Multi-Agent Based Operation and Control of Electric Power Systems

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Intelligent Systems Application to Power Systems

- 1. Expert System (Rule Base System)
- 2. Fuzzy Inference & Fuzzy Reasoning
- 3. Artificial Neural Network
- 4. Heuristic Approach (Genetic Algorithm, Tabu Search, SA)
- 5. Multi-Agent System (Intelligent Agent)

International Conferences

ESAP(Expert System Application to Power Systems)

ANNPS(Application of Neural Network to Power Systems)

Now, ISAP(Intelligent Systems Application to Power Systems)

Research Topics

- 1. Real Time Wide Area Stability Monitoring System
- 2. Real Time Stability Margin Control of Electric Power Systems
- 3. Operation, Control and Management of Dispersed Power Sources including Renewable Energy Power Sources and Energy Storage Device
- 4. Artificial Neural Network Based Identification of Fault Location
- 5. Application of Energy Capacitor System to Power System Control
- 6. Multi-Agent Based Wide Area Operation, Control and Management of Electric Power Systems
- 7. Multi-Agent Based Hierarchical Stabilization Control of Power Systems
- 8. Multi-Agent Based AGC for Isolated Power Systems including Renewable Energy Power Sources and Energy Storage Device
- 9. Rule Based Voltage and Power Flow Management
- 10. Remote Tuning of Power System Controllers through Computer Network
- 11. Artificial Neural Network Based Estimation of Power Demand and Electricity
 Cost
- 12. Artificial Neural Network Based Diagnosis of Induction Machines
- 13. Development of Real Time PV System Simulator and MPPT Control under Partially Shaded Condition

Facilities for Experimental Studies

1. 5kVA Laboratory One Machine Power System:

5kVA Synchronous Generator 7kW DC Motor

Transmission Line Modules

Load Modules

2. 70Wh New Energy Storage Device(ECS):

Electrical Double Layer Capacitors

Maximum Charging/Discharging Power: 7kW

- 3. AC/DC Conversion Unit for Real Power Control
- 4. AC/DC Conversion Unit:

Active/Reactive Power Control

- 5. Wind Turbine Generators: 600W
- 6. PV System: 400W
- 7. VPN: Virtual Private Network

Analog Power System Simulator at the Research Laboratory of Kyushu Electric Power Co. for Joint Research Projects: 10 to 14 weeks a year

Experimental Facilities (1)

MG Set



ECS and AC/DC Conversion Unit



ECS: Energy Capacitor System Electrical Double-Layer Capacitors

We have several AC/DC conversion units for PV and WTG system and also for different control purposes.

Experimental Facilities (2)



Small Sized Wind Turbine Generating Units



Photo-Voltaic Generating Units

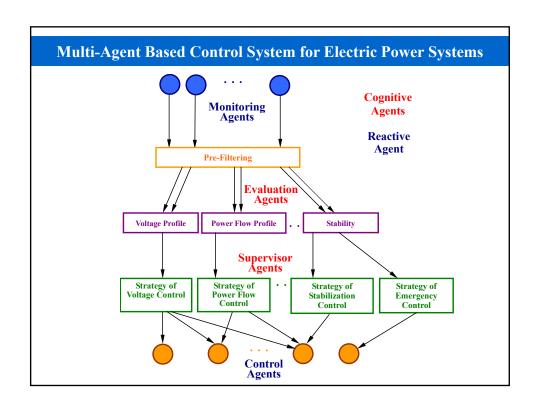
Analog Power System Simulator at the Research Laboratory of Kyushu Electric Power Co. (10 to 15 weeks per year)



Preparation of Simulator Test

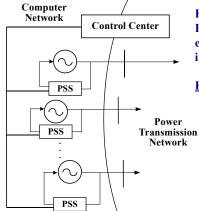
Overview of Analog Power System Simulator





Intelligent Agent Based Remote Tuning of Power System Stabilizer through Computer Network

Configuration of Intelligent Agent Based Remote Tuning System



Power Station PSS Site For the remote tuning of each PSS, the PSS standard tests should be activated and evaluated regularly at the PSS sites by the intelligent agent from the control center.

Evaluation of Control Performance:

$$J = \sum \Delta T^2 \Delta P_e^2$$

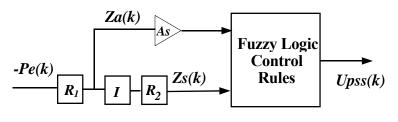
△Pe: generator real power deviation

 ΔT : sampling interval of the PSS.

Here, it must be noted that the proposed remote tuning system utilizes the virtual private network(VPN) among different network groups and also for the security reasons.

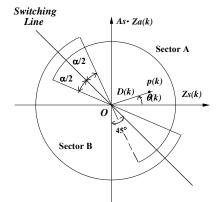
Basic Configuration of Fuzzy Logic PSS

Pre-Filtering for Acceleration and Speed Deviation Signals

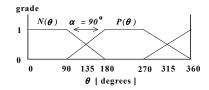


R: Reset Filter I: Integrator **Upss:** Switching Control Signal

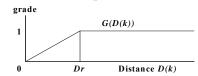
Membership Functions and Control Signal



Angle Membership Functions



Radius Membership Function

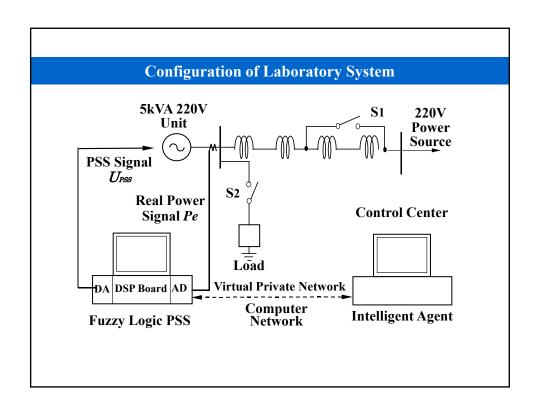


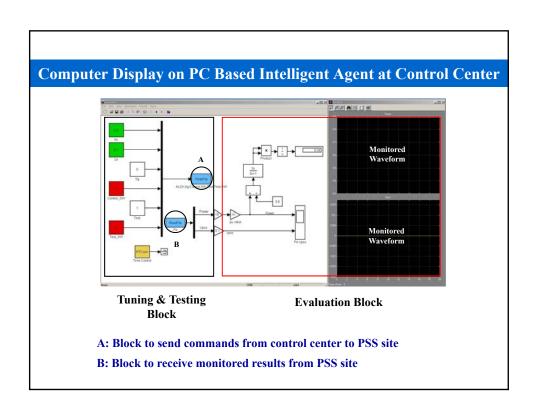
Speed/Acceleration Phase Plane

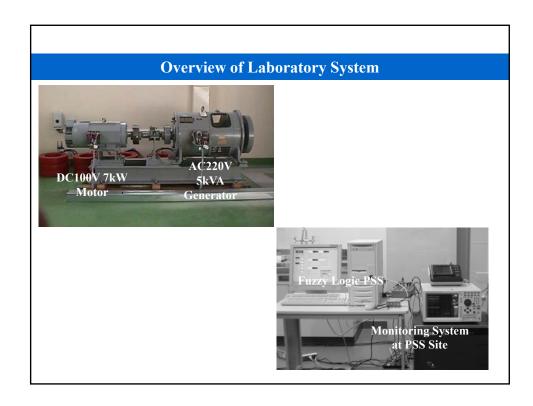
$$D(k) = \sqrt{Z_{s}(k)^{2} + (A_{s} \cdot Z_{a}(k))^{2}}$$

