

$$Z = \alpha_1 - 2\alpha_2 + 4\alpha_3 + 1\alpha_1 + 2\alpha_2 + 3\alpha_3 + \alpha_4 + \alpha_5$$

Energy trading between multiple microgrids for enhancing the grid resilience

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May 7, 2021

problem
→ sub prob
fun f
main
→ sub 1
→ sub 2
→ sub 3

1 Research Objective

The research objective is to develop a framework to keep supplying the electric energy to maximize the social welfare even in case of disasters by trading the electric energy among multiple microgrids.

2 Research plan

- May 2021 : Making the program by considering the imputation in the core of the line between microgrids and solving the problem by gurobi optimizer

3 Research results

- Considering about the interconnection lines between microgrids (1.4851JPN calculated by gurobi)
- The imputation to the interconnection lines between the microgrids : the investment to the lines to increase the capacity by using this imputation, more efficient microgrid operation because If the profit of the grand coalition increases as the capacity of the line increases, the imputation to the line will increase.

Nomenclature

Suffix

i: The number of the players

Constant

c_i : The price coefficient C [JPY/p.u.]

b_i : The resource capacity [p.u.]

$Sline$: Line limit capacity [p.u.]

$Bline$: Line susceptance [p.u.]

KKT > wlmn
form dual
dual

$$\lambda = \frac{\partial J(\alpha)}{\partial \alpha}$$

Variable

y_i : The shadow price [JPY/p.u.]

θ_i : Phase angle of node i[Rad]

y_{line1} : The shadow price of the line at node 1[JPY/p.u.]

y_{line2} : The shadow price of the line at node 2[JPY/p.u.]

4 Formulation

The profit maximization problem of each microgrid can be formulated as a bi-level mathematical programming problem with the leader as each microgrid and the follower as a transaction manager as follows.

$$\underset{badd_{i=1}^{100}}{\text{Maximize}} \quad \sum_{i=1}^{100} (b_i + badd_i) y_i + yline1 \left(\frac{Sline}{Bline} \right) + yline2 \left(\frac{Sline}{Bline} \right) \quad (4.1)$$

$$\text{subject to } badd_i \geq 0 \quad \forall i \in N \quad (4.2)$$

$$\underset{x_{i=1}^{100}, \theta}{\text{Maximize}} \quad \sum_{i=1}^{100} c_i x_i \quad (4.3)$$

$$\text{subject to } 0 \leq x_i \leq b_i + badd_i \quad \forall i \in N \quad (4.4)$$

$$\sum_{i=1}^{25} x_{2i-1} - \sum_{i=26}^{50} x_{2i-1} = Bline(\theta_1 - \theta_2) \quad (4.5)$$

$$\sum_{i=1}^{25} x_{2i} - \sum_{i=26}^{50} x_{2i} = Bline(\theta_2 - \theta_1) \quad (4.6)$$

$$\left(\frac{-Sline}{Bline} \right) \leq \theta_2 \leq \left(\frac{Sline}{Bline} \right) \quad (4.7)$$

$$\theta_1 = 0 \quad (4.8)$$

The bi-level programming problem can be converted into equivalent constraints by applying KKT (Karush-Kuhn-Tucker) conditions to its lower level problem (4.3 - 4.8). The two-level programming problem can then be converted to a mixed integer linear programming problem.

(1) KKT condition

The KKT (Karush-Kuhn-Tucker) condition is the optimality condition that the optimal solution of various optimization problems should satisfy. The KKT condition is a necessary condition, and is a first-order condition, that is, using first-order derivative vectors and matrices. It is a condition to be formulated. [?]

Consider the following general mathematical programming problem.

$$\text{Minimize } f(x) \quad (4.9)$$

$$\text{subject to } g(x) \leq 0 \quad (4.10)$$

$$h(x) = 0 \quad (4.11)$$

The Lagrange function corresponding to this problem is

$$L(x, \lambda, \mu) = f(x) + \lambda g(x) + \mu h(x) \quad (4.12)$$

The KKT condition in question is

$$\nabla_x f(x) + \lambda \nabla_x g(x) = 0 \quad (4.13)$$

$$h(x) = 0 \quad (4.14)$$

$$g(x) \leq 0 \quad (4.15)$$

$$\mu g(x) = 0 \quad (4.16)$$

$$\mu \geq 0 \quad (4.17)$$

Where λ, μ is the Lagrange multiplier vector for equations and inequalities, and ∇ is the gradient for x . The constraint expression (4.15)-(4.17) is a complementarity condition,

