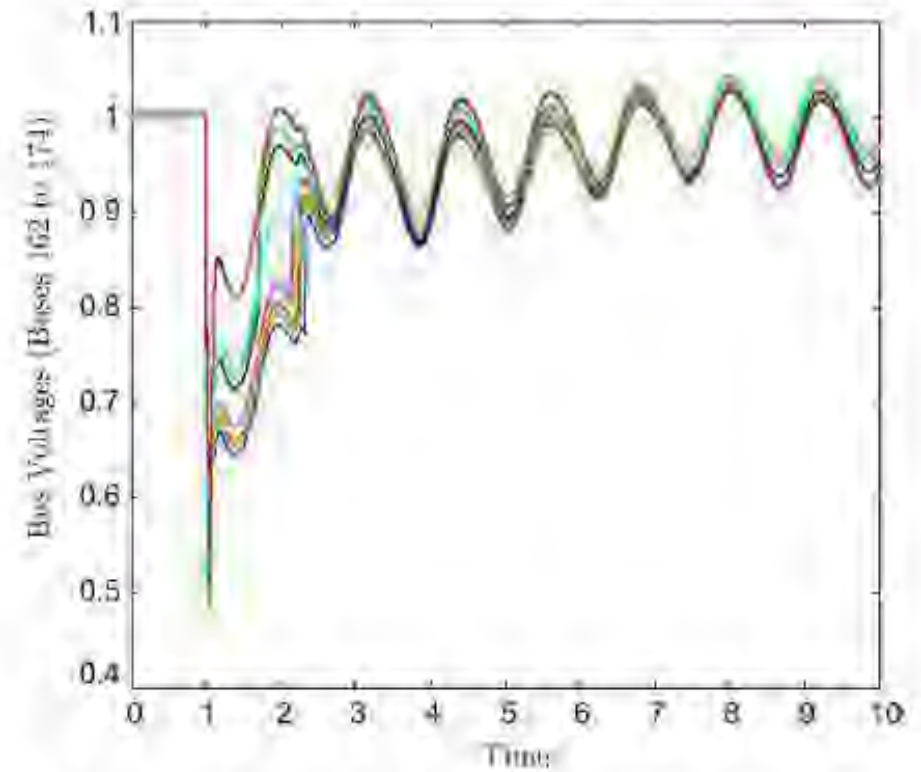


Fig. 2. Evolution of bus voltages (for buses 162-174) for clearing time $t_{cl} = 1.080$ s for IEEE 162-bus system.

IEEE 9 bus system results

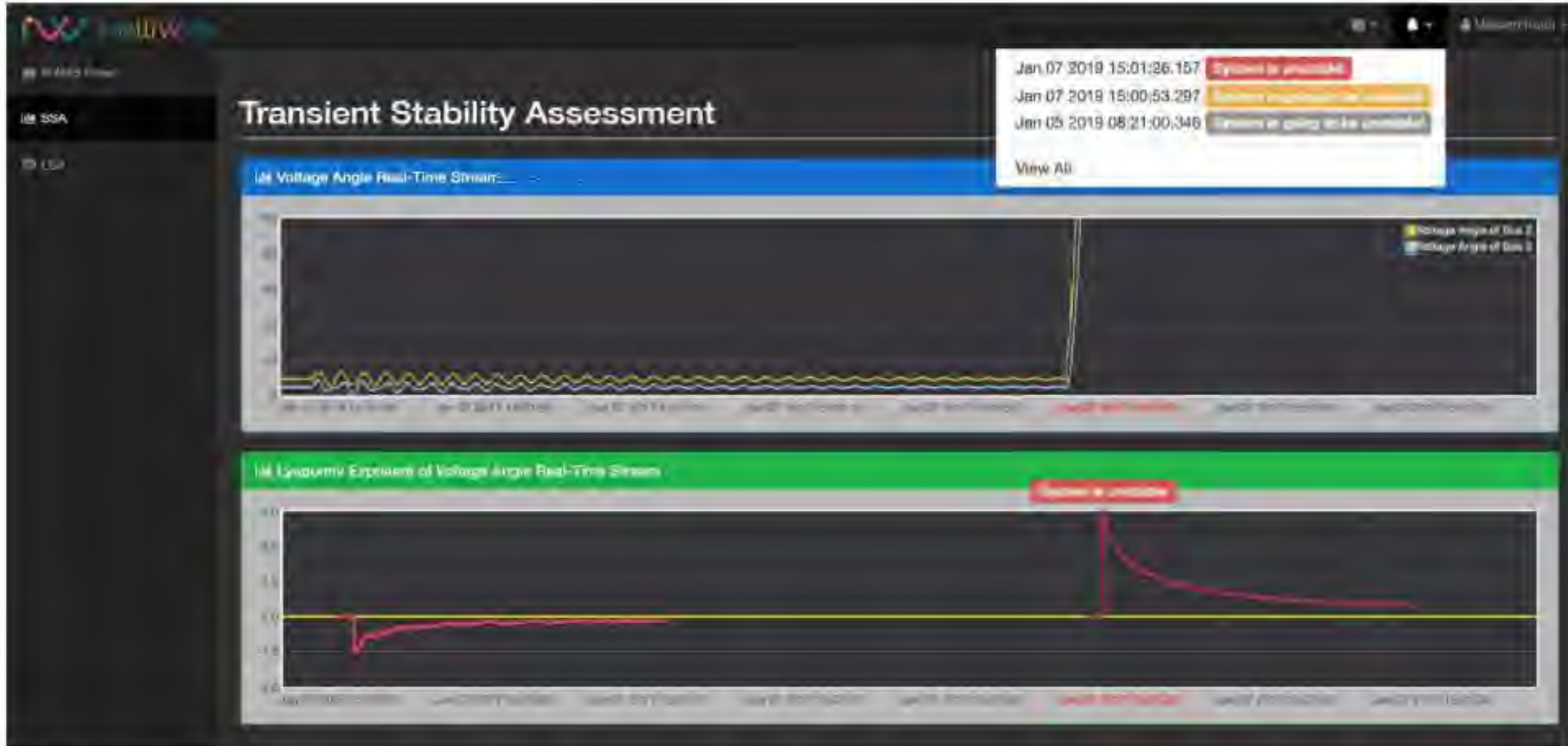


Fig. 11 GUI for TSA (Scenario 2)

IEEE 9 bus system results

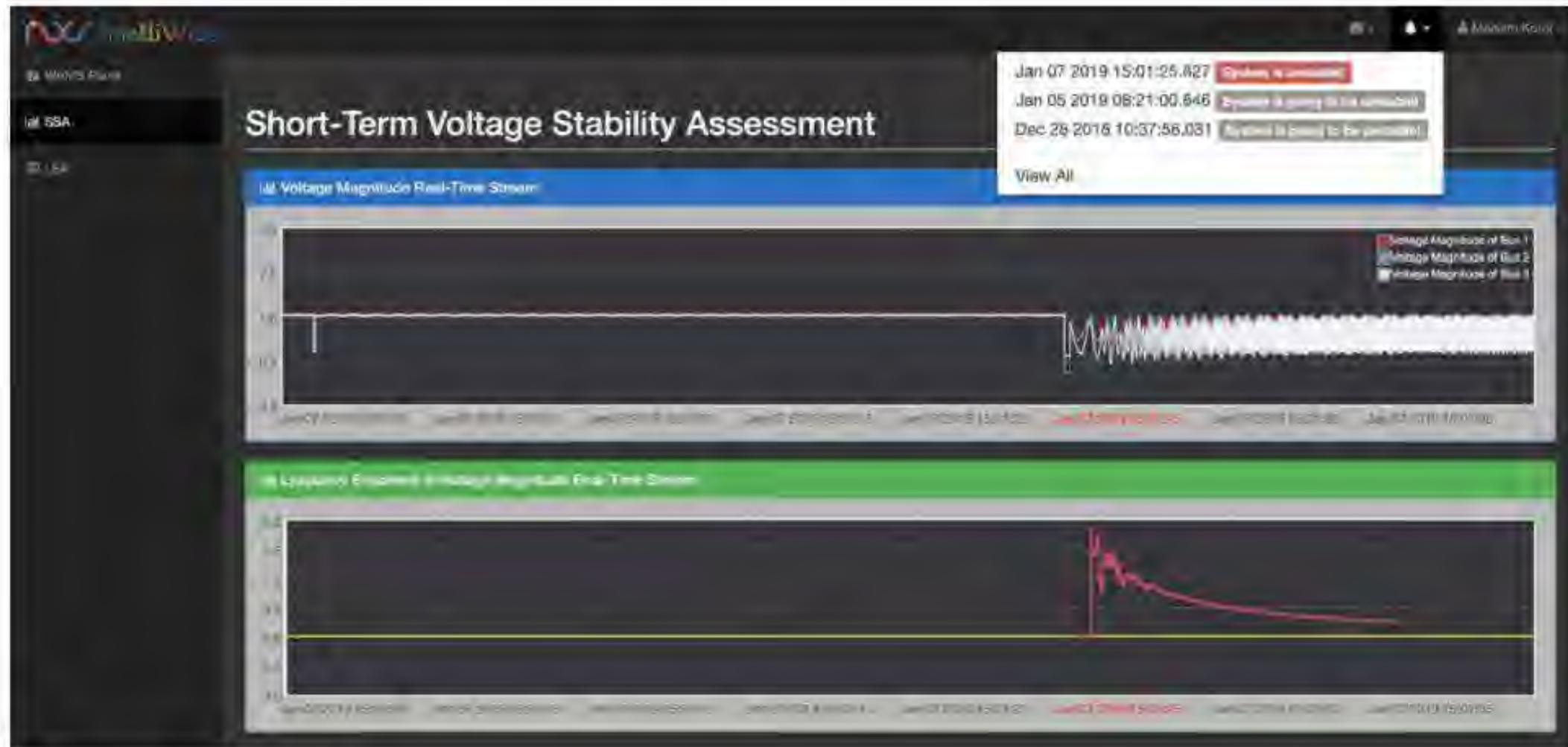


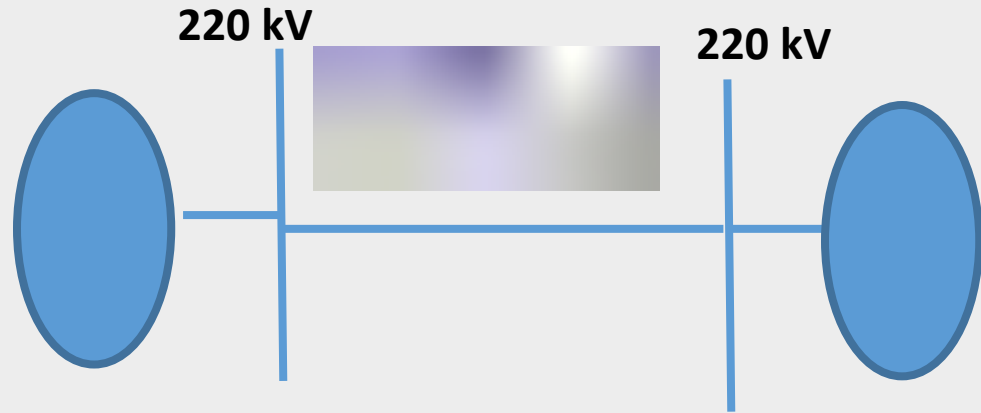
Fig. 12 GUI for SVSA (Scenario 2)

- **Challenges remaining**

1. Develop a tool for real-time monitoring and early warning system based on PMU measurements
2. How to make system more observable with limited number of PMUs?
3. Effect of PMU noise and communication latency in estimation
4.

DYNAMIC LINE RATING ESTIMATION

- Over head lines normally are limited by **thermal limit (Heat)**
- They are usually sized based on worst criterion (**65 degree celsius, low wind velocity and full radiation**)



Ampacity= 506 A

Thermal limit = $\sqrt{3}$ *220* 506 = 193 MW

✓ However, conductor can carry more current when the solar radiation is low and wind velocity is high
(Cooling effect)

Code Wind	Current Carrying Capacity		
	0° Deg. C	75 Deg. C	90 Deg. C
	(Amps)	(Amps)	(Amps)
Booster	707	400	145
Forest	185	155	35
Alm.	187	108	50
Horn	166	135	16
Bever	105	105	20
QHE	307	257	38
Co	250	25	25
Leopard	315	375	110
Suzale	298	357	12
Tiger	311	373	150
Tryo	305	475	30
100	465	574	600
	0	0	0
Gold	345	724	170
Shore	255	300	120
Dev	722	295	610
Elk	787	970	1015
Garbat	707	970	1015
Suzrov	110	140	174
Fox	115	144	180
Guard	38	111	38
100	500	110	100