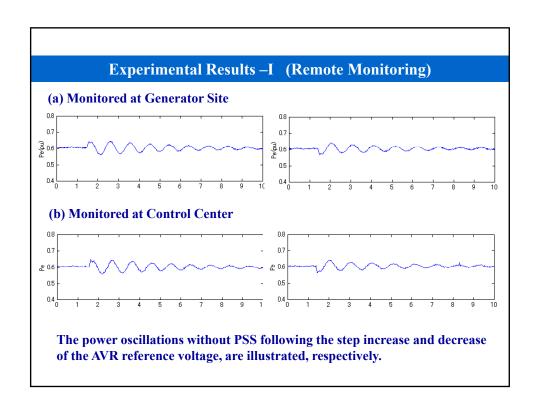
$$\theta(k) = \tan^{-1}(A_s \cdot Z_a(k) / Z_s(k))$$

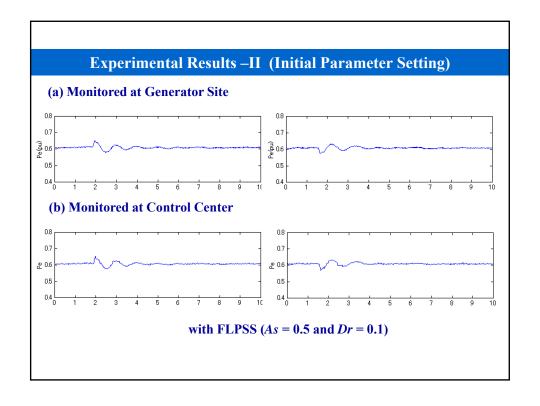
$D(k) = \sqrt{Z_{s}(k)^{2} + (A_{s} \cdot Z_{a}(k))^{2}} \qquad Upss(k) = \frac{N(\theta(k)) - P(\theta(k))}{N(\theta(k)) + P(\theta(k))} \cdot G(D(k)) \cdot U_{max}$

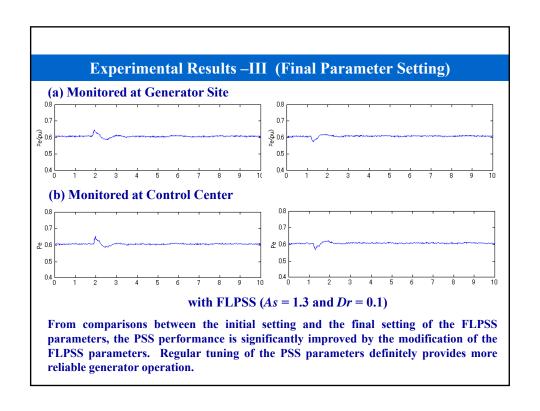


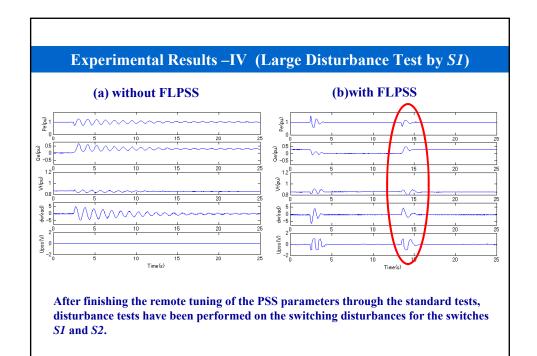


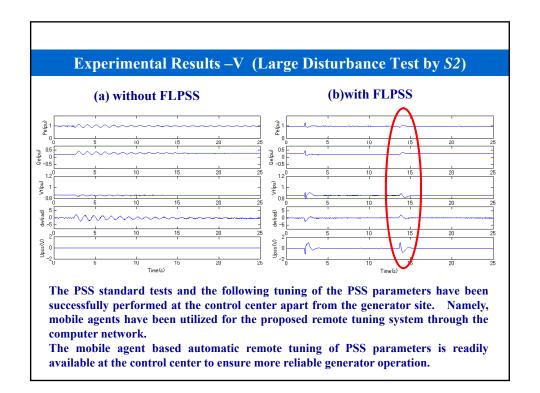




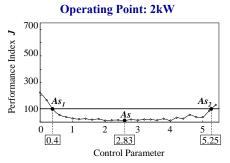


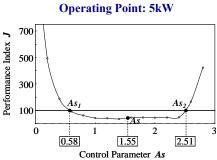






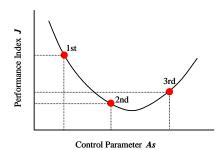
Experimental Results: Detailed Tuning of PSS Parameters at Site





- 1. The performance index *J* has a quite wide bottom within a certain range of the parameter *As*. Therefore, the robustness of the fuzzy logic PSS is clearly recognized from the results.
- 2. At the middle points of the bottom, the parameter As are found as 2.83 and 1.55 for the output setting of 2kW and 5kW, respectively. These two values might be considered as the optimal setting of As at those two different generator operating points.

Experimental Results: Intelligent Agent Based Remote Tuning



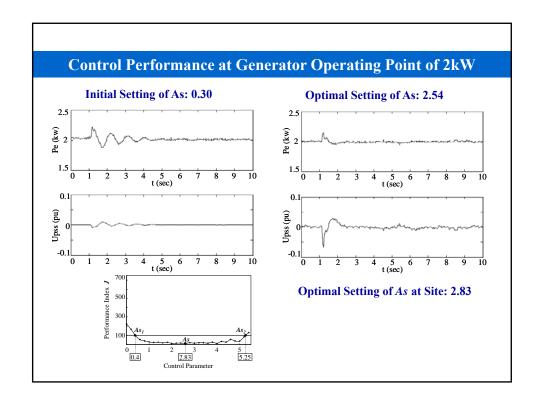
For the optimization of the parameter As, the standard tests have been activated by the intelligent agent from the control center.

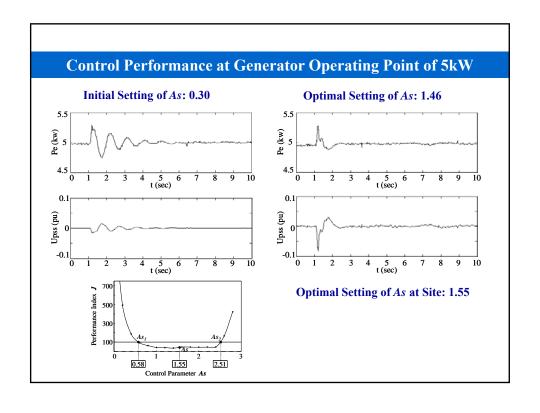
To determine the optimal value for the parameter As, the quadratic approximation has been utilized as shown in the figure.

No. of Standard Test	As	Index J (x 100)
1	0.3	151
2	0.35	137.2
3	0.45	105.9
4	0.65	68.3
5	1.05	38.6
6	1.85	20.1
7	3.45	23.9
Optimal Setting	2.54	19.7

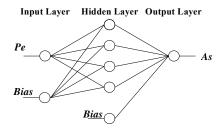
(a) Generator operating point: 2kW

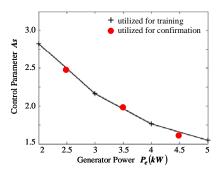
(b) Generator operating point: 5 kW		
No. of Standard Test	As	Index J (x100)
1	0.3	315.6
2	0.35	276.1
3	0.45	172.4
4	0.65	110.7
5	1.05	31
6	1.85	26.2
7	3.45	850.5
Optimal Setting	1.46	25.3











The training of the neural network is performed by the intelligent agent at the control center after having enough data for the training.

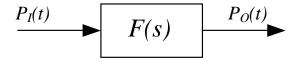
Whenever the retraining is required, the retrained result is transferred to the corresponding adaptive fuzzy logic PSS for its renewal.

Eigenvalue-based Wide Area Stability Monitoring of Power Systems

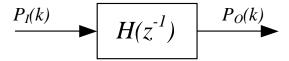
Journal of Control Engineering Practice 13 (2005) 1515-1523

Real Time Stability Evaluation - 1

(a) Continuous Time Transfer Function



(b) Discrete Time Transfer Function



Through the monitoring of the input signal PI(t), and the output signal PO(t), the parameters of the target system can be easily identified. Therefore, the stability of the target system can be evaluated from the eigenvalues of the identified transfer function.