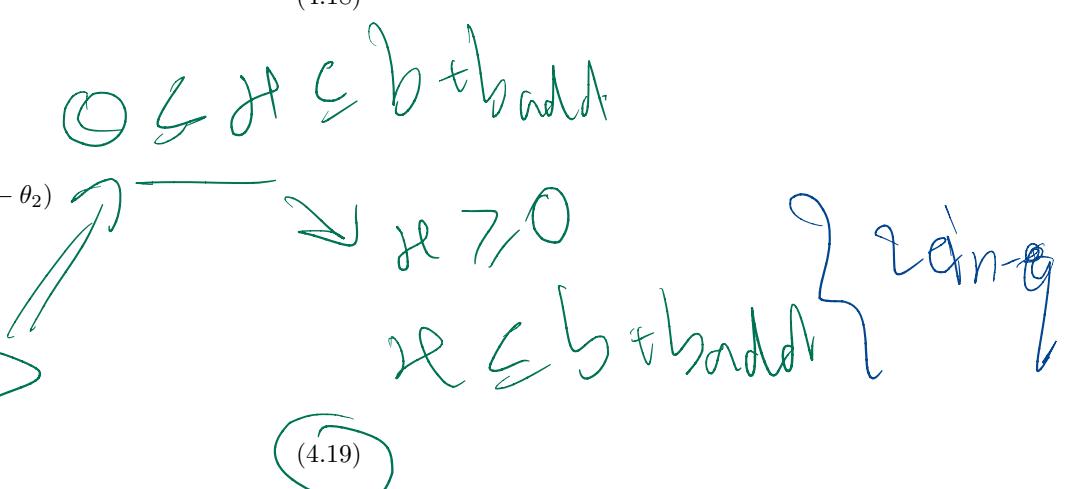
$$0 \leq \mu \perp g(x) \geq 0 \quad (4.18)$$

Here, \perp indicates complementarity, that is, $\mu g(x) = 0$.

(2) Lagrange function of microgrid i

The Lagrange function of the expression (4.19) is

$$\begin{aligned} L_i = & \sum_{i=1}^{100} c_i x_i + \mu^{Bline1} \left(\sum_{i=1}^{25} x_{2i-1} - \sum_{i=26}^{50} x_{2i-1} - Bline(\theta_1 - \theta_2) \right) \\ & + \mu^{Bline2} \left(\sum_{i=1}^{25} x_{2i} - \sum_{i=26}^{50} x_{2i} - Bline(\theta_2 - \theta_1) \right) \\ & + \rho_{b_i+badd}^{\min} \sum_{i=1}^{100} (-x_i) + \rho_{b_i+badd}^{\max} \sum_{i=1}^{100} (x_i - (b_i + badd_i)) \\ & + \phi_{\theta_2}^{\min} \left(\left(\frac{-Sline}{Bline} \right) - \theta_2 \right) + \phi_{\theta_2}^{\max} \left(\theta_2 - \left(\frac{Sline}{Bline} \right) \right) \end{aligned} \quad (4.19)$$



Where μ_i^{Bline1} , μ_i^{Bline2} , $+\rho_{b_i+badd}^{min}$, $\rho_{b_i+badd}^{max}$, $\phi_{\theta_2}^{min}$ and $\phi_{\theta_2}^{max}$ are Lagrange multiplier

(3)KKT condition of microgrid i

From the Lagrange function(4.19), the KKT condition is as follows

$$\frac{\partial L}{\partial \theta_2} = \mu_i^{Bline1}(Bline) - \mu_i^{Bline2}(Bline) - \phi_{\theta_2}^{min} + \phi_{\theta_2}^{max} = 0, \quad \forall i \in N \quad (4.20)$$

$$\frac{\partial L}{\partial x_i} = c_i + \mu_i^{Bline1} - \rho_{b_i+badd}^{min} + \rho_{b_i+badd}^{max} = 0, \quad \forall i \in N, i = 1, 3, 5, \dots, 47, 49 \quad (4.21)$$