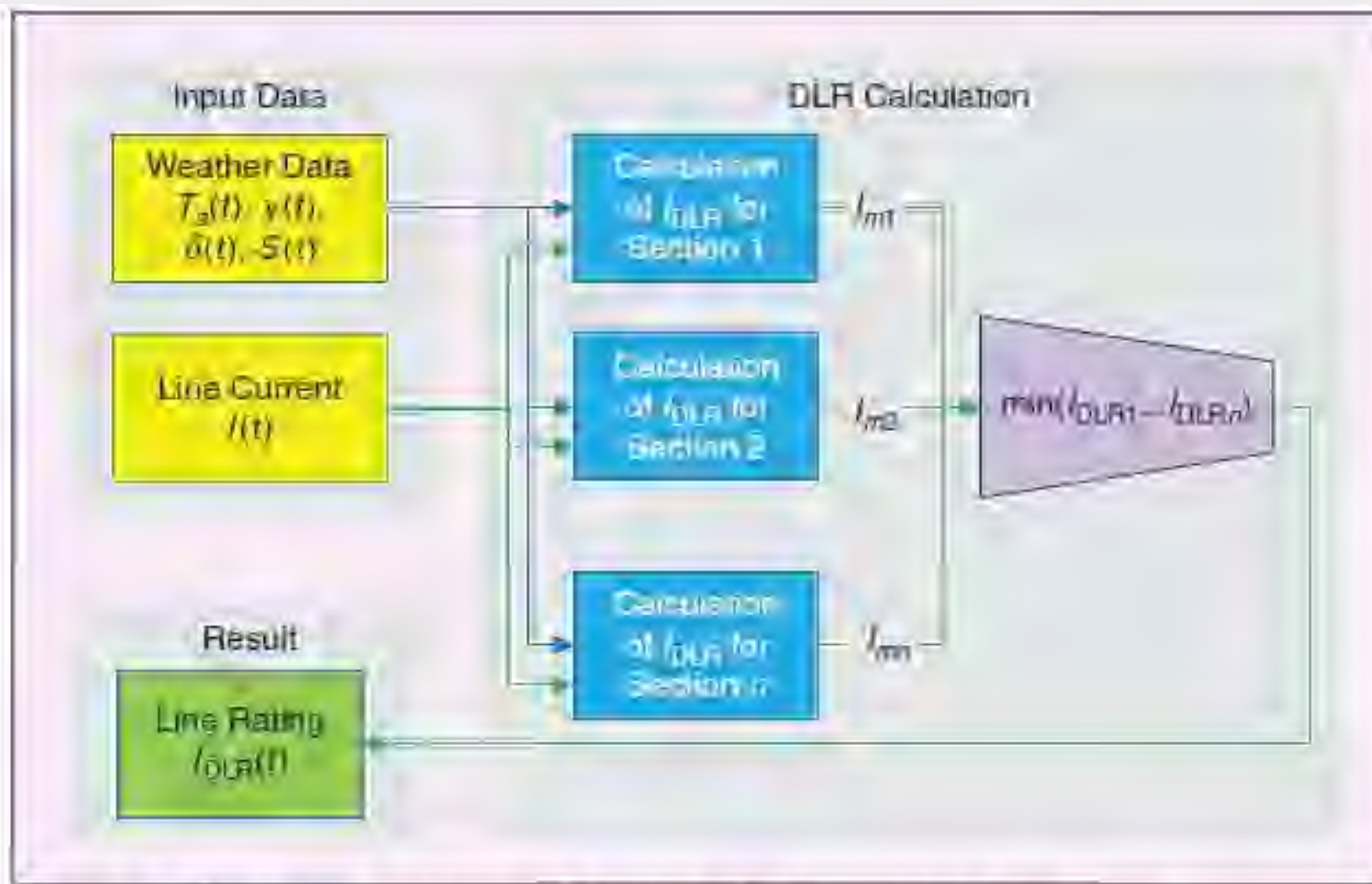


Figure 9. The general principle of dynamically determining line rating.

Technology developed by the Slovenia grid

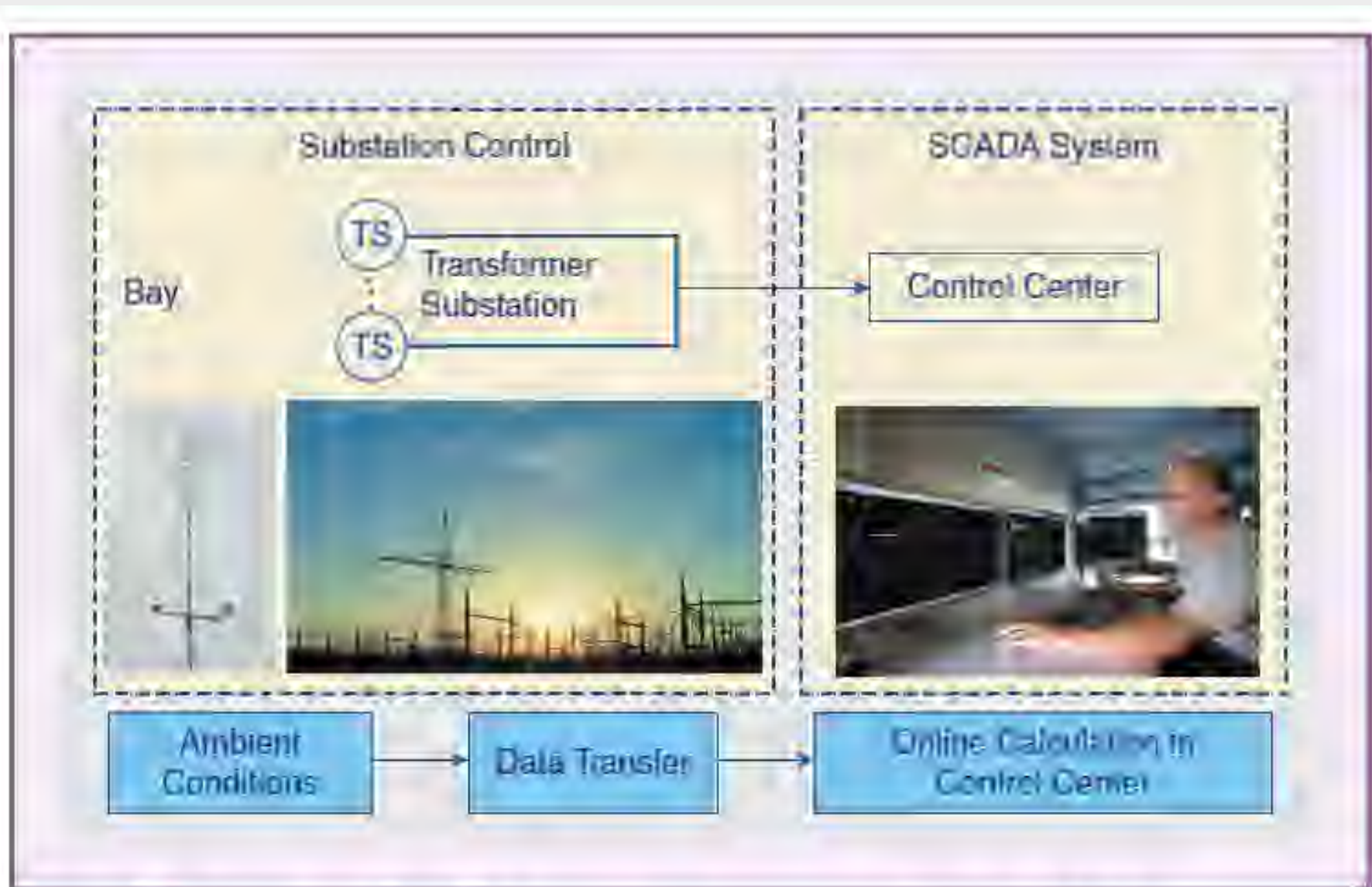


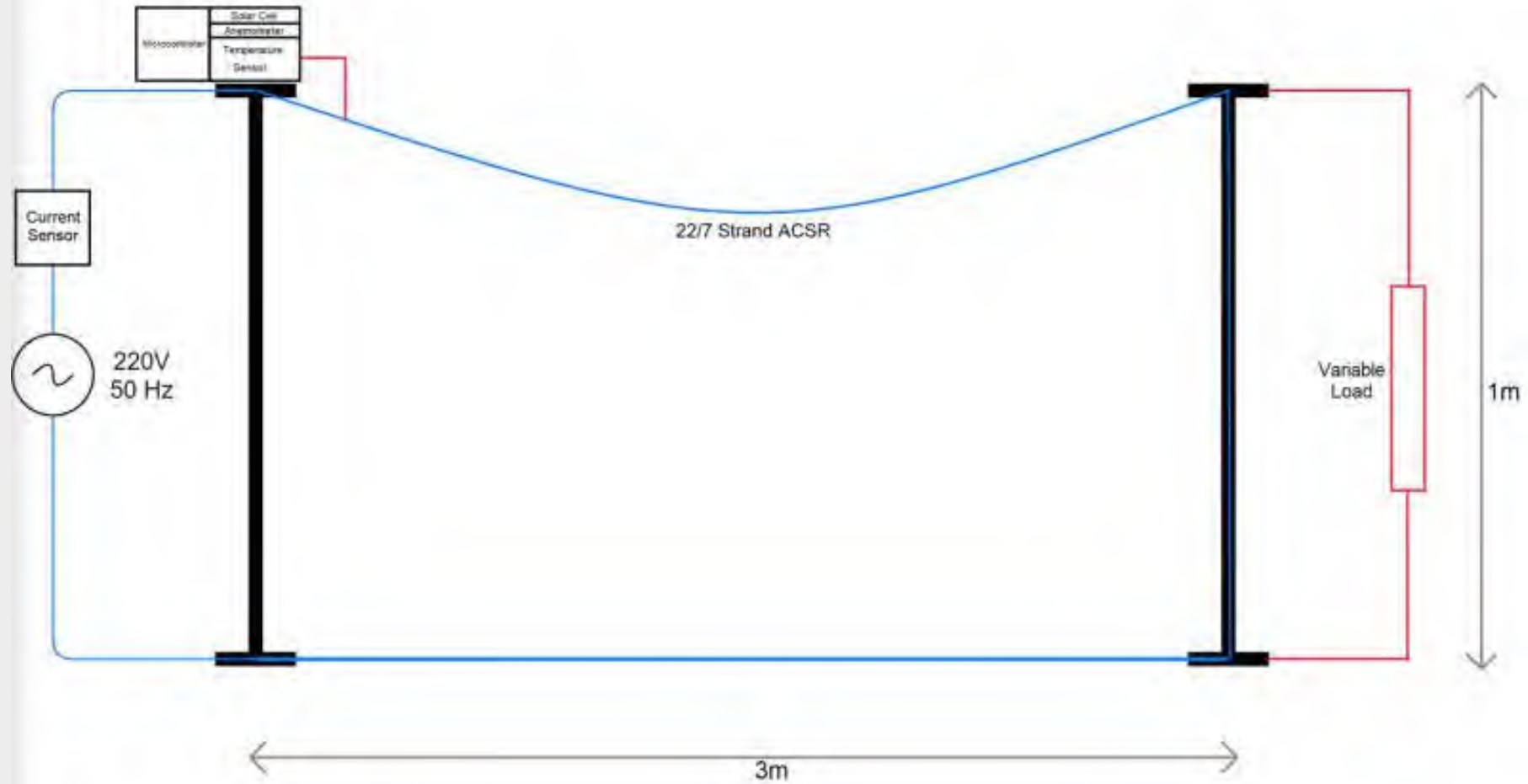
Figure 15. The principle of the dynamic rating system for 380-kV OHLs.



Figure 16. The dynamic current rating compared with the static rating over one month for TenneT OHLs.



Figure 14. Application of dynamic line rating in Germany.



Undergraduate project work of
Phurba, Sristhi and Polarj (Running), KU

ADVANCEMENTS IN DISTRIBUTION SYSTEMS

STAYING BIG OR GETTING SMALLER

Expectations for change in energy systems made possible by [unclear] of [unclear]