Real Time Stability Evaluation - 2

After monitoring the output PO(t) for an input signal PI(t), the relation between the input PI(t) and the output PO(t) can be expressed in a discrete manner as follows:

$$P_O(k) = a_1 P_O(k-1) + a_2 P_O(k-2) + \dots + a_n P_O(k-n) + b_0 P_I(k) + b_1 P_I(k-1) + \dots + b_n P_I(k-n)$$

After identifying the above model parameters by using the least square method, the discrete time transfer function H(z-1) can be derived as follows:

$$H(z^{-1}) = \frac{b_0 + b_1 z^{-1} + \dots + b_n z^{-n}}{1 - a_1 z^{-1} - a_2 z^{-2} - \dots - a_n z^{-n}}$$

By solving the following characteristic equation, the stability of the discrete time system with the transfer function H(z-1) can be evaluated.

$$1 - a_1 z^{-1} - a_2 z^{-2} - \dots - a_n z^{-n} = 0$$

Real Time Stability Evaluation - 3

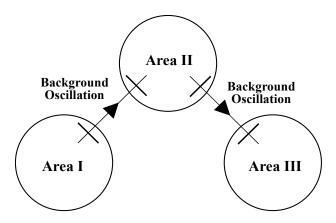
The discrete time eigenvalues can be easily converted to their corresponding continuous time eigenvalues as follow:

$$z_i = x_i + j y_i$$

$$\alpha_{i} = \frac{\ln\left(\sqrt{x_{i}^{2} + y_{i}^{2}}\right)}{T} \qquad \beta_{i} = \frac{\tan^{-1}\left(\frac{y_{i}}{x_{i}}\right)}{T}$$

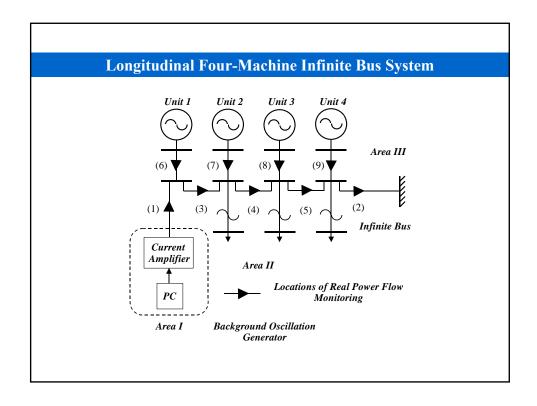
where α is the estimated damping coefficient and β gives the estimated frequency of the oscillation modes of the continuous study system F(s). Here, it must be noted that the system is stable when all the damping coefficients α have negative values. In addition, T denotes the sampling interval for the discrete time system.

Study System

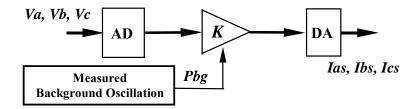


Background oscillations are measured at the locations between Area I and Area II, and also between Area II and Area III.

The stability of the study system is evaluated based on those measured background oscillations.

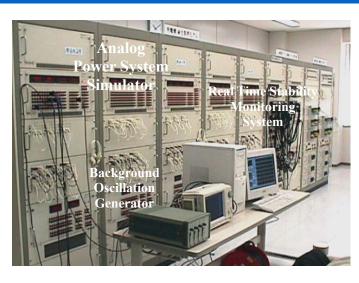


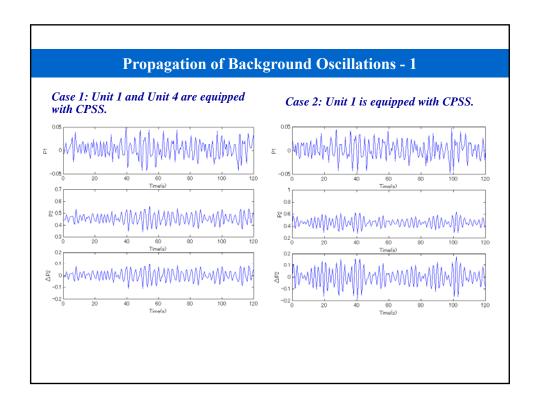
Basic Configuration of Background Oscillation Generator

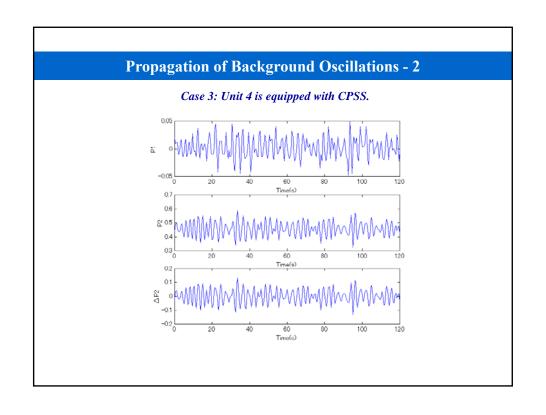


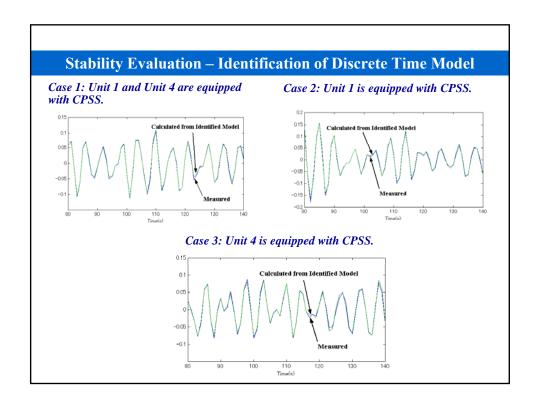
To generate background oscillations, real power flow actually measured on a 500kV trunk line were utilized.

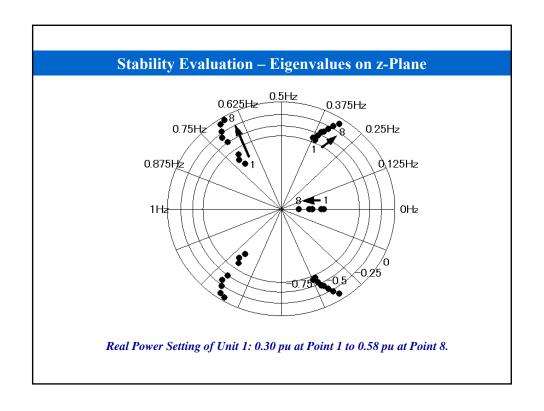
Overview of Analog Power System Simulator