$$\frac{\partial L}{\partial x_i} = c_i + \mu_i^{Bline2} - \rho_{b_i+badd}^{min} + \rho_{b_i+badd}^{max} = 0, \quad \forall i \in N, i = 2, 4, 6, \dots, 48, 50 \quad (4.22)$$

$$\frac{\partial L}{\partial x_i} = c_i - \mu_i^{Bline1} - \rho_{b_i+badd}^{min} + \rho_{b_i+badd}^{max} = 0, \quad \forall i \in N, i = 51, 53, 55, \dots, 97, 99 \quad (4.23)$$

$$\frac{\partial L}{\partial x_i} = c_i - \mu_i^{Bline2} - \rho_{b_i+badd}^{min} + \rho_{b_i+badd}^{max} = 0, \quad \forall i \in N, i = 52, 54, 56, \dots, 98, 100 \quad (4.24)$$

$$0 \leq x_i \perp \rho_{b_i+badd}^{min} \geq 0, \quad \forall i \in N \quad (4.25)$$

$$0 \leq (b_i + badd_i) - x_i \perp \rho_{b_i+badd}^{max} \geq 0, \quad \forall i \in N \quad (4.26)$$

$$0 \leq \theta_2 + \left(\frac{Sline}{Bline} \right) \perp \phi_{\theta_2}^{min} \geq 0, \quad \forall i \in N \quad (4.27)$$

$$0 \leq \left(\frac{Sline}{Bline} \right) - \theta_2 \perp \phi_{\theta_2}^{max} \geq 0, \quad \forall i \in N \quad (4.28)$$

$$\sum_{i=1}^{25} x_{2i-1} - \sum_{i=26}^{50} x_{2i-1} - Bline(\theta_1 - \theta_2) = 0 \quad (4.29)$$

$$\sum_{i=1}^{25} x_{2i} - \sum_{i=26}^{50} x_{2i} - Bline(\theta_2 - \theta_1) = 0 \quad (4.30)$$

The equation (4.25 - 4.28) are complementarity constraints which are non - linear equation . By replace them with linear constraint , we need to use Big - M method to solve the problem (4.31 - 4.32)

$$0 \leq \mu \perp g(x) \geq 0 \iff \begin{cases} 0 \leq \mu \leq Mu \\ 0 \leq g(x) \leq M(1-u) \end{cases} \quad (4.31)$$

$$(4.32)$$

The complementarity conditions (4.25 - 4.28) are replaced with the following linear constraints(4.33 - 4.40)

$$0 \leq \rho_{b_i+badd}^{min} \leq Mu_{b_i+badd}^{min}, \quad \forall i \in N \quad (4.33)$$

$$0 \leq x_i \leq M(1 - u_{b_i+badd}^{min}), \quad \forall i \in N \quad (4.34)$$

$$0 \leq \rho_{b_i+badd}^{max} \leq Mu_{b_i+badd}^{max}, \quad \forall i \in N \quad (4.35)$$

$$0 \leq (b_i + badd) - x_i \leq M(1 - u_{b_i+badd}^{max}), \quad \forall i \in N \quad (4.36)$$

$$0 \leq \phi_{\theta_2}^{min} \leq Mu_{\theta_2}^{min}, \quad \forall i \in N \quad (4.37)$$

$$0 \leq \theta_2 + \left(\frac{Sline}{Bline} \right) \leq M(1 - u_{\theta_2}^{min}), \quad \forall i \in N \quad (4.38)$$

$$0 \leq \phi_{\theta_2}^{max} \leq Mu_{\theta_2}^{max}, \quad \forall i \in N \quad (4.39)$$

$$0 \leq \left(\frac{Sline}{Bline} \right) - \theta_2 \leq M(1 - u_{\theta_2}^{max}), \quad \forall i \in N \quad (4.40)$$