today

tomorrow



centralized generation

generation



decentralized generation



central energy distribution

market



decentral energy distribution



high-voltage power line and system

transmission



including direct local transmission and supply



one-way

distribution



two-way



passive consumers

consumer



active consumers

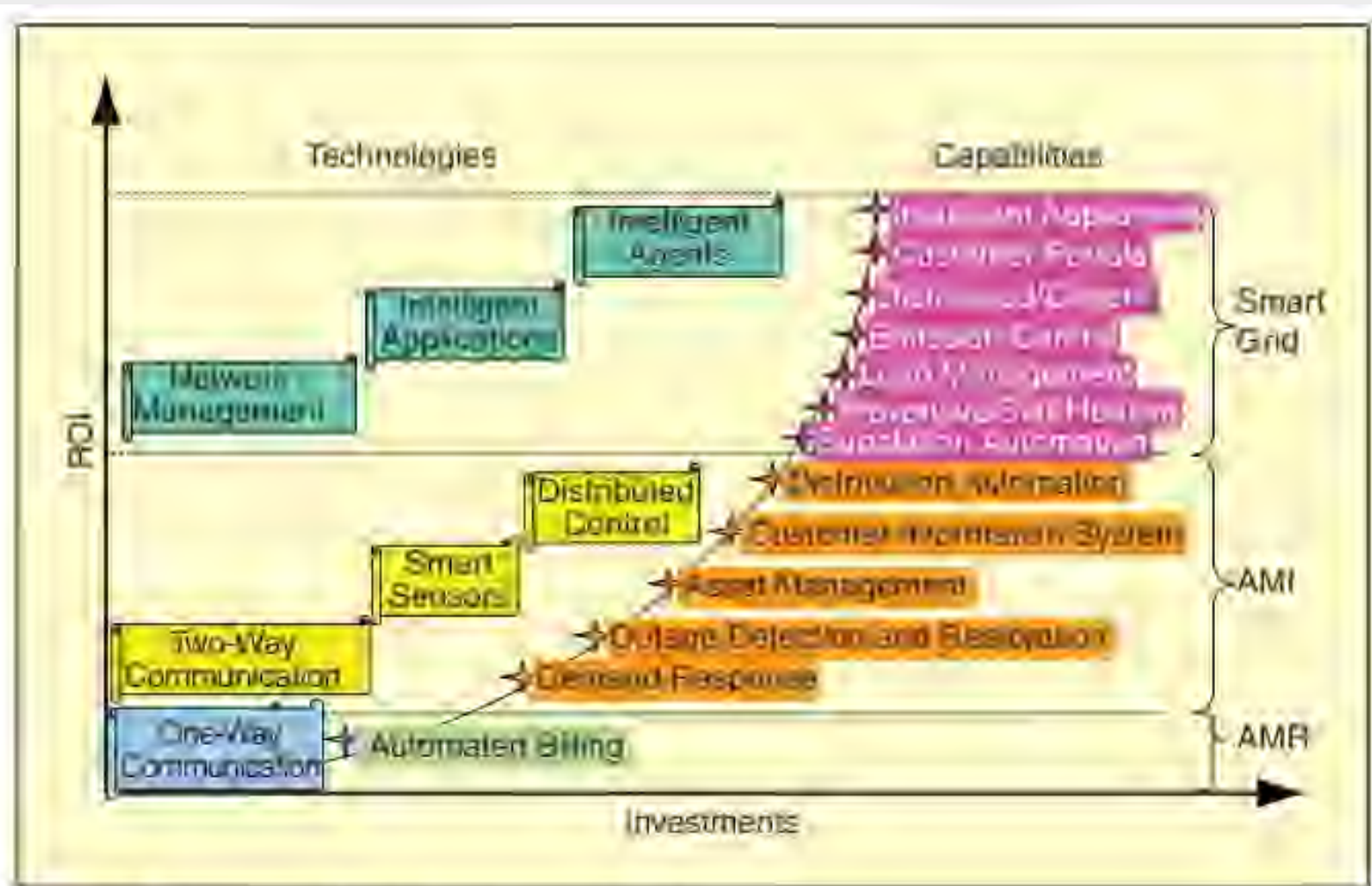
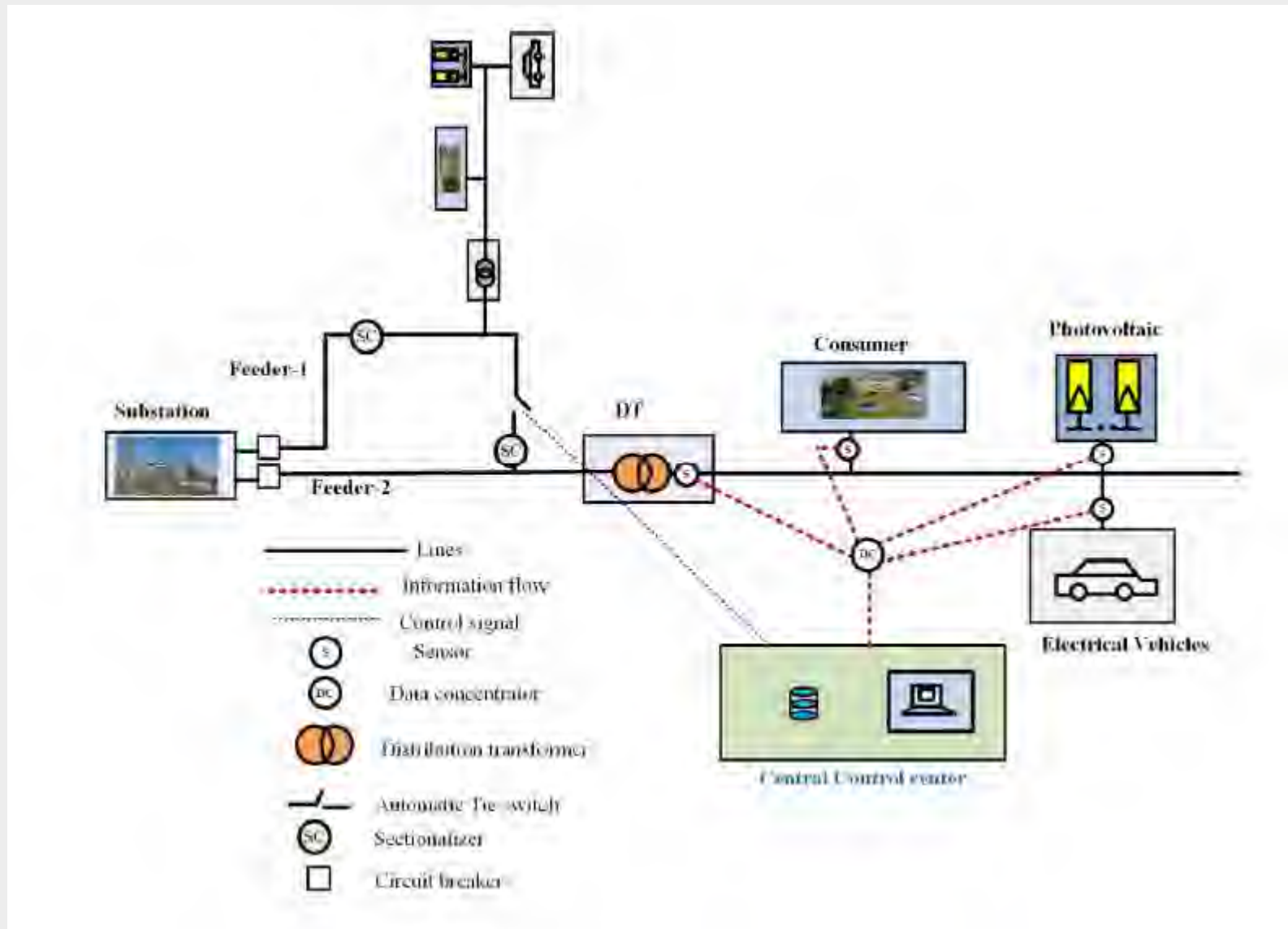


Figure 7. Smart grid return on investments.

- Distribution Grid Visualization (Ongoing Project)



- Transmission Grid Visualization



<https://www.rechargenews.com/transition/the-final-piece-of-the-puzzle-how-will-the-future-grid-work-/2-1-515137?zephrossoott=cKZMLw>

• Distribution Grid Visualization



Smart meter

<https://techlekh.com/nea-smart-electricity-meters/>



Micro phasor measurement unit

<https://powerside.com/products/power-monitoring/micropmu/>

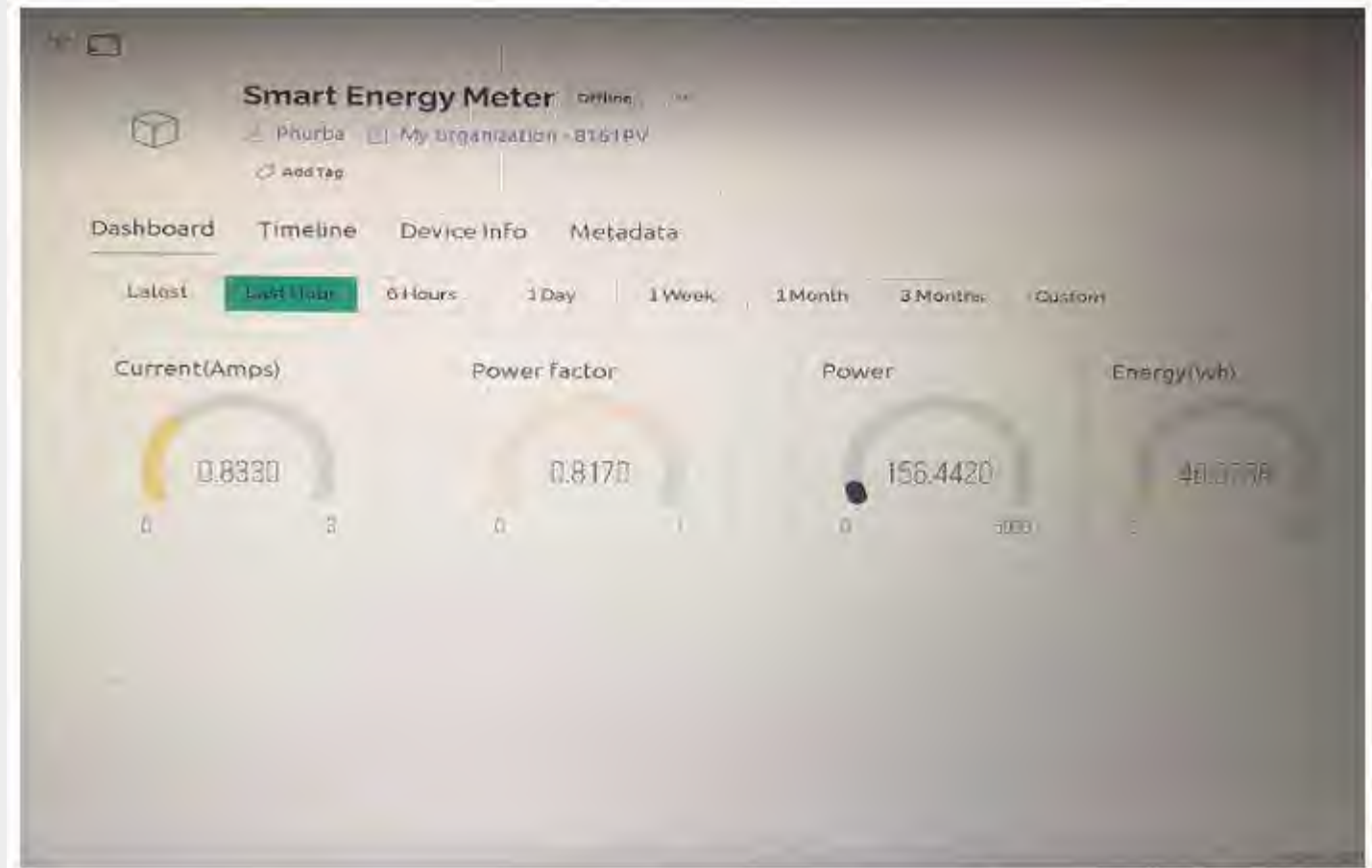
- Distribution Grid Visualization (Ongoing Project)



- Distribution Grid Visualization (Ongoing Project)

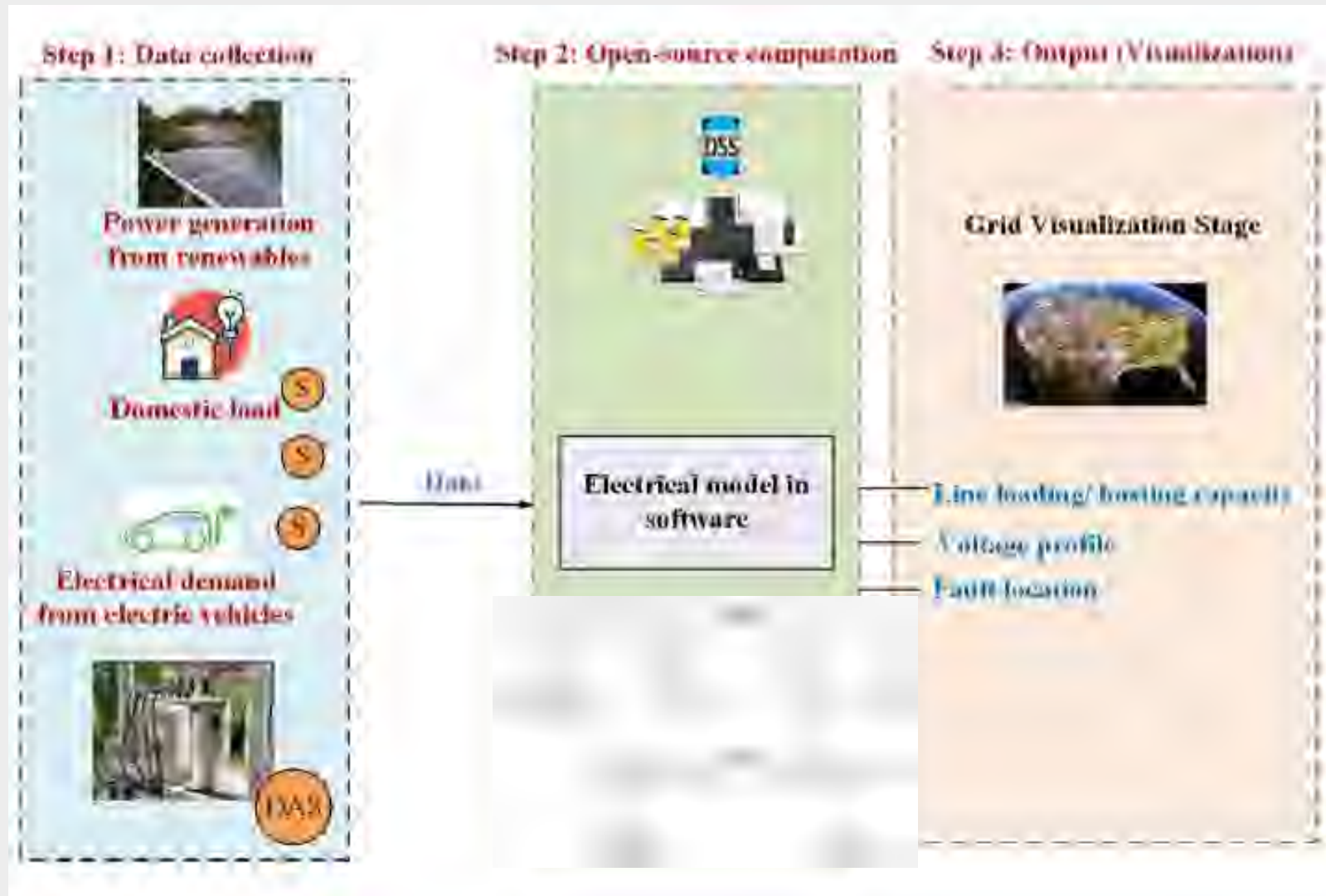


Fig 13: Circuit Setup of SEM



Undergraduate project work of
Phurba, Sristhi, Abhisekh and Prakash, KU

- Distribution Grid Visualization (Ongoing Project)