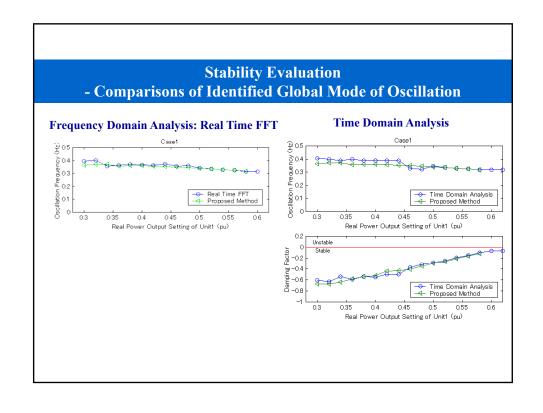


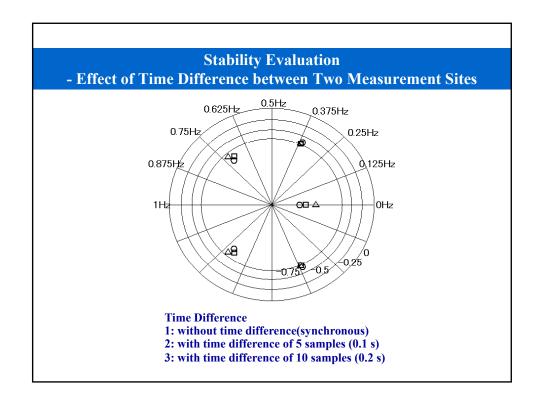


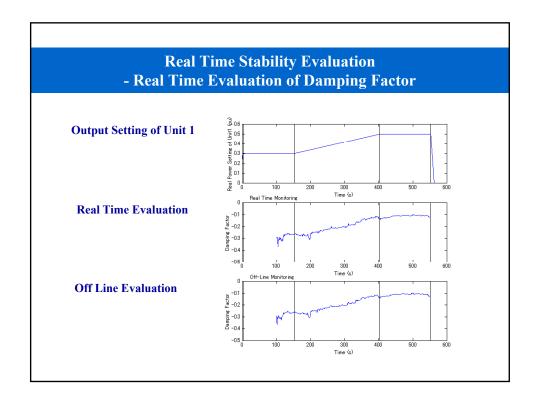




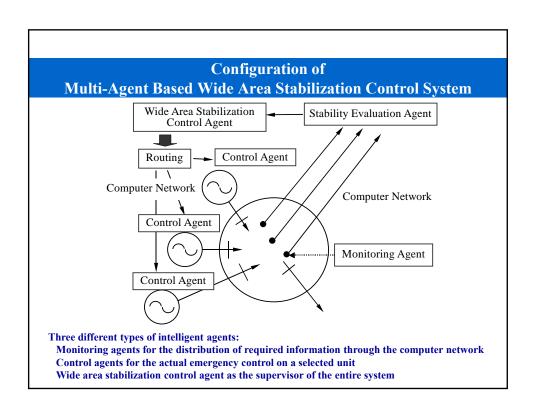








Multi-Agent Based Wide Area Stabilization Control of Electric Power Systems



Selection of Target Unit for Emergency Control

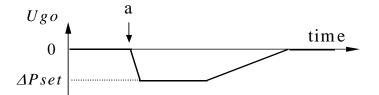
Following performance index has been utilized to evaluate the stability of each generating unit. The unit with the lowest stability level is selected as the target unit for the proposed emergency control.

$$J_i = \sum \Delta \omega_i^2$$

 $\Delta\omega_i$: speed deviation of the *i*-th unit

Emergency Control after Detecting Instability on Selected Unit - 1

The emergency controller on the selected unit shifts its operating point to its new stable equilibrium by modifying the real power output setting in the corresponding speed governing control system.



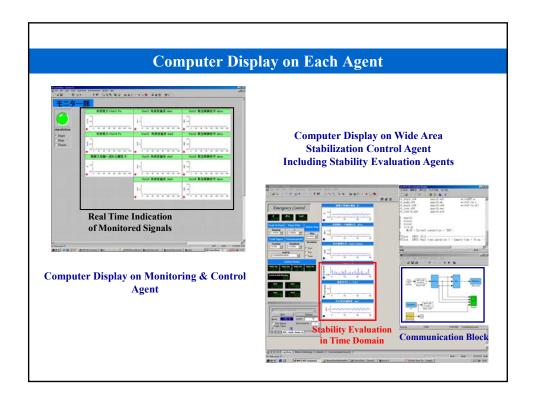
Insatbility is detected at t = a.

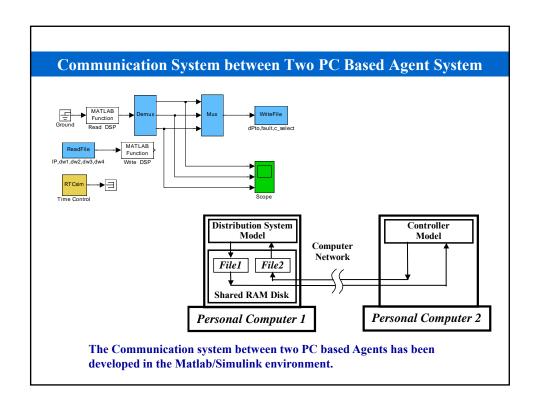
PC Based Wide Area Stabilization Control System for Analog Simulator Tests

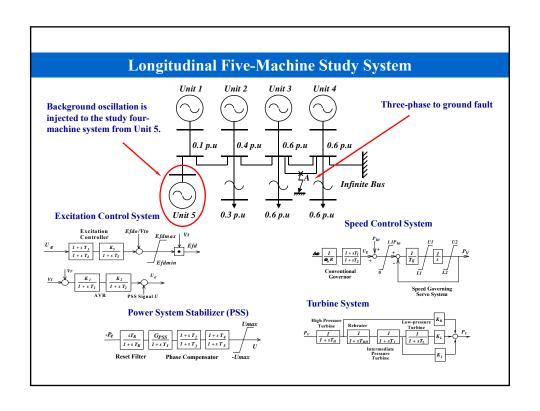


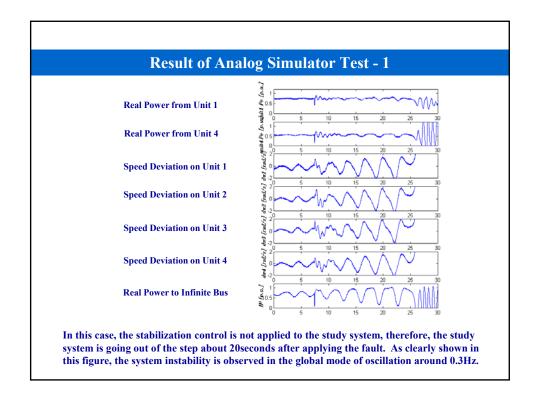
A DSP board with AD/DA conversion interfaces has been installed on the PC based Monitoring & Control Agent as the interfaces to and from the Analog Power System Simulator.

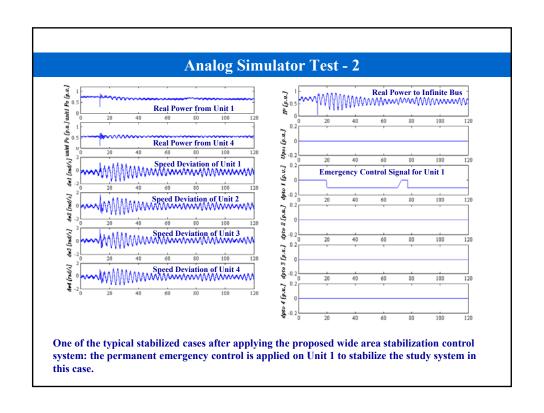
The computer network has been utilized as the interfaces between these two personal computers.











Multi-Agent Based Operation and Control of Distribution Systems with Dispersed Power Sources

- With Renewable Energy Power Sources & Energy Storage Device

Introduction

Renewable Energy Power Sources:

Photo-Voltaic Generation & Wind Turbine Generation

Conventional Power Sources:

Diesel Generation & Gas Turbine Generation

Energy Storage Devices:

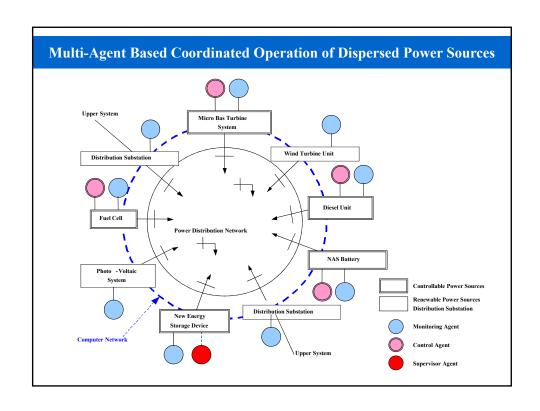
Energy Capacitor System (Electrical Double Layer Capacitors)

A number of new distributed power generation technologies are currently available to offer integrated performance and flexibility for the power consumers.