5 Future tasks

Making the program by using gurobi optimizer

References

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$$\begin{aligned}
 \text{Larg}(b_{\text{add}}, \gamma, \gamma_{\text{line}}, \Delta) = & \sum_{i=1}^{100} (b_i + b_{\text{add},i}) \gamma_i + \gamma_{\text{line}1} \left(\frac{s_{\text{line}}}{B_{\text{line}}} \right) \\
 & + \gamma_{\text{line}2} \left(\frac{s_{\text{line}}}{B_{\text{line}}} \right) \\
 & + \sum_{i=1,3,\dots,49} \gamma_i (c_i + \mu_i^{\text{Blne1}} - p_{b_i+b_{\text{add}}}^{\min} + p_{b_i+b_{\text{add}}}^{\max}) \\
 & + \sum_{i=2,4,\dots,50} \gamma_i (c_i + \mu_i^{\text{Blne2}} - p_{b_i+b_{\text{add}}}^{\min} + p_{b_i+b_{\text{add}}}^{\max}) \\
 & + \sum_{i=51,\dots,99} \gamma_i (c_i - \mu_i^{\text{Blne1}} - p_{b_i+b_{\text{add}}}^{\min} + p_{b_i+b_{\text{add}}}^{\max}) \\
 & + \sum_{i=52,\dots,100} \gamma_i (c_i - \mu_i^{\text{Blne2}} - p_{b_i+b_{\text{add}}}^{\min} + p_{b_i+b_{\text{add}}}^{\max})
 \end{aligned}$$

$\{ \mu(b_{\text{add}})$

$p^{\min}(b_{\text{add}})$

$p^{\max}(b_{\text{add}})$

$$\min -C^T x.$$

$$x_i \stackrel{100}{=} 1, 0$$

subject to

$$\begin{cases} Ax \leq a \\ Bx = b \end{cases}$$



$$\begin{aligned} \text{Lang}(x, \lambda, v) &= -C^T x + \lambda^T (Ax - a) + v^T (Bx - b) \\ &= -C^T x + \lambda^T A x - \lambda^T a + v^T B x - v^T b \\ &= -\lambda^T a - v^T b + (-C^T + \lambda^T A + v^T B) x. \\ &= -\lambda^T a - v^T b + (-C + A^T \lambda + B^T v)^T x. \end{aligned}$$



$$\begin{array}{ll} \max_h & -\lambda^T a - v^T b \\ \text{subject} & -C + A^T \lambda + B^T v = 0 \end{array} \Rightarrow \begin{array}{ll} \min & \lambda^T a + b^T v \\ \lambda, v & \end{array}$$

$\lambda, v \Rightarrow$ no $y_i : y_i \in \text{line}_i$

\Rightarrow no margin

margin

(a) May/21

$$\min_{\alpha} \quad C^T \alpha$$

subject to $\alpha \geq 0$; (γ_1)
 $\alpha \leq b + b_{\text{add}}$; (γ_2) vector
 $k_1^T \alpha - k_2^T \alpha = -B \text{line}_2$; (γ_3) scalar
 $k_3^T \alpha - k_4^T \alpha = B \text{line}_2$; (γ_4) scalar

$$-\frac{\text{Sline}}{B \text{line}} \leq \theta_2 \leq \frac{\text{Sline}}{B \text{line}}; \quad (\gamma_5) \text{ scalar}$$

$$L = C^T \alpha - \gamma_1^T \alpha + \gamma_1^T (-b - b_{\text{add}}) + \gamma_2^T (k_1^T \alpha - k_2^T \alpha + B \text{line}_2) \\ + \gamma_3^T (k_3^T \alpha - k_4^T \alpha - B \text{line}_2) \\ + \gamma_{\text{line}_1}^T (-\theta_2 - \frac{\text{Sline}}{B \text{line}}) + \gamma_{\text{line}_2}^T (\theta_2 - \frac{\text{Sline}}{B \text{line}})$$

$$= (C^T - \gamma_1^T + \gamma_2^T k_1^T - \gamma_2^T k_2^T + \gamma_3^T k_3^T - \gamma_3^T k_4^T) \alpha \\ + (\gamma_2^T B \text{line} - \gamma_3^T B \text{line} - \gamma_{\text{line}_1}^T - \gamma_{\text{line}_2}^T) \theta_2$$

$$+ \gamma_1^T (-b - b_{\text{add}}) - \gamma_{\text{line}_1}^T \frac{\text{Sline}}{B \text{line}} + \gamma_{\text{line}_2}^T \frac{\text{Sline}}{B \text{line}}$$

maximize:

$$\max \gamma_1^T (-b - b_{\text{add}}) - \gamma_{\text{line}_1}^T \frac{\text{Sline}}{B \text{line}} - \gamma_{\text{line}_2}^T \frac{\text{Sline}}{B \text{line}}$$

$$\min \gamma_1^T (b + b_{\text{add}}) + \gamma_{\text{line}_1}^T \frac{\text{Sline}}{B \text{line}} + \gamma_{\text{line}_2}^T \frac{\text{Sline}}{B \text{line}}$$