- Distribution Grid Visualization

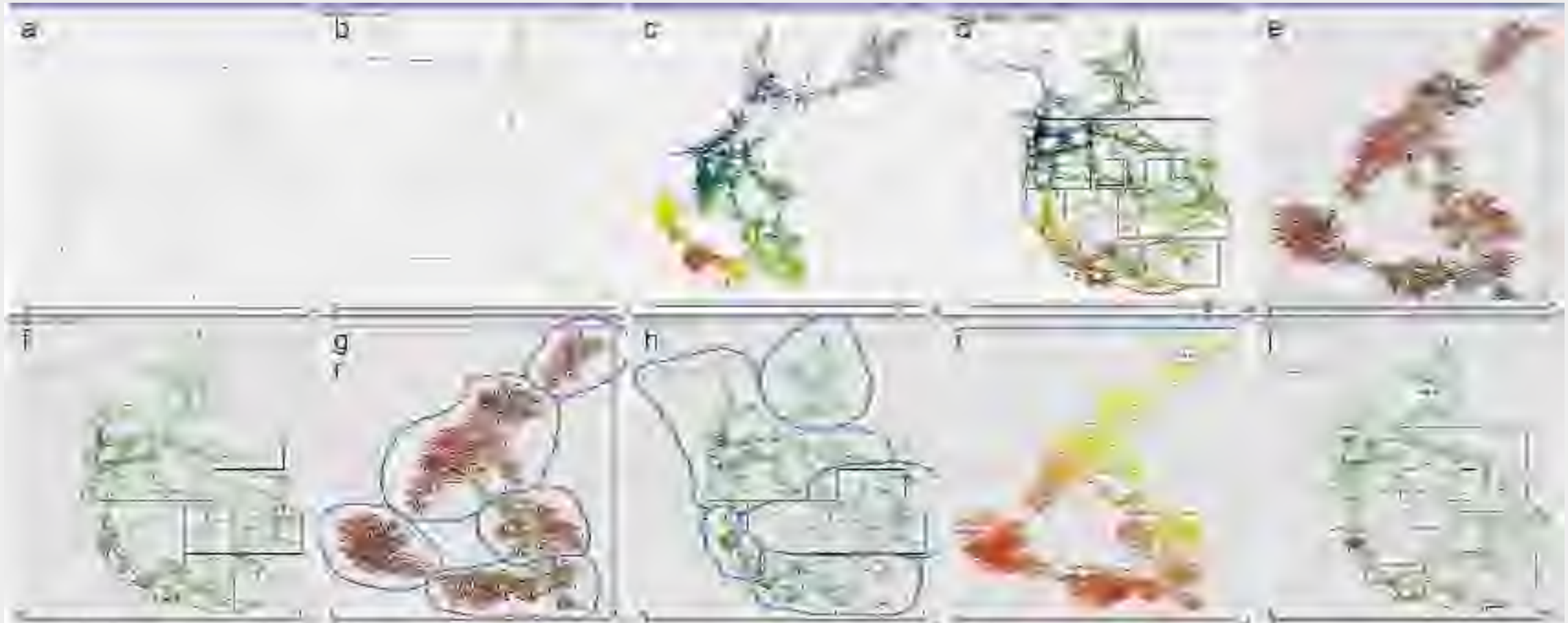
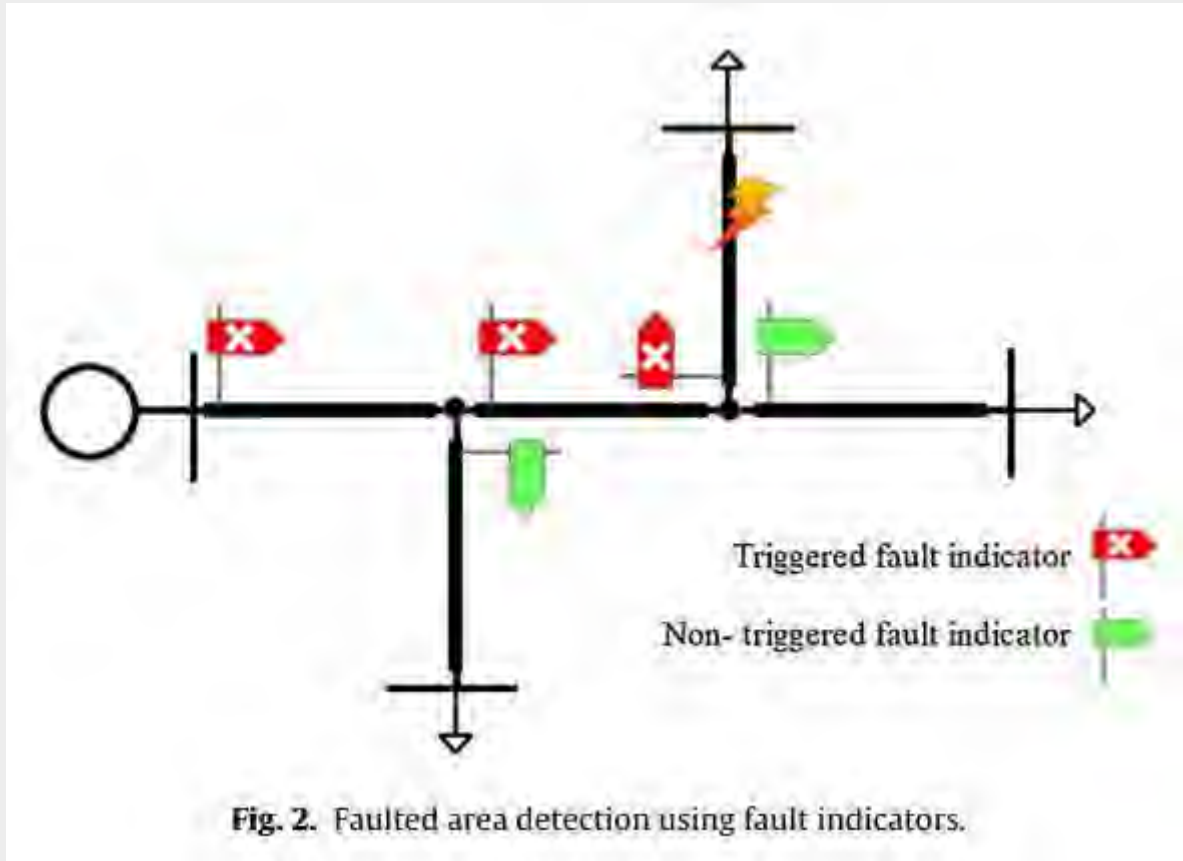


Figure 7. Sample set of GreenGrid and nematic-like layout visualizations applied used in usability evaluation. Grids show simulated electric power

<https://www.semanticscholar.org/paper/A-Novel-Visualization-Technique-for-Electric-Power-Wong-Schneider/36c2e19e3d9fd4b10e3fe9d9b0422c979f0bafae>

- Distribution Automation -> Outage Management System



- Distribution Automation (Ongoing Project) -> Outage Management System

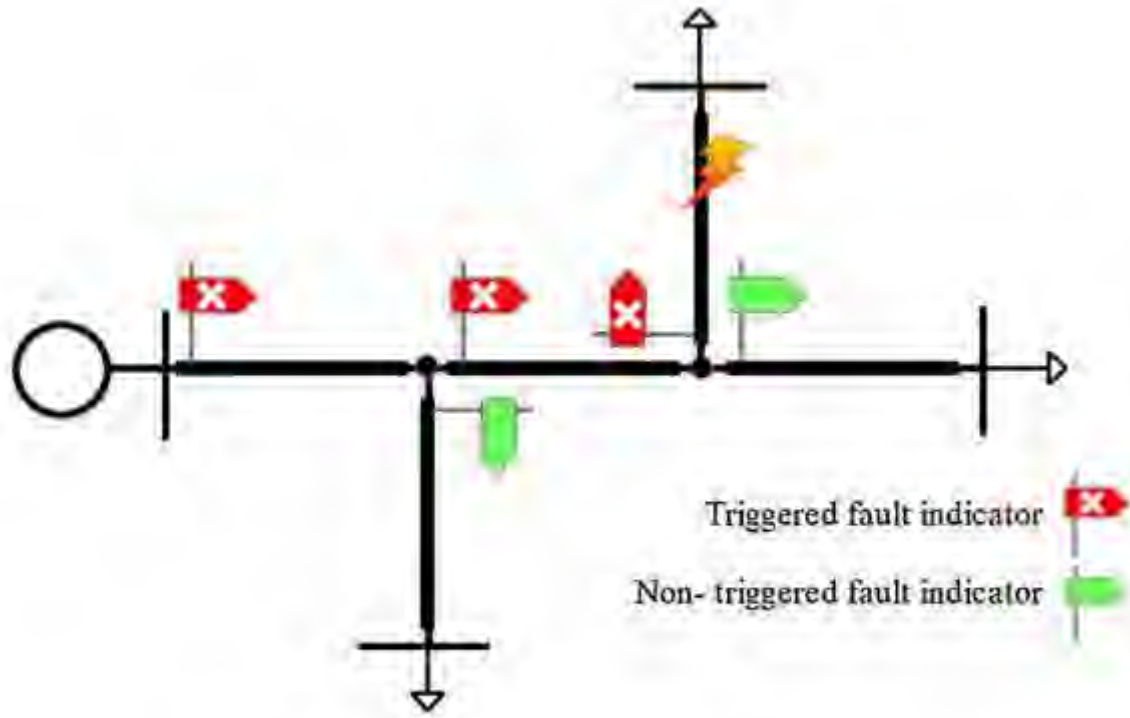
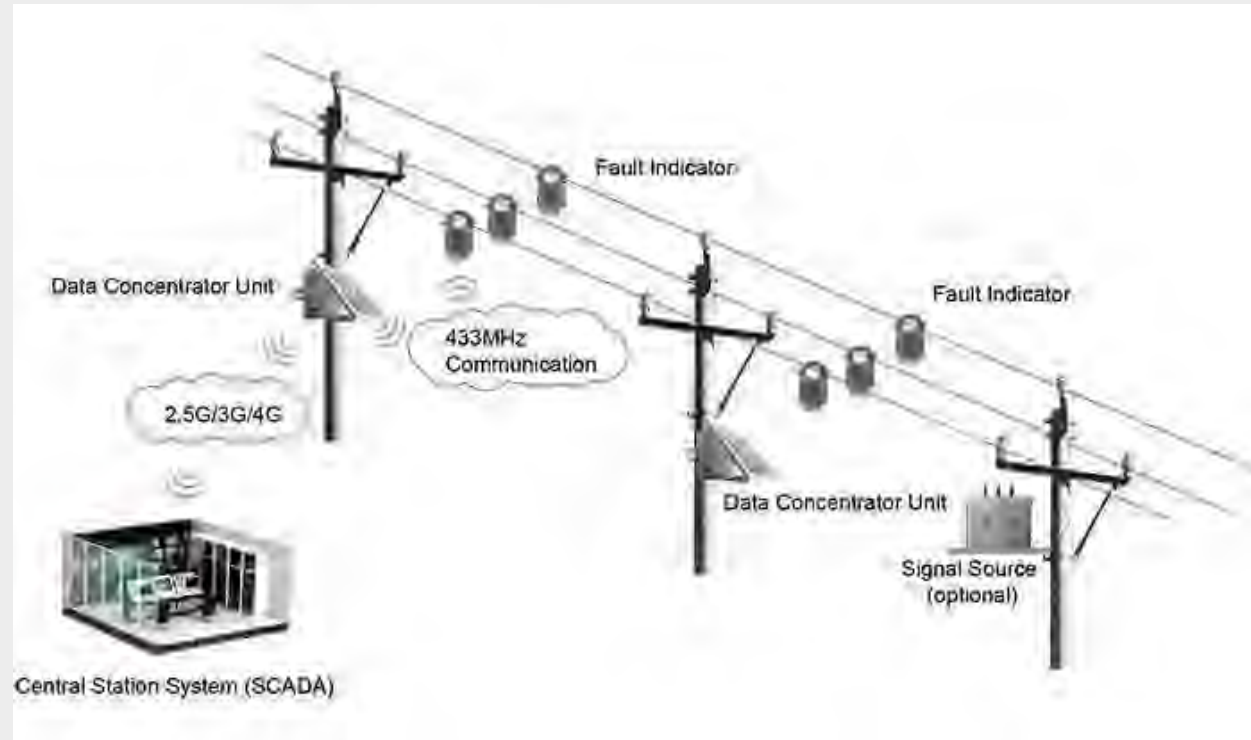
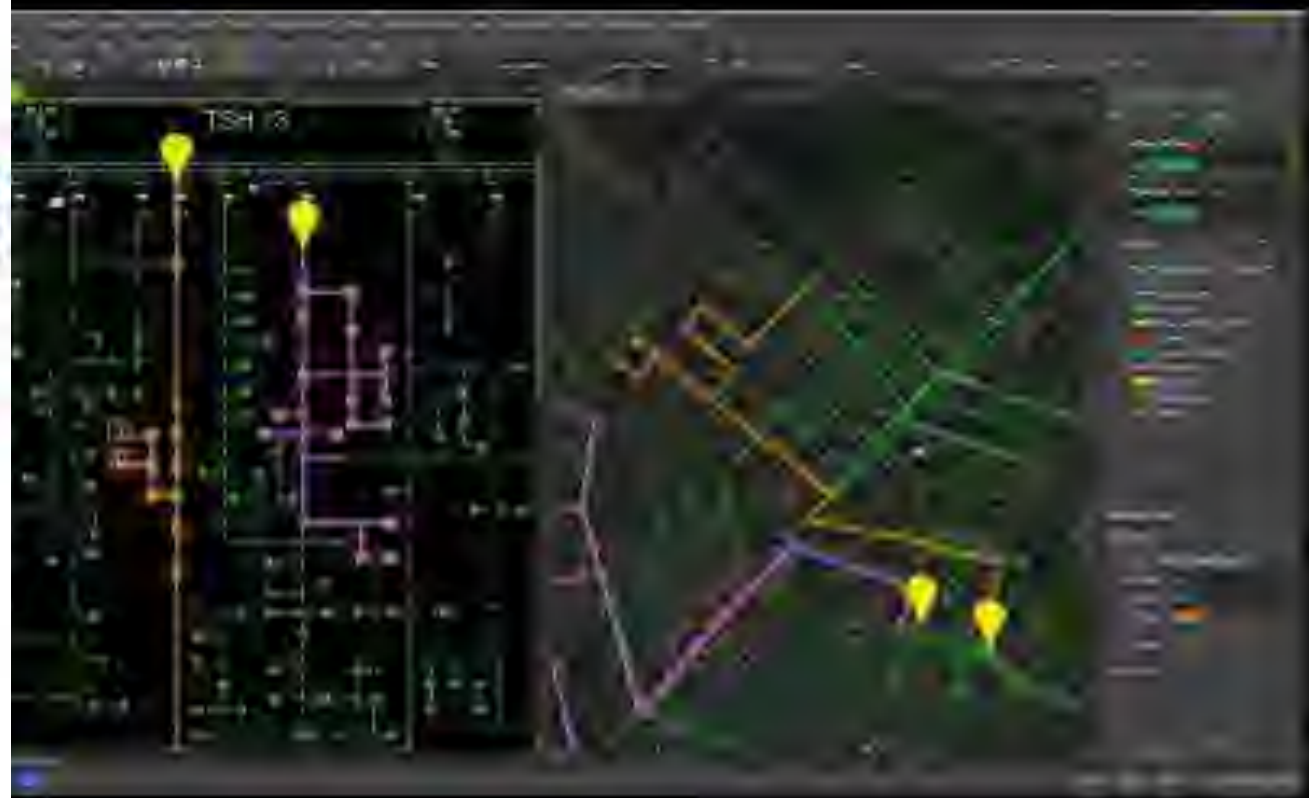
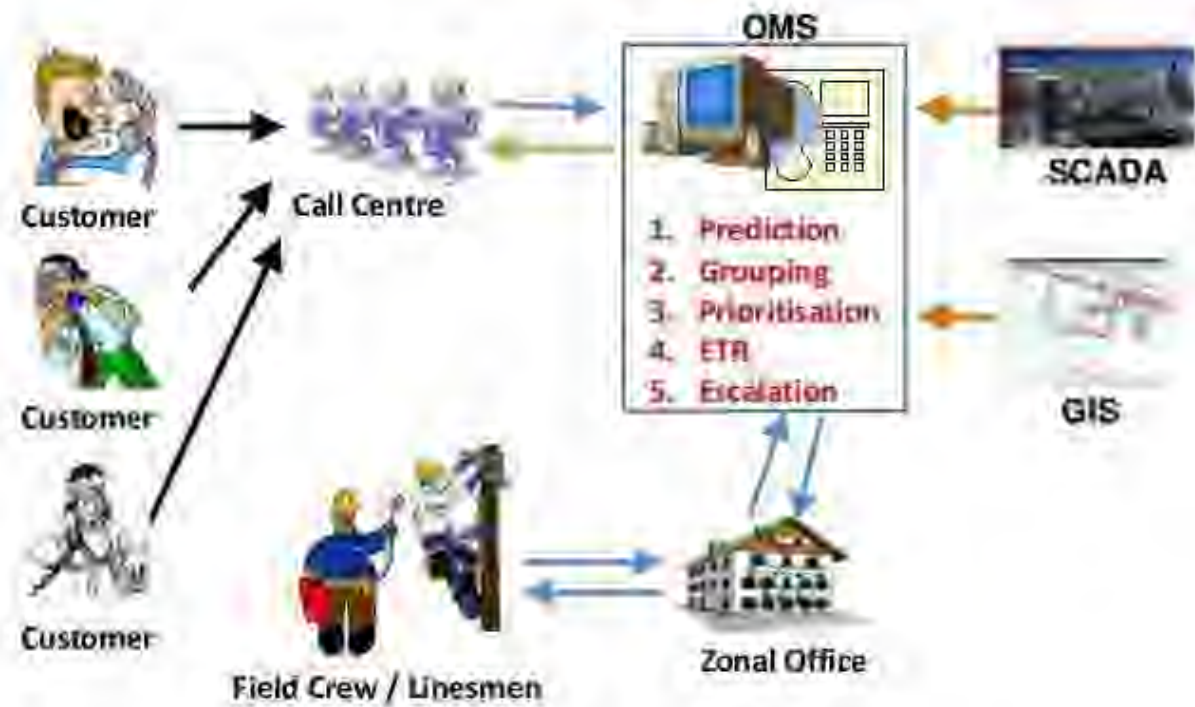