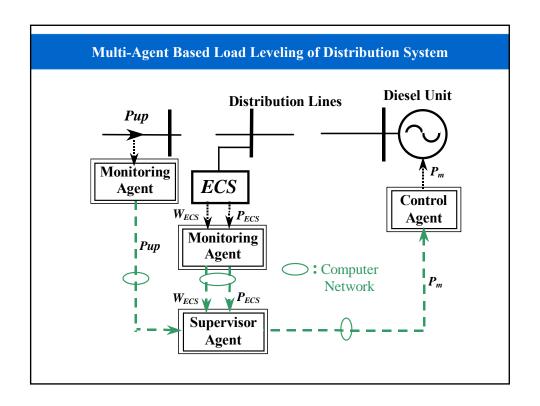
Multi-Agent System

Three Types of Agents:

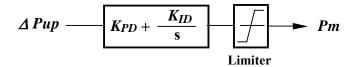
<u>Monitoring Agents</u> for the distribution of required information through the computer network (Reactive Agent)

<u>Control Agents</u> for the charging/discharging operation on the ECS and also for the power regulation on the diesel units (Reactive Agent)

<u>Supervisor Agent</u> for the coordination between the ECS and the diesel units (Cognitive Agent)



Conventional Load Leveling Control on Diesel Unit



△ Pup: Deviation of Power Flow from Upper System Pm: Output Setting of Diesel Unit

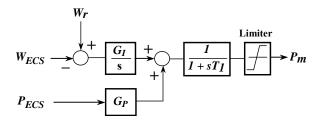
Proposed Load Leveling Control (Supervisor Agent)

$$\Delta P_{up}$$
 $K_P + \frac{K_I}{s}$ Limiter

△ Pup: Power Flow Deviation from Upper System PSECS: Output Setting of ECS

Coordination between ECS and Diesel Unit (Supervisor Agent)

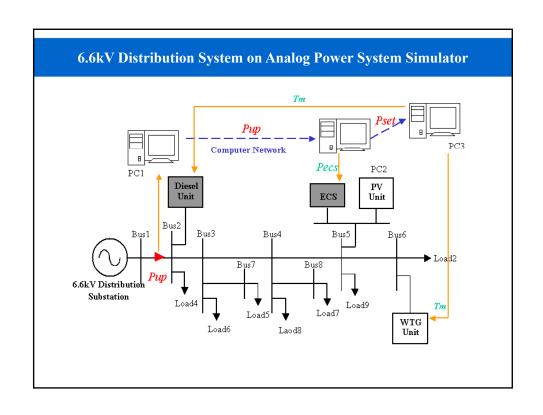
- 1. A small sized ECS is considered in this study, therefore, the regulation of the output from the diesel unit is required to keep the stored energy level of the ECS in a proper range.
- 2. The ECS provides the main function of load leveling control and the diesel unit provides its supplementary function to support the load leveling control on the ECS.

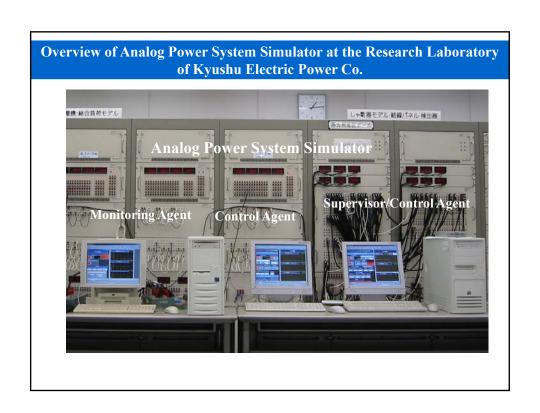


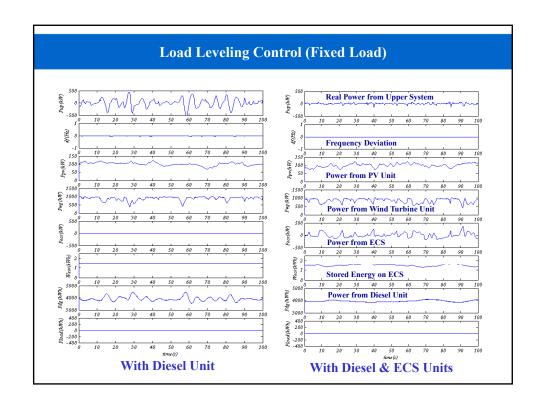
Wr: Target Stored Energy, WECS: Current Stored Energy

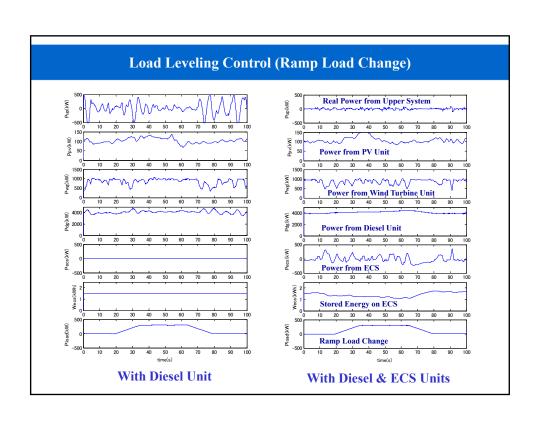
PECS: Power from ECS

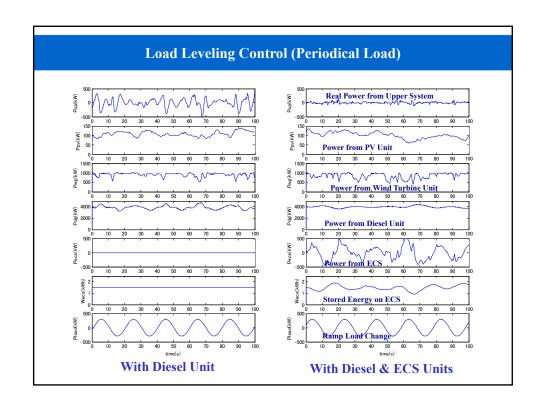
Pm: Power Regulation on Diesel Unit for Coordination with ECS

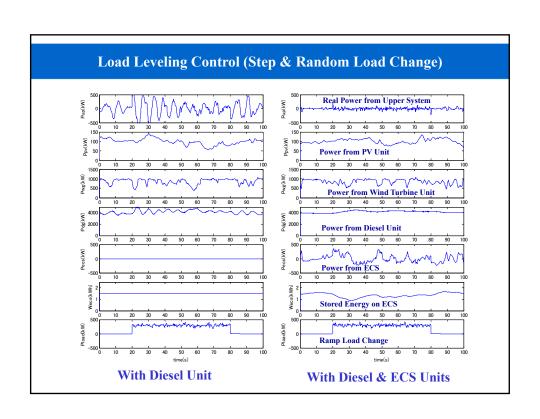


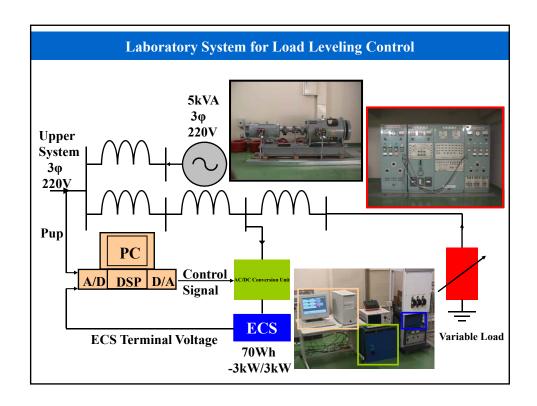












Conclusion

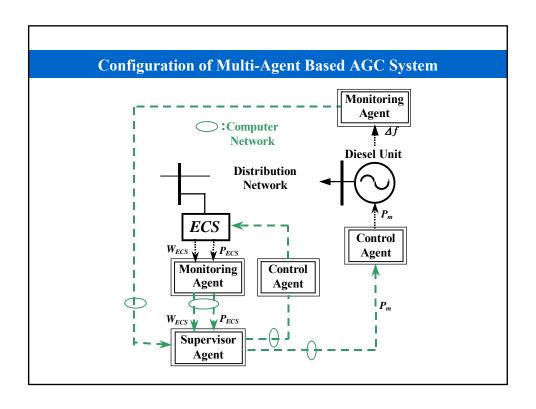
A multi-agent based load leveling control scheme has been proposed for distribution systems.