Sub to $x_1 \geq 0, y_1 \geq 0$

$y_{\text{line}_1} \geq 0, y_{\text{line}_2} \geq 0$

$$C + x_1 + y_1 = (k_1 - k_2)y_2 + (k_3 - k_4)y_3 = 0$$

$$0 \leq y_2^{\text{Blne}} = y_3^{\text{Blne}} - y_{\text{line}_1} + y_{\text{line}_2} \perp 0 \geq 0$$

then we can write the L.P. function

(maximize obj func)

\rightarrow max λy_2 s.t. $0 \leq y_2 \leq 0$, $\lambda > 0$

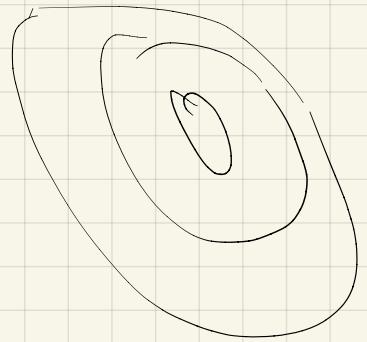
$$C + y - (k_1 - k_2)y_2 + (k_3 - k_4)y_3 \leq 0$$

min
sub $f(x)$
 $g(x) \leq 0$ \Rightarrow dual

obj:

$$h(x) = \begin{cases} 0 & \text{dual in eq} \\ \infty & \text{dual} \end{cases}$$

$$\mathcal{L}(x, \lambda, v) = f(x) + \lambda^T g(x) + v^T h(x)$$



$$\nabla_x \mathcal{L} = \nabla f(x) + \lambda^T \nabla g(x) + v^T \nabla h(x) = 0$$

$\nabla f(x)$

PRIMAL	maximize	minimize	DUAL
constraints	$\leq b_i$ $\geq b_i$ $= b_i$	≥ 0 ≤ 0	
	≥ 0 ≤ 0	<i>unconstrained</i>	<i>variables</i>
variables	$unconstrained$	$\geq c_j$ $\leq c_j$ $= c_j$	<i>constraints</i>

Note, using the rules in the above table, the dual of

$$\begin{array}{ll} \text{minimize} & c^T x \\ \text{subject to} & A^1 x \leq b^1 \\ & A^2 x \geq b^2 \\ & A^3 x = b^3 \end{array}$$

- Primal form(lower level) x_i, θ_2 $\neq \text{top}(\text{top})$

$$\text{Minimize}_{\substack{x^{100} \\ -\theta}} -c^T x \quad (4.9)$$

$$\text{subject to } x \geq 0 \quad -x_i \leq 0, \forall i \in N \quad ; \quad (\lambda_1) \bar{\lambda}_i^t \quad (4.10)$$

$$x_i \leq b_i + b_{add}, \forall i \in N \quad ; \quad (\mu_1) \lambda_i^+ \quad (4.11)$$

$$k_1^T x - k_2^T x = -Bline \theta_2 \quad ; \quad (\mu_2) \mu_1 \quad (4.12)$$

$$k_3^T x - k_4^T x = Bline \theta_2 \quad ; \quad (\mu_3) \mu_2 \quad (4.13)$$

$$\left(\frac{-Sline}{Bline} \right) \leq \theta_2 \leq \left(\frac{Sline}{Bline} \right) \quad ; \quad (y_{line_1}, y_{line_2}) \quad (4.14)$$

where

$$C = (c_1, c_2, \dots, c_{100})^T$$

$$k_1 = [1_{1 \times 25} \ 0_{1 \times 25} \ 0_{1 \times 25} \ 0_{1 \times 25}]^T$$

$$k_2 = [0_{1 \times 25} \ 1_{1 \times 25} \ 0_{1 \times 25} \ 0_{1 \times 25}]^T$$

$$k_3 = [0_{1 \times 25} \ 0_{1 \times 25} \ 1_{1 \times 25} \ 0_{1 \times 25}]^T$$

$$k_4 = [0_{1 \times 25} \ 0_{1 \times 25} \ 0_{1 \times 25} \ 1_{1 \times 25}]^T$$

use $\neq \text{top}$ instead of T
(top)

- Using ~~Lagrange multiplier~~ to convert to dual form

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Done

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strained variables in the dual is but one aspect that can be observed by looking at the above dual. Other features are summarized in the table below, which describes how to take the dual of a general linear program.