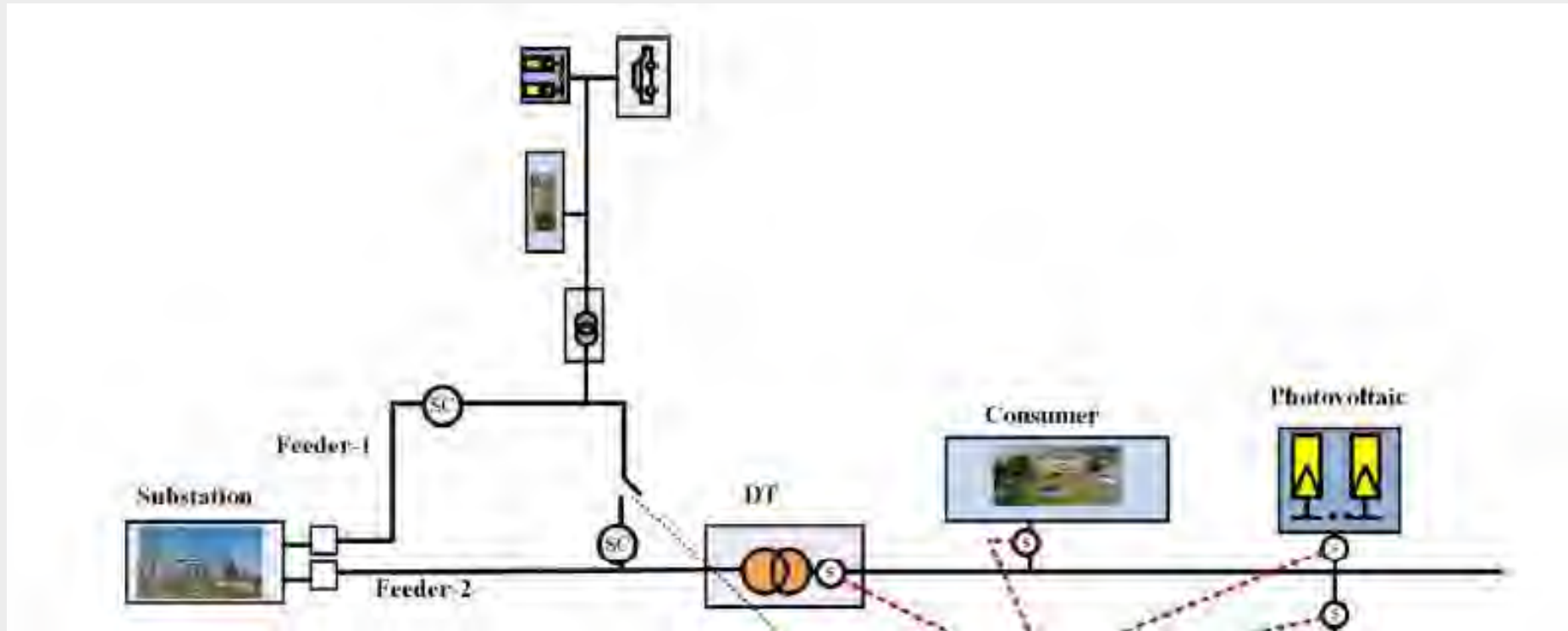
Fig. 2. Faulted area detection using fault indicators.



Outage Management System



- **Distribution/ Feeder Automation (In lab)**
- More complex algorithm/ AI (maybe) required in complex system (Mesh, loop etc)



Lab model

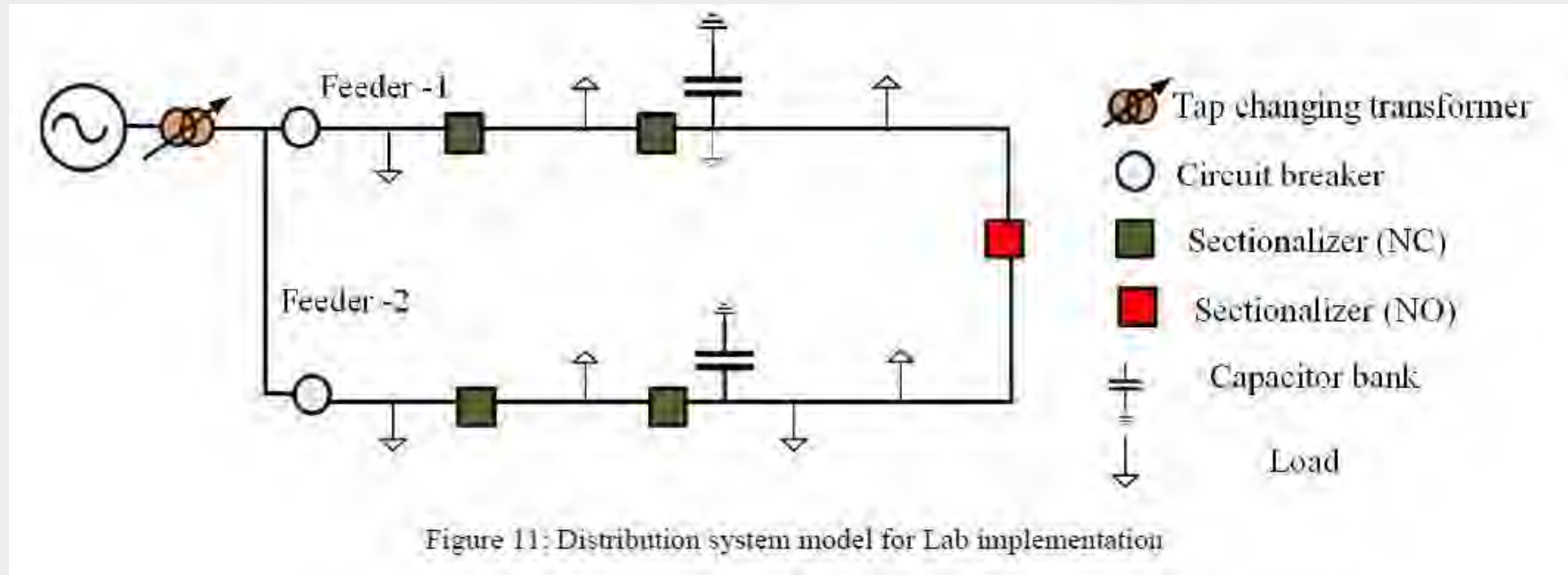


Table 2: Equivalent distribution components for DA lab implementation

Components	Equivalent
Supply (11kV)	12V (AC)
Feeder	Resistor and inductor
Sectionalizer	Contact
Load	Resistive bank



Figure 42: Monitoring the system status from Web app.

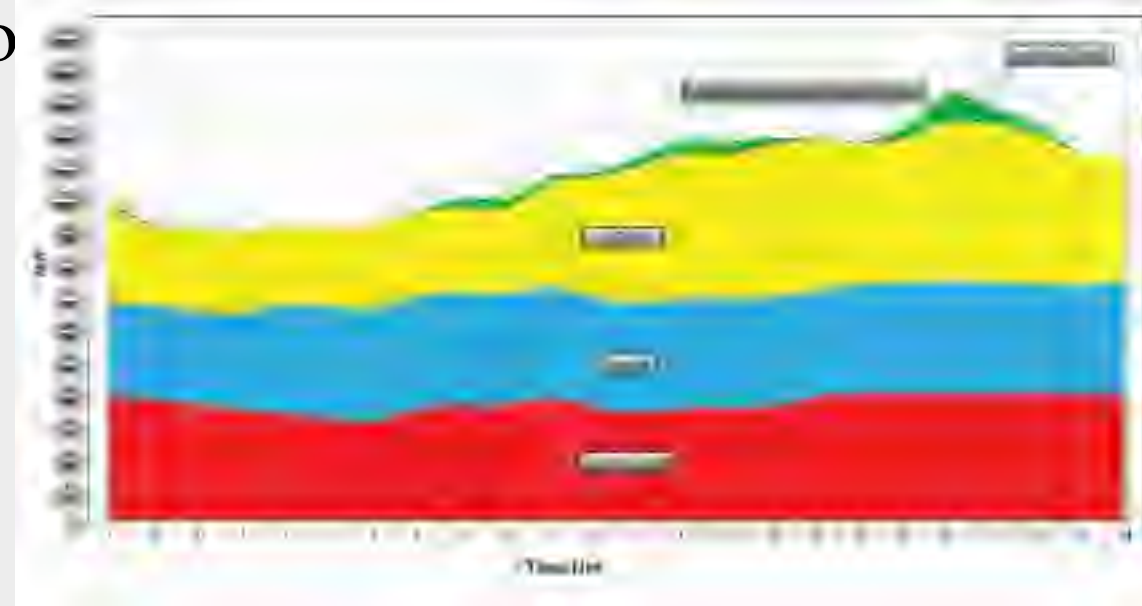
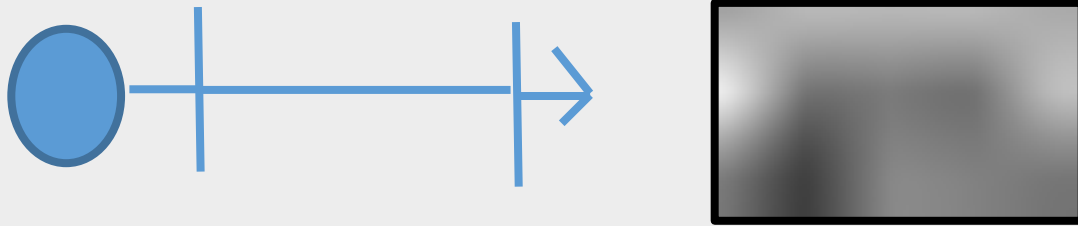


Figure 43: Overall hardware

Undergraduate project work of
Yubraj, Aship and Abhijeet, KU

PROBABILISTIC METHODS

- Provide statistical information about system → Better tool for decision making
- Better method to analyze the effect of (renewable, load etc)



- **Task:** Find **voltage profile** for load for **every hour**
 - **Solution:** Take hourly load data of every hour and run the loadflow (Newton Raphson)
- **Tedious work**

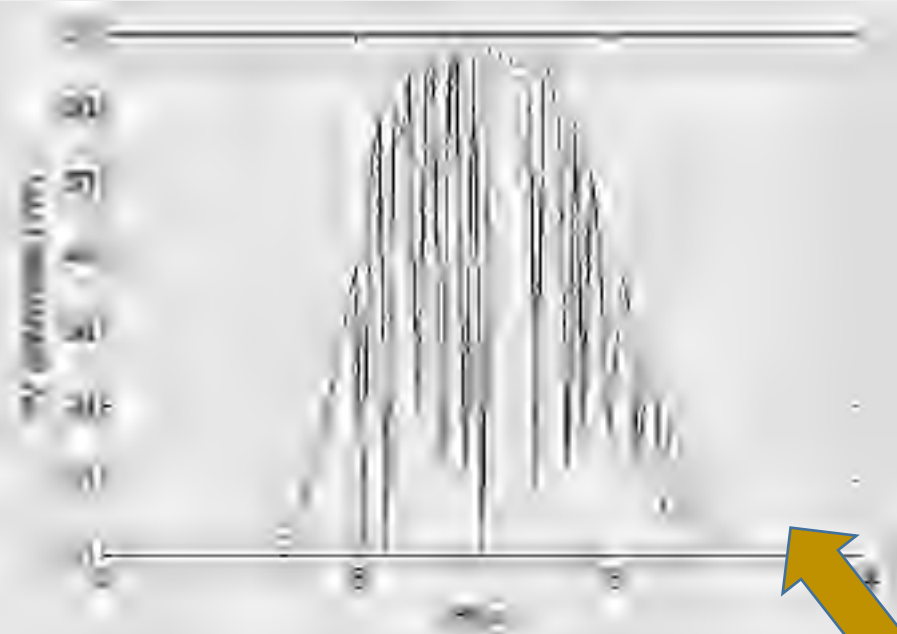


Fig. PV output power [Fan et al. (2012)]