The real time simulation results, performed on the Analog Power System Simulator at the Research Laboratory of Kyushu Electric Power Co., clearly indicate the advantages of the proposed multiagent based load leveling control scheme even in the existence of the communication delay to a certain level.

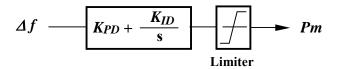
Further studies are now ongoing for the compensation of the long communication delay.

Experimental Studies on Multi-Agent Based AGC for Isolated Power System with Dispersed Power Sources

Journal of Engineering Intelligent Systems, Vol. 13, No. 2 (2005)

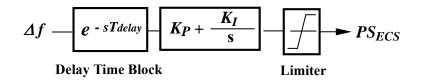


Conventional AGC on Diesel Unit



△f: Frequency Deviation on Diesel Unit **Pm**: Output Setting of Diesel Unit

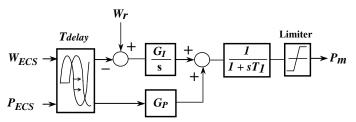
Proposed AGC (Supervisor Agent)



△f: Frequency Deviation of Diesel Unit PSECS: Output Setting of ECS Tdelay: Communication Delay Time

Coordination between ECS and Diesel Unit (Supervisor Agent)

- 1. A small sized ECS is considered in this study, therefore, the regulation of the output from the diesel unit is required to keep the stored energy level of the ECS in a proper range.
- 2. The ECS provides the main function of AGC and the diesel unit provides a supplementary function to support the AGC on the ECS.

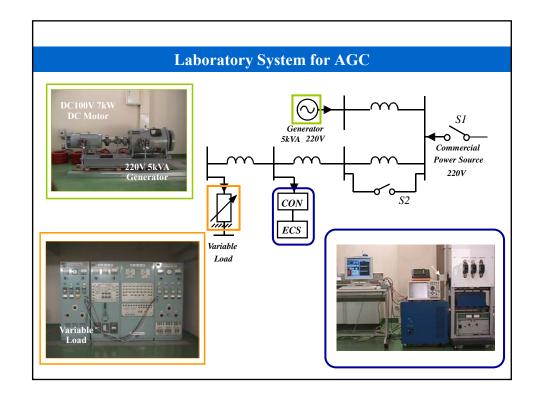


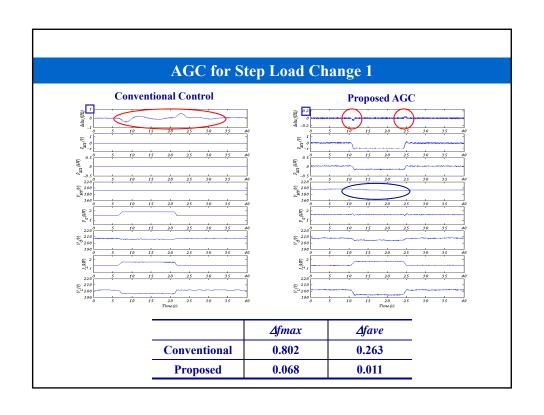
Wr: Target Stored Energy, WECS: Current Stored Energy

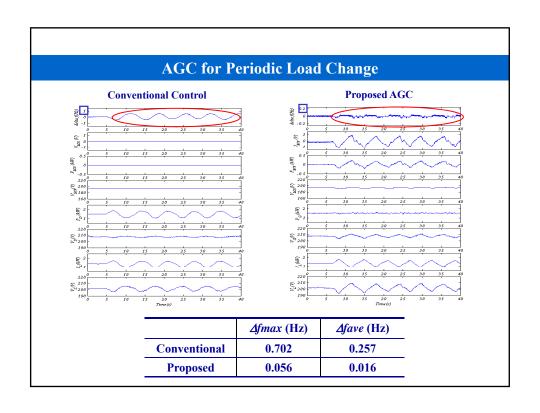
PECS: Power from ECS

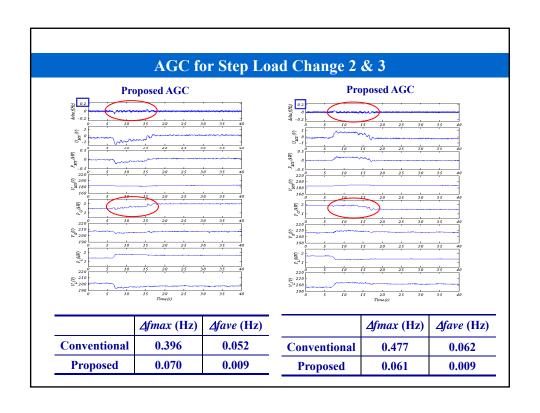
Pm2: Power Regulation on Diesel Unit for Coordination with ECS

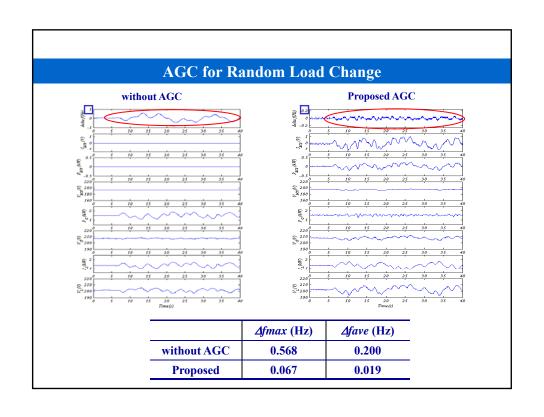
Tdelay: Communication Delay Time

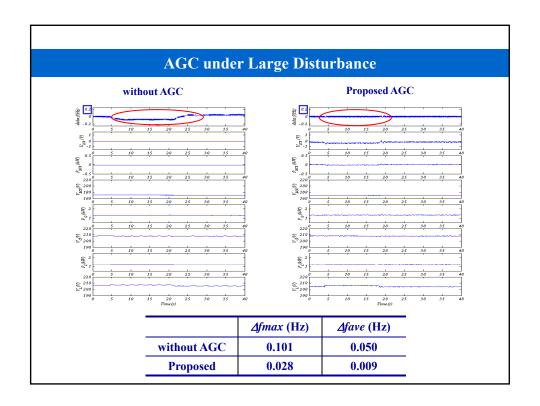










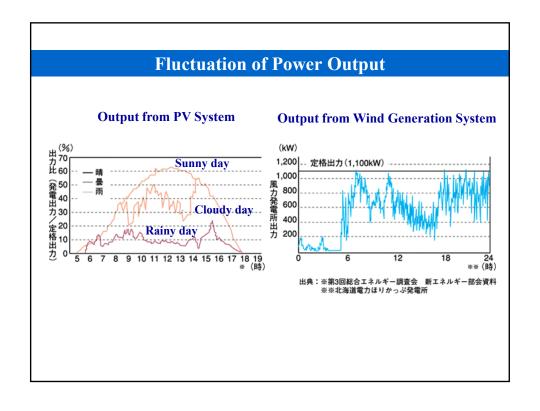


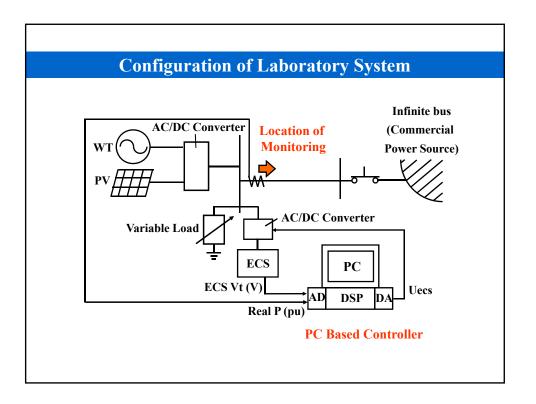
Conclusion

A multi-agent based AGC scheme has been proposed for isolated power systems. The experimental results, performed on the laboratory system, clearly indicate the advantages of the proposed multi-agent based AGC scheme even in the existence of the communication delay to a certain level.