PRIMAL	maximize	minimize	DUAL
constraints	$\leq b_i$ $\geq b_i$ $= b_i$	≥ 0 ≤ 0	
variables	≥ 0 ≤ 0 $unconstrained$	$\geq c_j$ $\leq c_j$ $= c_j$	<i>variables</i>

Note, using the rules in the above table, the dual of

$$\begin{array}{ll} \text{minimize} & c^T x \\ \text{subject to} & A^1 x \leq b^1 \\ & A^2 x \geq b^2 \\ & A^3 x = b^3 \end{array}$$

becomes

$$\begin{array}{ll} \text{maximize} & (b^1)^T y^1 + (b^2)^T y^2 + (b^3)^T y^3 \\ \text{subject to} & (A^1)^T y^1 + (A^2)^T y^2 + (A^3)^T y^3 = c \\ & y^1 \leq 0 \\ & y^2 \geq 0 \end{array}$$

Since the dual of the dual is the primal, reorganizing the above table yields an alternative procedure for converting primals that involve minimization to their duals.

$$\text{Min} \quad -c^T x$$

$$x_i, \theta$$

$$\text{subject to } \begin{cases} x \geq 0 & ; \quad \lambda \\ x \leq b_i + b_{add} & ; \quad \mu \end{cases}$$

$$(k_1 - k_2)^T Bline \theta_2 \leq k_1^T x - k_2^T x = -Bline \cdot \theta_2 \quad ; \quad \mu_1$$

$$k_3^T x - k_4^T x - Bline \theta_2 \leq k_3^T x - k_4^T x = Bline \cdot \theta_2 \quad ; \quad \mu_2$$

$$[(k_1 - k_2)^T Bline] \theta_2 \geq -\frac{Sline}{Bline} \quad ; \quad \mu_1 \geq -\frac{Sline}{Bline}$$

$$[(k_3 - k_4)^T Bline] \theta_2 \leq \frac{Sline}{Bline} \quad ; \quad \mu_2 \leq \frac{Sline}{Bline}$$

R

$$\text{Max} \quad \underline{\lambda^T \cdot \lambda^-} + \lambda^{+T} (b + b_{add}) + 0 \mu_1 + 0 \mu_2 + y_{line_1} \cdot \left(-\frac{Sline}{Bline} \right) + y_{line_2} \cdot \left(\frac{Sline}{Bline} \right)$$

$$\lambda^- \leq 0$$

$$y_{line_1} \leq 0$$

$$\lambda^+ \geq 0$$

$$y_{line_2} \geq 0$$

$\mu_1 \Rightarrow \text{unconstrained}$

$\mu_2 \Rightarrow \text{unconstrained}$

$$\begin{bmatrix} 1 \\ 0 \end{bmatrix} \lambda^- + \begin{bmatrix} 1 \\ 0 \end{bmatrix} \lambda^+ + \begin{bmatrix} k_1 - k_2 \\ Bline \end{bmatrix} \mu_1 + \begin{bmatrix} k_3 - k_4 \\ Bline \end{bmatrix} \mu_2 = \begin{bmatrix} -C \\ 0 \end{bmatrix}$$

LP

max

\hat{J}_x



$$A_1 x \leq b_1$$

$$A_2^T x = b_2$$

$$A_3^T x = b_3$$

$$x \geq 0$$

→ how to solve solver

Python -
gekko -

min.