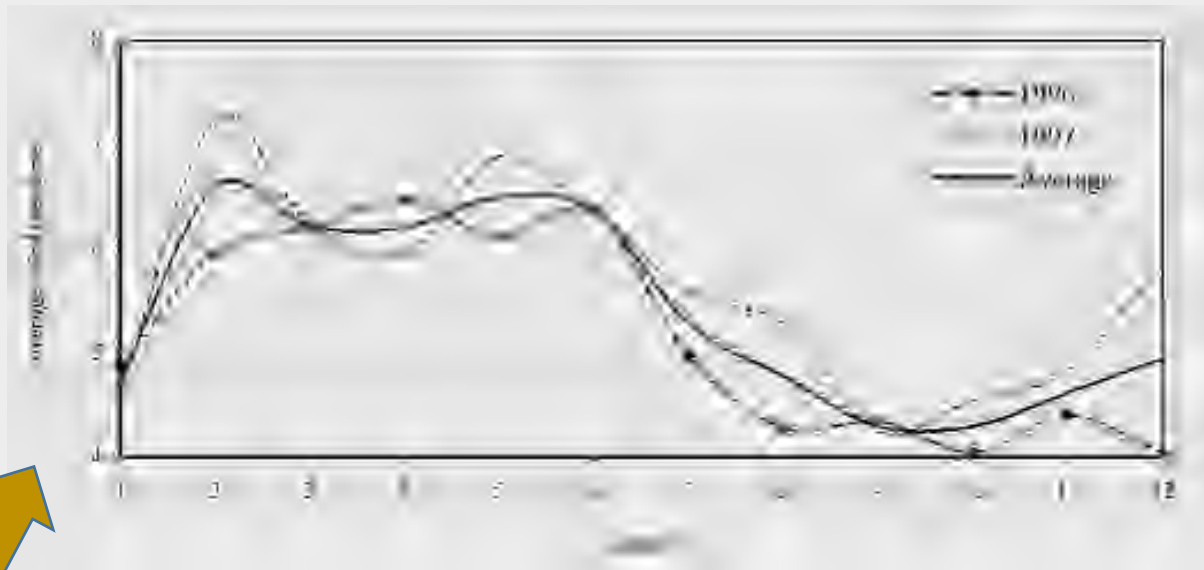
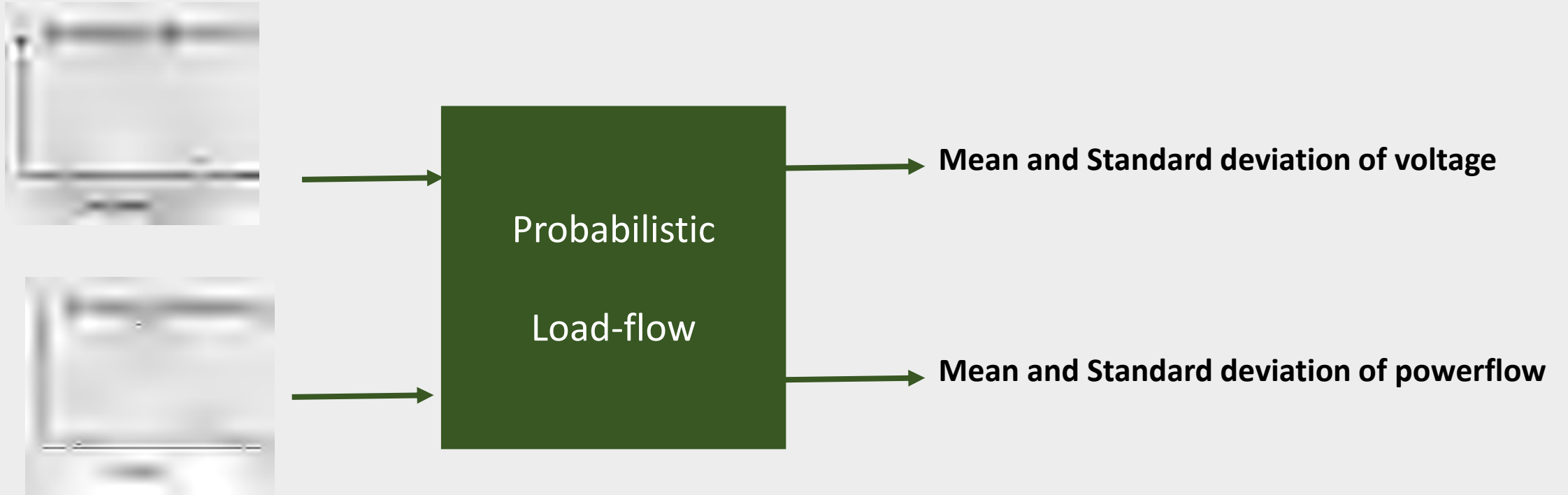


Fig. Wind velocity [Weisser et al. (2002)]

**POWER SYSTEM
UNCERTAINTIES**

PROBABILISTIC LOADFLOW



- ✓ Uses **Monte Carlo Simulation** or some analytical methods such as **cumulant, point estimate** etc

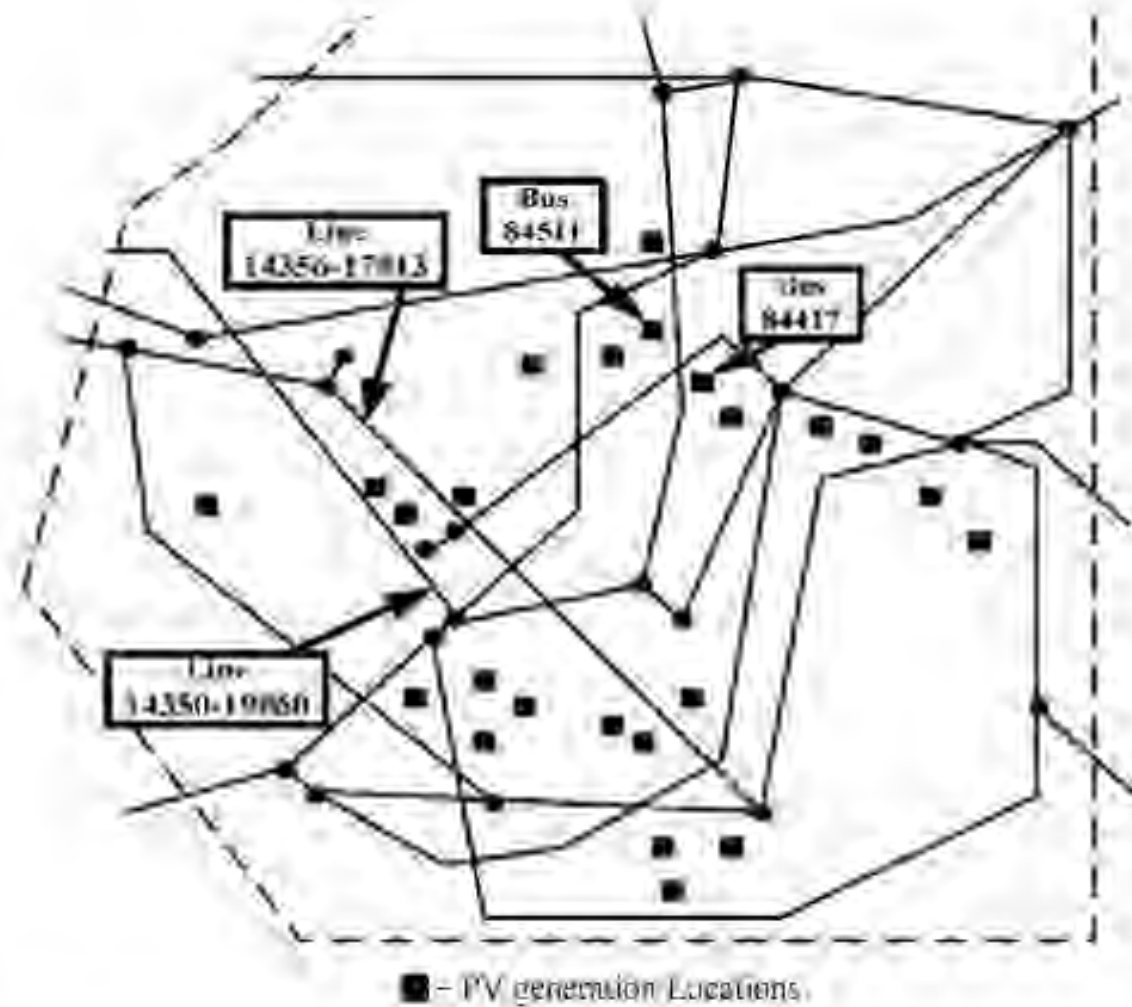


Fig. 6. Simplified portion of the WECC in Arizona.

PVG supplies 20% of the load demand

INPUTS

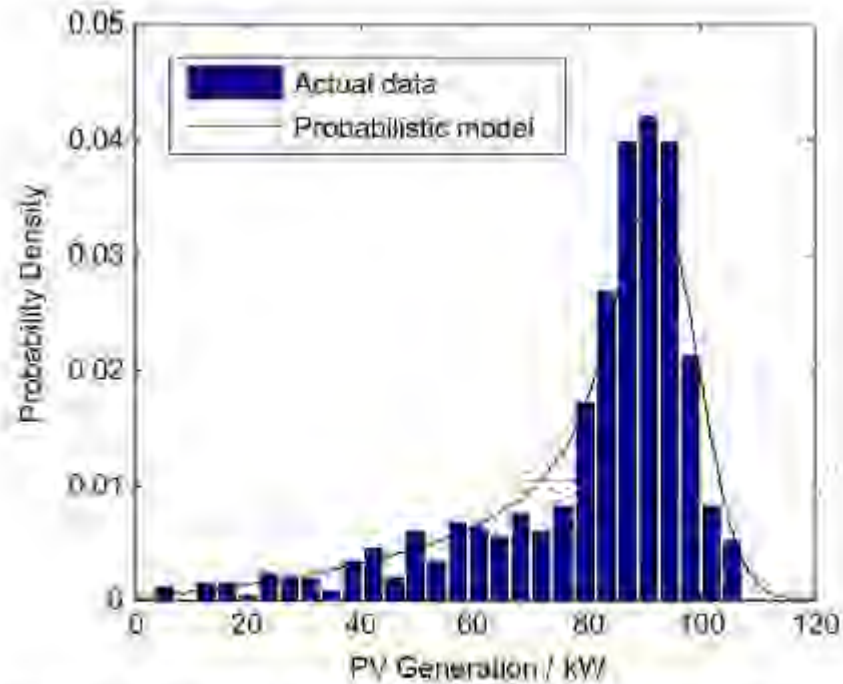


Fig 4 PDF curve of PV generation for both probabilistic model and actual data

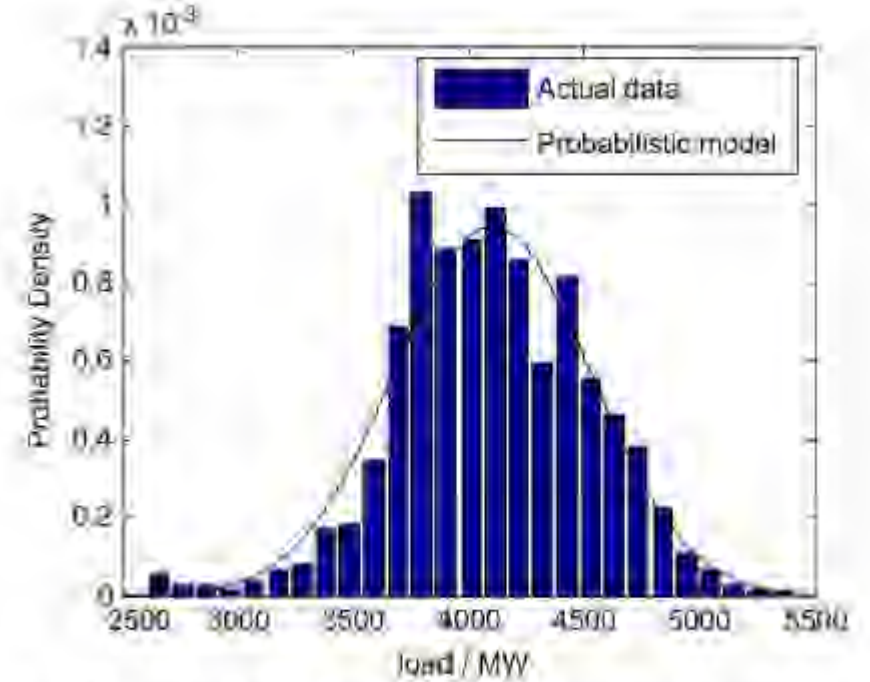


Fig 5 PDF curve of load for both probabilistic model and actual data

TABLE II

COMPARISON OF THE RESULTS FOR THE VOLTAGE MAGNITUDE AT BUS 44117

	MCS	
ARMS	1.0	
10% CL (pu)	1.0055	
90% CL (pu)	1.1506	
OVP (1.1 pu)	83.23%	
Time (s)	2147.20	

CL - confidence level; OVP - steady state overvoltage probability; GC - Gram-Charlier; BW - Edgeworth; CF - Cornish-Fisher.