Further studies are now ongoing for the development of hierarchical control where the multi-agent based control system performs its upper level control and the local control system performs its lower level control.

Operation and Control for Hybrid Power Source Including Renewable Power Sources - Rule-based Regulation & PI-type Regulation





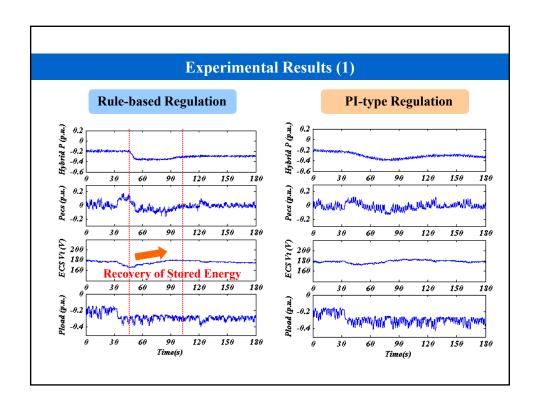
Estimation of Stored Energy

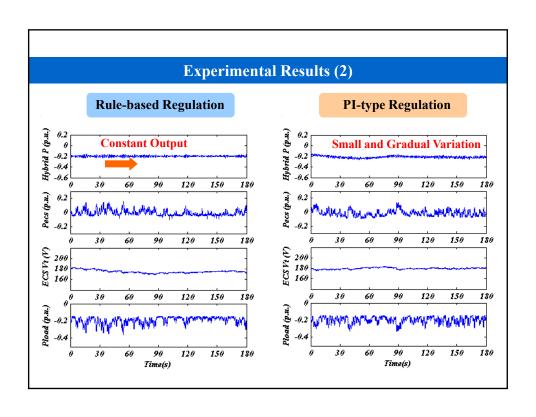
Stored Energy in Wh

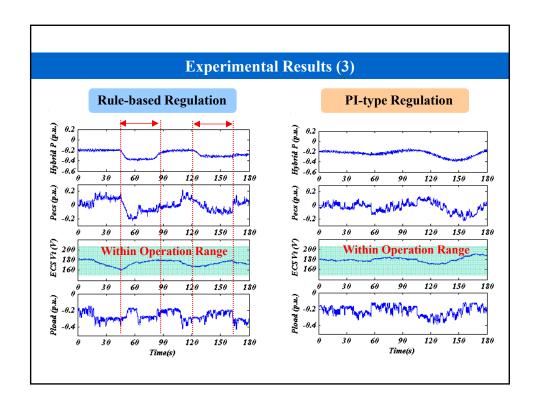
$$Wh = \frac{1}{2}CV^2 * \frac{1}{3600}$$

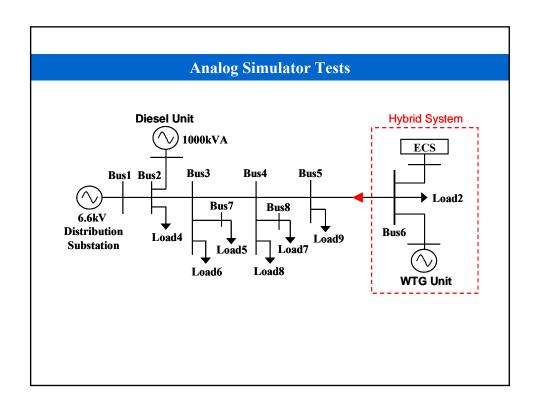
C: Capacitance (F)

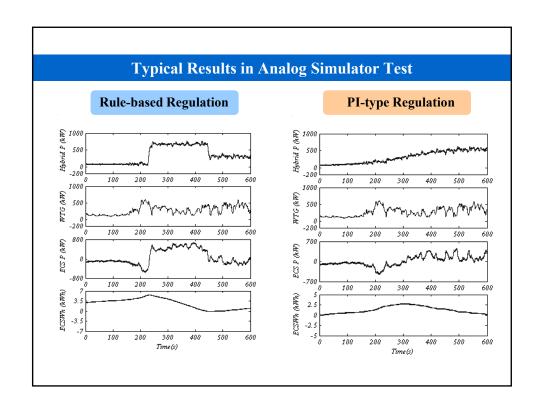
 $V: \; {
m DC} \; {
m side} \; {
m terminal} \; {
m voltage} \; \; {
m Monitoring} \; {
m of} \; {
m DC} \; {
m side} \; {
m voltage}$

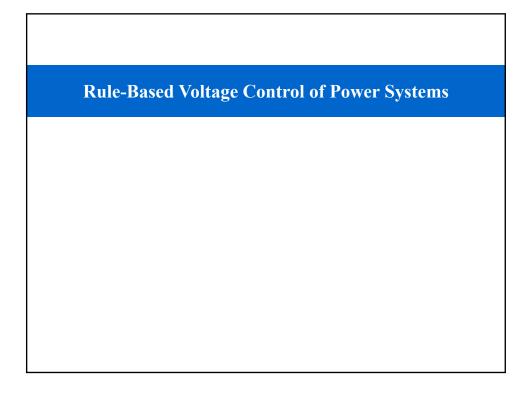


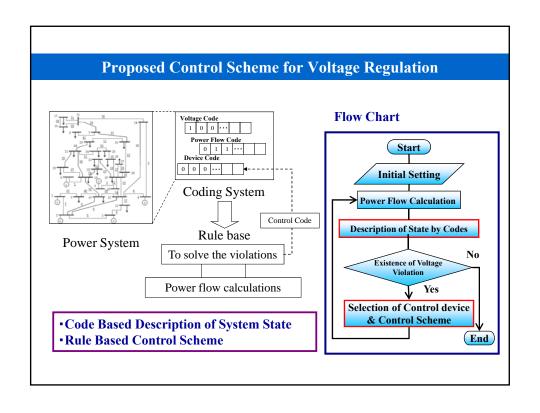


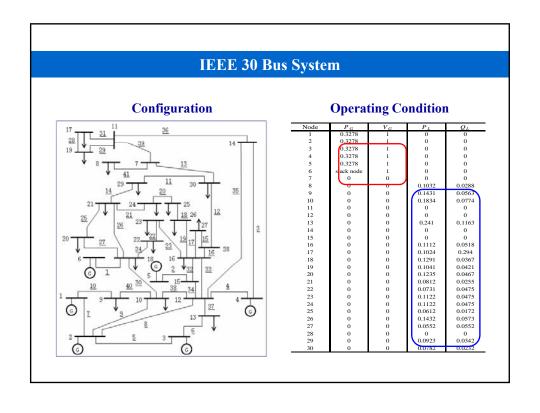


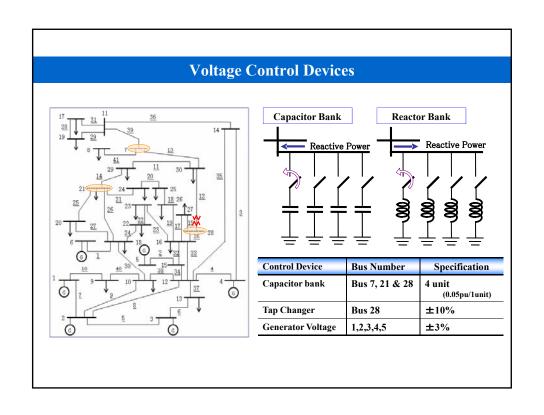


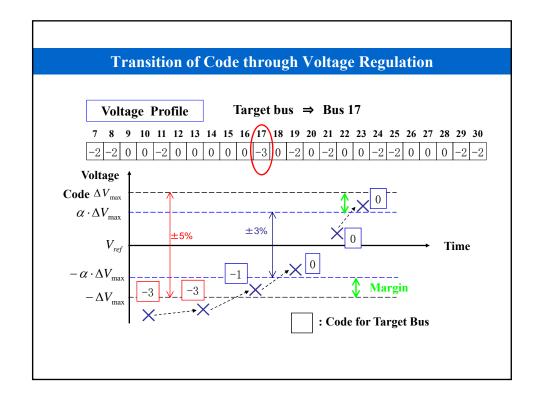












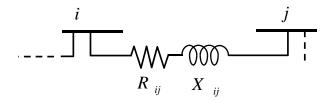
Selection of Control device - Measure of Electrical Distance

The controlled device is selected by using the distance measure.

The distance Lij of the line between Bus i and Bus j is defined as follows:

$$L_{ij} = \sqrt{R_{ij}^2 + X_{ij}^2}$$

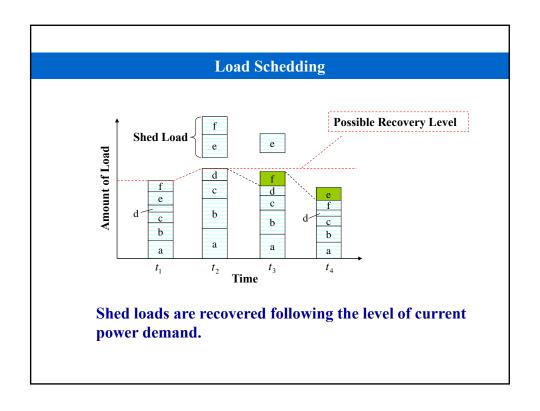
where the terms Rij and Xij give the resistance and the reactance of the line between Bus i and Bus j. The minimum distance for any pair of Bus i and Bus j can be calculated by using the Dijkstra Method.

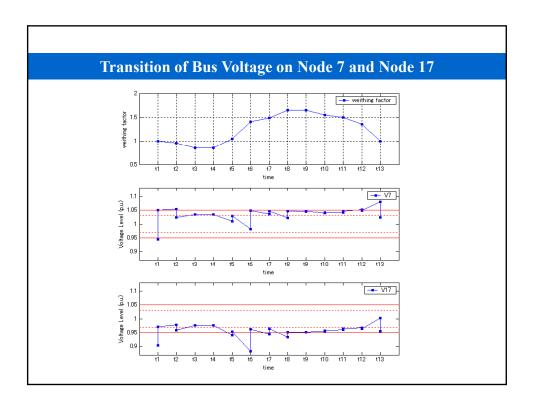


Process to Solve Voltage Violation

Violated Node	Ranking of Nearest Voltage Control Distances
17	7, 28, 21, 4, 5, 2, 1,3

- 1	Cap. Bank		Generator	Trans.	Cod	le V																						\neg	
	7	21	28	Unit	Тар	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30
	0	0	0	0	1	-2	-2	0	0	-2	0	0	0	0	0	-3	0	-2	0	-2	0	0	-2	-2	0	0	0	-2	-2
1	1	0	0	0	1	0	-2	0	0	-2	0	0	0	0	0	-3	0	-2	0	0	0	0	-2	0	0	0	0	o	0
2	2	0	0	0	1	0	-2	0	0	0	0	0	0	0	0	-3	0	-2	0	0	0	0	0	0	0	0	0	0	0
3	3	0	0	0	1	0	0	0	0	0	0	0	0	0	0	-3	0	-2	o	0	0	0	0	0	0	0	0	0	0
4	4	0	0	0	1	0	0	0	0	0	0	0	0	0	0	-3	0	-2	0	0	0	0	0	0	0	0	0	0	0
5	4	0	0	0	1.01	0	0	0	0	0	0	0	0	0	0	-,3	0	-2	o	0	0	0	0	0	0	0	0	0	0
6	4	0	0	0	1.02	0	o	0	0	o	0	0	0	0	0	-3	0	-2	o	0	0	0	0	0	0	0	0	o	0
7	4	0	0	0	1.03	0	0	0	0	0	0	0	0	0	0	-3	0	-2	0	0	0	0	0	0	0	0	0	0	0
8	4	0	0	0	1.04	0	0	0	0	0	0	0	0	0	0	-3	0	-2	0	0	0	0	0	0	0	0	0	0	0
9	4	0	0	0	1.05	0	0	0	0	0	0	0	0	0	0	-3	0	-2	0	0	0	0	0	0	0	0	0	0	0
10	4	0	0	0	1.06	0	0	0	0	0	0	0	0	0	0	-3	0	0	0	0	0	0	0	0	0	0	0	0	0
11	4	0	0	0	1.07	0	0	0	0	0	0	0	0	0	0	-3	0	0	0	0	0	0	0	0	0	0	0	0	0
12	4	0	0	0	1.08	0	0	0	0	o	0	0	0	0	0	-1	0	0	o	0	0	0	0	0	0	0	0	0	0
13	4	0	0	0	1.09	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	0
14	4	0	0	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	0
15	4	0	1	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	o	0
16	4	0	2	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	o	0	0	0	0	0	0	0	0	0	0
17	4	0	3	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	0
18	4	0	4	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	0
19	4	1	4	0	1.1	0	0	0	0	0	0	0	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	0
20	4	2	4	0	1.1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0





Multi-Agent Based Stabilization Control using Energy Capacitor System

Fuzzy Logic Control Scheme - Operating Point

Operating Point on Polar Coordinates

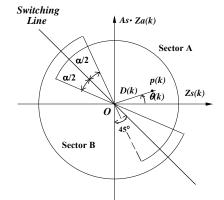
$$p(k) = [Zs(k) \ AsZa(k)]$$

$$D(k) = \sqrt{Z_s(k)^2 + (A_s \cdot Z_a(k))^2}$$

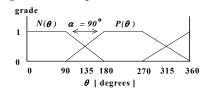
$$\theta(k) = \tan^{-1}(A_s \cdot Z_a(k)/Z_s(k))$$
Sector A
$$S(k) = \frac{\Delta S}{\delta(k)}$$
Sector B
$$S(k) = \frac{\Delta S}{\delta(k)}$$
Sector B

Speed/Acceleration Phase Plane

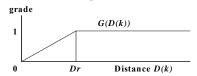
Membership Functions and Control Signal



Angle Membership Functions

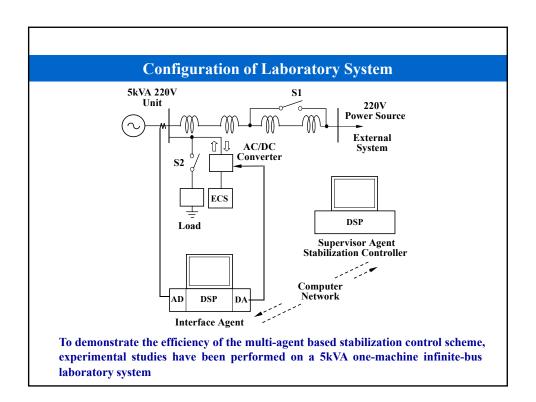


Radius Membership Function



Speed/Acceleration Phase Plane

$$\begin{aligned} Uecs(k) &= \frac{N(\theta(k)) - P(\theta(k))}{N(\theta(k)) + P(\theta(k))} \cdot G(D(k)) \cdot U_{max} \\ &= [1 - 2P(\theta(k))] \cdot G(D(k)) \cdot U_{max} \end{aligned}$$







ECS: Energy Capacitor System

Electrical Double Layer Capacitors

Capasity of ECS : 70 Wh (250kJ)

Maximum Charging Power: 3 kW

Maximum Discharging Power: 3 kW