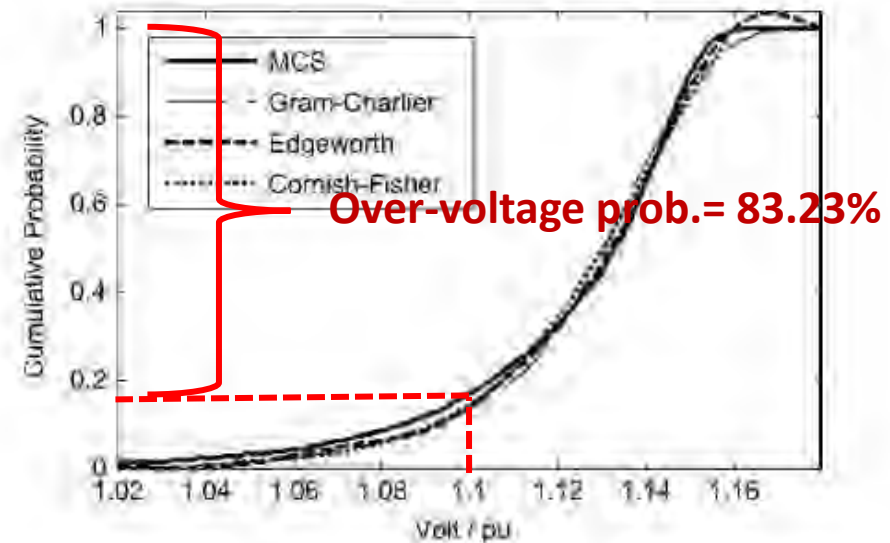


Fig. 7. CDF curves of the voltage magnitude at bus 44117 in three different types of expansions.

TABLE III

COMPARISON OF THE RESULTS FOR THE LINE
Flow Through Line 44556-17013

	MCS	
ARMS	1.0	
10% CL (pu)	1.0071	
90% CL (pu)	1.2877	
OVL (1.2 pu)	92.06%	

OVL - overload probability

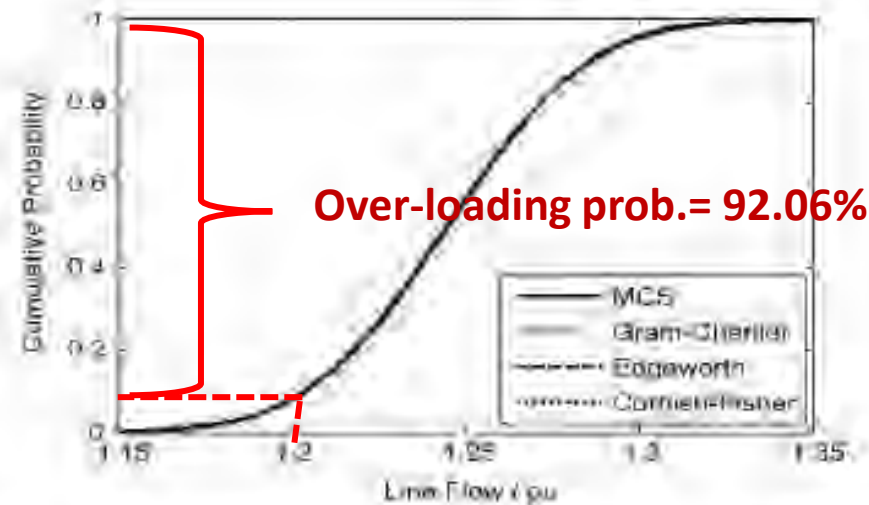


Fig. 8. CDF curves of the line flow through line 44556-17013 in three different types of expansions.

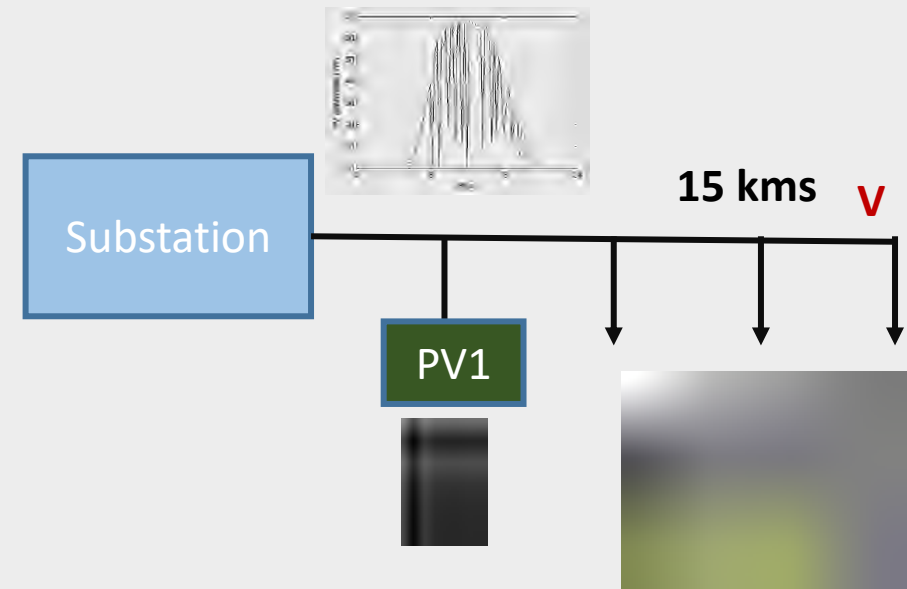
Other applications

□ Probabilistic optimization

- Task: Put a capacitor bank of appropriate size to keep the voltage near desired voltage (1 pu)
- System has a lot of photovoltaic generation

■ Solution:

- *Minimize* $\sum_{i=1}^{N_{bus}} P(V_i < 0.95)$
- Constraints: $Q_{min} \leq Q_{capacitor} \leq Q_{max}$



- ✓ This problem can be solved by different optimization algorithms such as **particle-swarm optimization, genetic algorithm** etc.

- Volt-Var Control (Distribution Automation)

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Review article

State-of-the-art technologies for volt-var control to support the penetration of renewable energy into the smart distribution grids

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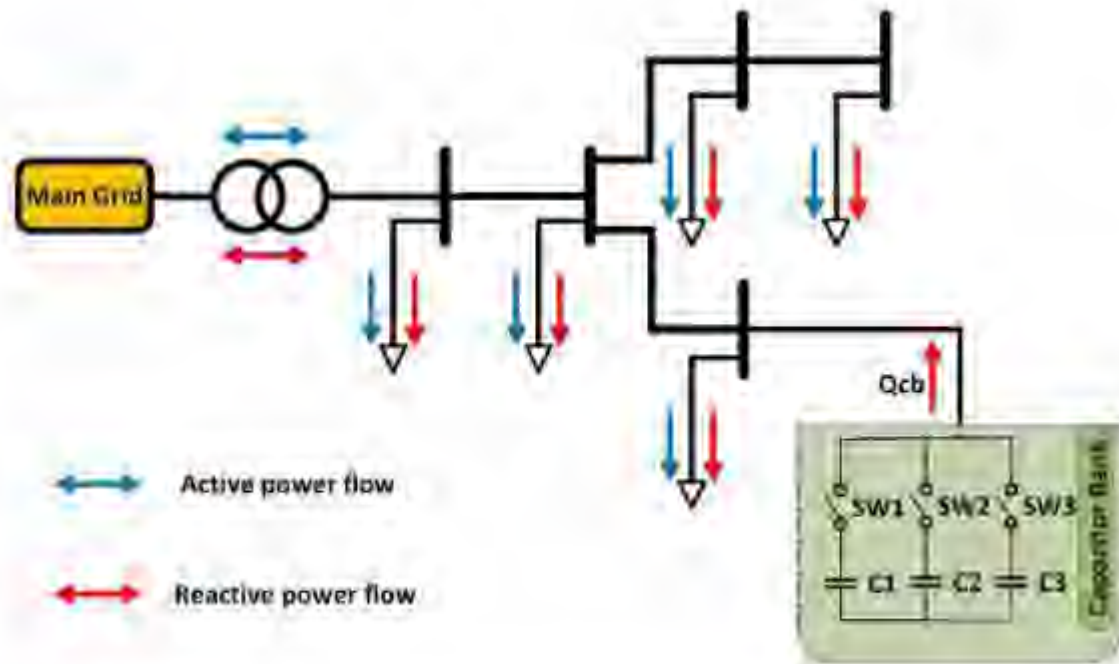


Fig. 3. Capacitor bank in a distribution network for reactive power compensation.

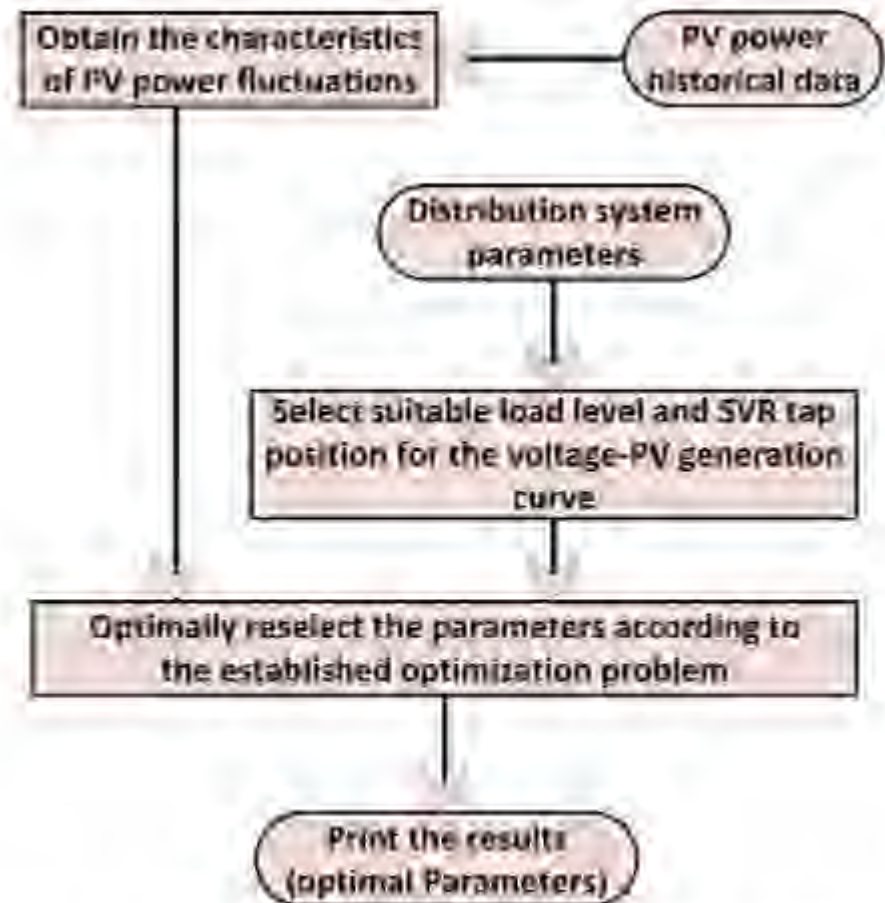


Fig. 5. The process of managing step voltage regulators and the PVs reactive power.

OTHER RESEARCH AREAS

- ✓ Machine learning in power systems
- ✓ Home Energy Management Systems
- ✓ Electric Vehicles (Vehicle 2 Grid)...



Thank you for your attention



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MACHINE LEARNING FOR POWER SYSTEMS

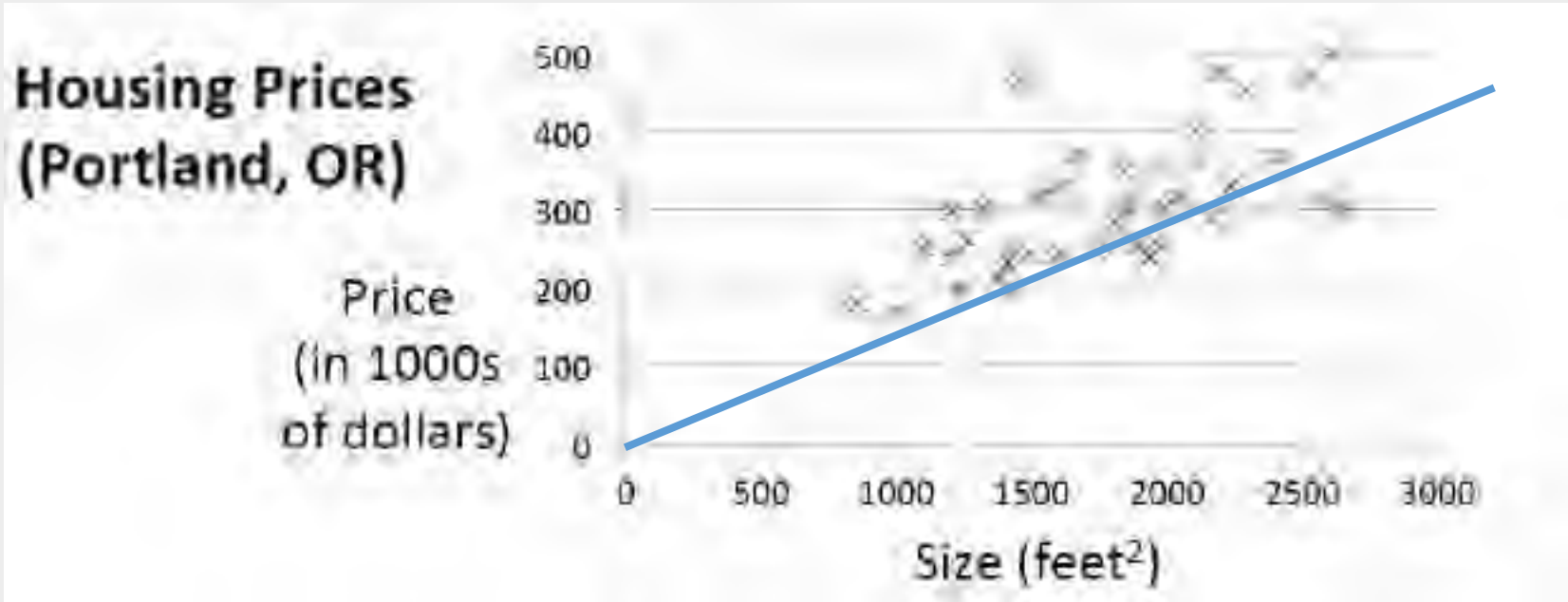
✓ **Artificial Intelligence** is a bigger concept of **machines** that can simulate human thinking behavior

✓ **Machine learning** is an application or subfield that allows **machines** to learn from data without being explicitly programmed

✓ Machine learning algorithms are categorized into **Supervised and Unsupervised**



Regression



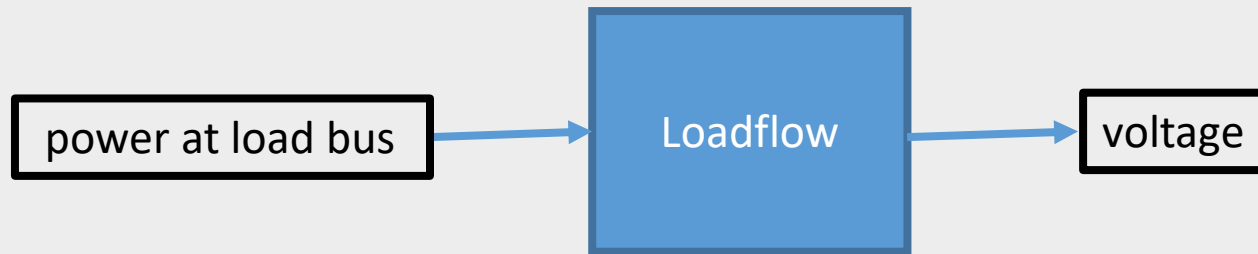
X	Y
1000	200
2500	300

Regression Problem

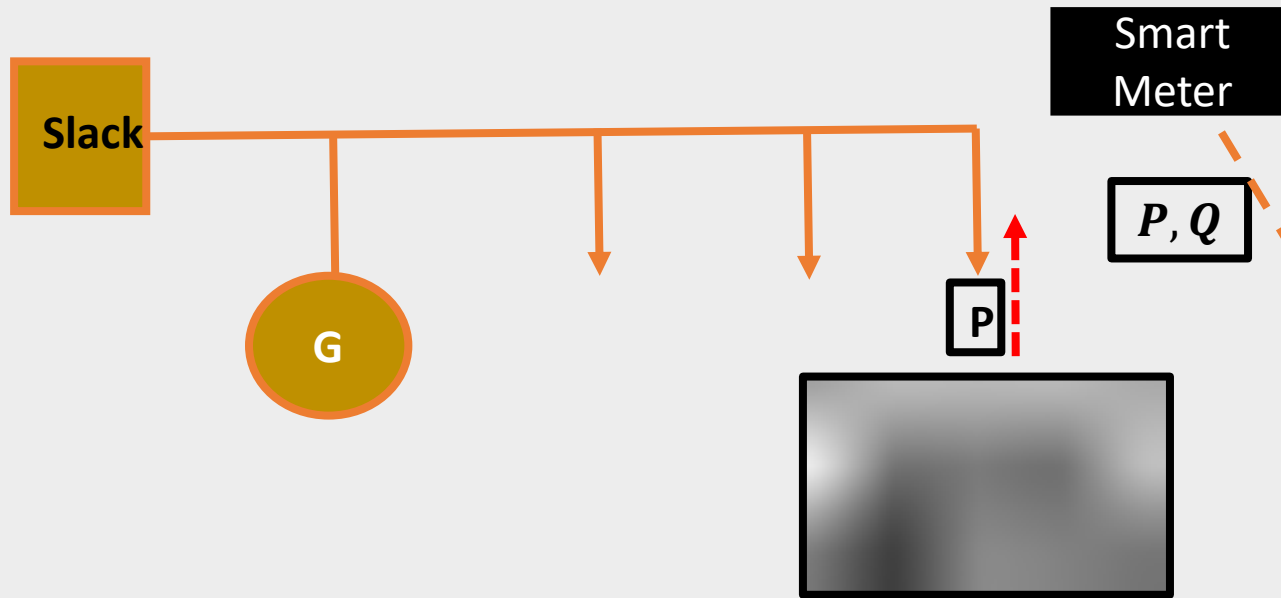
Predict real-valued output



CONVENTIONAL CPF



- Takes large **computational time** for large system
- **Security assessment** is done every 5 to 15 minutes



Real-time implementation



Prepare the model offline

$$\text{Voltage} = 0.5 * P + 0.8 * Q$$

Calculate the voltage (output)

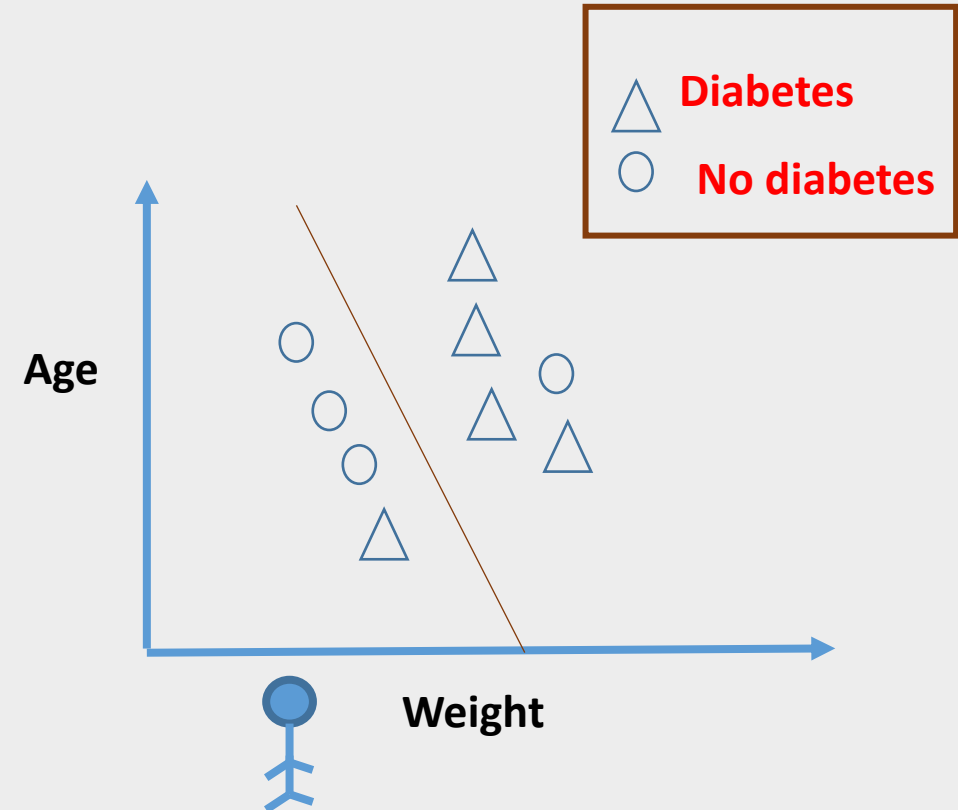
$$\text{Voltage} = f(P, Q)$$

Active Power (P)	Reactive Power (Q)	Voltage
5MW	1MVAR	1.02 pu

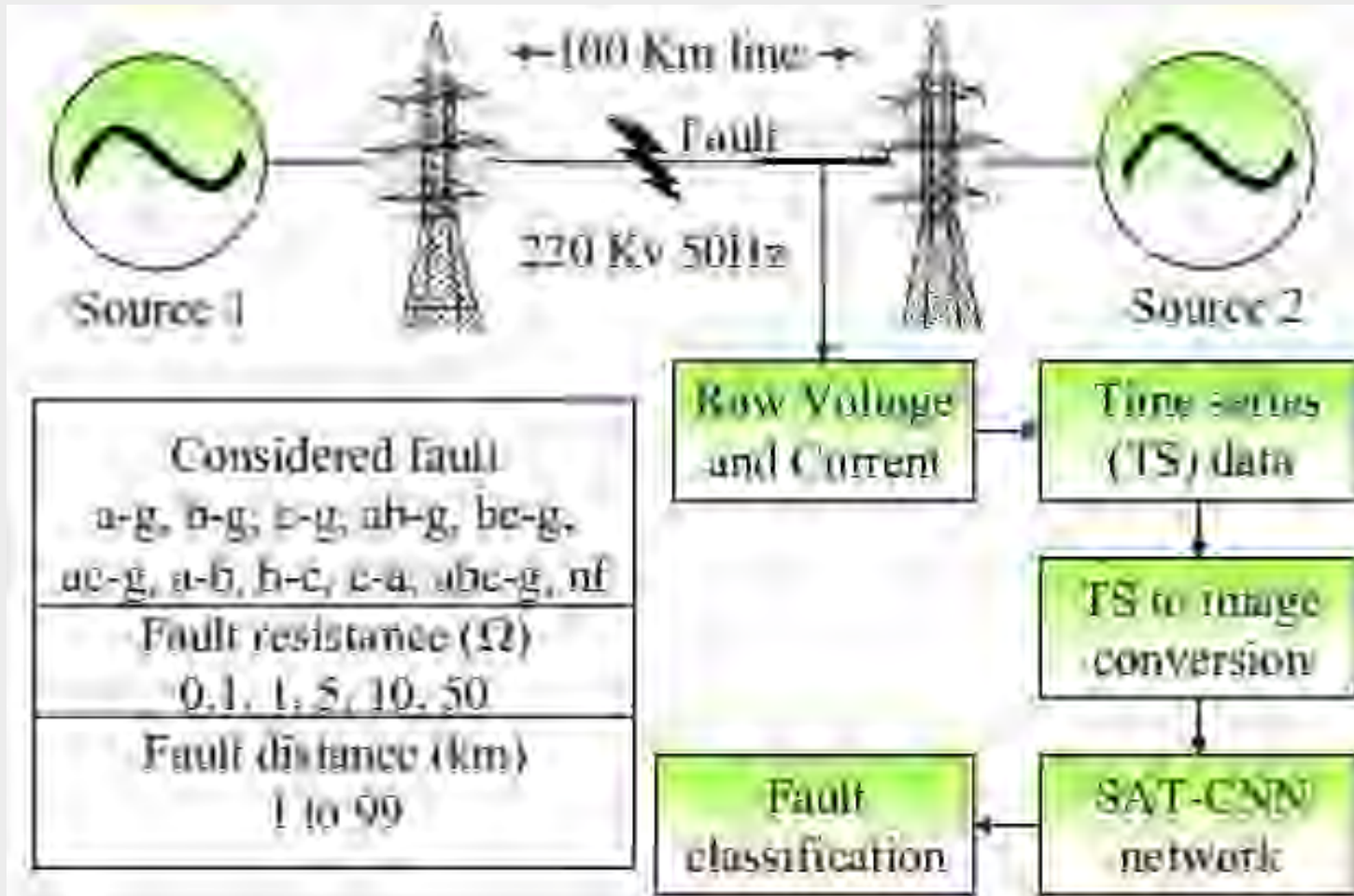
“Data-Driven Method”

Classification: Supervised learning

Age	Weight	Diabetes
30	90	1 (Yes)
50	60	0 (No)
.	.	.



Classification: Line fault classification



Lab model

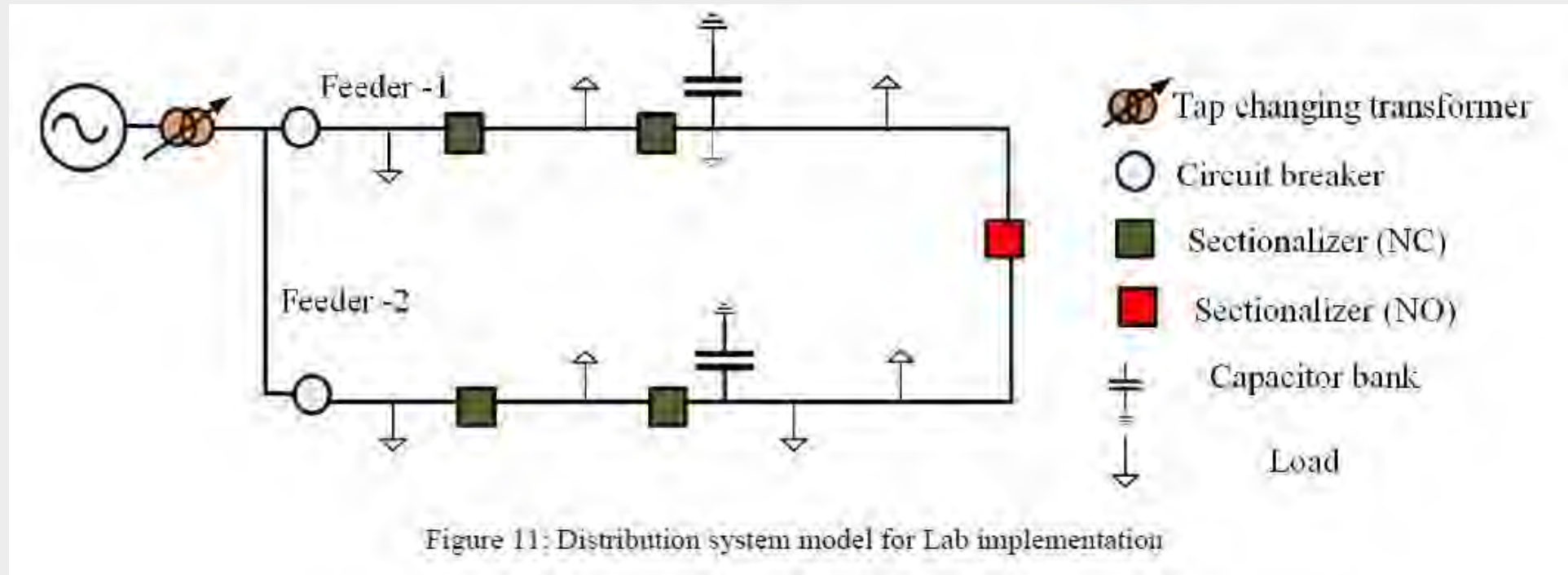


Table 2: Equivalent distribution components for DA lab implementation

Components	Equivalent
Supply (11kV)	12V (AC)
Feeder	Resistor and inductor
Sectionalizer	Contact
Load	Resistive bank



Figure 42: Monitoring the system status from Web app.



Figure 43: Overall hardware

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Yubraj, Aship and Abhijeet, KU