\hat{J}_x

dual.

max

$$b^T y_1 + b^T y_2 \rightarrow \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

$$A_1 x \geq b^T$$

$$A_2 x = b^T$$

$$x \geq 0$$

$$y \leq 0$$

$$y_2 \rightarrow \text{unco}$$

$$A_1^T y_1 + A_2^T y_2 \geq 0$$

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$$\begin{array}{c} \mathcal{H} \leq 0 \\ (-\mathcal{H}) > 0 \\ \downarrow \\ \mathcal{H} \end{array} \quad \left| \quad \begin{array}{l} \mathcal{H} \Rightarrow \text{uncon.} \\ \mathcal{H} = \mathcal{H}^+ - \mathcal{H}^- \\ \mathcal{H}^+, \mathcal{H}^- \geq 0 \\ 0 \perp \end{array} \right.$$

$$\min c^T x \Rightarrow \begin{array}{l} \mathcal{H}^+ - \mathcal{H}^- \\ \Rightarrow c^T \mathcal{H}^+ - c^T \mathcal{H}^- \end{array}$$

$$\begin{array}{ll} A_1^1 x \leq b_1 & \Rightarrow A_1^1 \mathcal{H}^+ - A_1^1 \mathcal{H}^- \leq b_1 \\ A_2^2 x \geq b_2 & \Rightarrow A_2^2 \mathcal{H}^+ - A_2^2 \mathcal{H}^- \geq b_2 \\ \underline{A_3^3 x = b_3} & \Rightarrow \underline{A_3^3 \mathcal{H}^+ - A_3^3 \mathcal{H}^- \geq b_3} \\ \mathcal{H}^+, \mathcal{H}^- \end{array}$$

$$x^T + \mu_1^+ - \mu_1^- \leq -c_1$$

$$\mu_2^+ - \mu_2^- \leq$$

$$\begin{array}{l} y^{\text{lin}} + y^{\text{line}} \\ \epsilon B (\mu_1^+ - \mu_1^-) + B^{\text{lin}} (\mu_2^+ - \mu_2^-) \geq 0 \end{array}$$

$$\mu_1$$

$$\max_{b_{\text{add}}} (b + b_{\text{add}})^T y + Y_{\text{line}_1} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right) + Y_{\text{line}_2} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right)$$

sub to $b_{\text{add}} > 0$

$$\min (b + b_{\text{add}})^T y + Y_{\text{line}_1} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right) + Y_{\text{line}_2} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right)$$

$$\begin{bmatrix} 1 \\ 100 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} y \\ \mu \end{bmatrix} \leq \begin{bmatrix} -1 \\ 100 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \begin{bmatrix} y \\ \mu \end{bmatrix} \leq -c$$

Condition $\frac{b^T y}{y}$

$$[1 - 1 \ B_{\text{line}} : -B_{\text{line}}] \begin{bmatrix} Y_{\text{line}} \\ \mu \end{bmatrix} = 0 \quad y \geq 0, Y_{\text{line}}, \exists 0, Y_{\text{line}_2} \geq 0$$

$$L = (b + b_{\text{add}})^T y + Y_{\text{line}_1} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right) + Y_{\text{line}_2} \left(\frac{S_{\text{line}}}{B_{\text{line}}} \right) + f^T (-y + \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \\ 0 & -1 \end{bmatrix} \mu + c)$$

$$+ \omega (Y_{\text{line}_1} - Y_{\text{line}_2} + [B_{\text{line}} - B_{\text{line}}] \mu)$$

$$+ \beta_3^T y + \beta_f \cdot Y_{\text{line}_1} + \beta_g \cdot Y_{\text{line}_2}$$

$$\frac{dc}{dy} = b + b_{\text{add}} - f + \beta = 0$$

$$\max c^T f$$

$$Y_{\text{line}_1} - Y_{\text{line}_2} + [B_{\text{line}} - B_{\text{line}}] \mu = 0$$

$$0 \leq f + (y - \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \\ 0 & -1 \end{bmatrix} \mu) \geq 0$$

$$0 \leq y \leq M u_2$$

$$0 \leq -f \leq M(-u)$$

$$0 \leq f \leq M u_1$$

$$0 \leq y - \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \\ 0 & -1 \end{bmatrix} \mu \leq M(-u)$$

$$0 \leq Y_{\text{line}} \leq M u_2$$

$$0 \leq -f \leq M(-u)$$

Energy trading between multiple microgrids for enhancing the grid resilience

Kanyakorn Likhitluecha

June 11, 2021

1 Research Objective

The research objective is to develop a framework to keep supplying the electric energy to maximize the social welfare even in case of disasters by trading the electric energy among multiple microgrids.

2 Research plan

- May 2021 : Making the program by considering the imputation in the core of the line between microgrids and solving the problem by gurobi optimizer

3 Research results

- Considering about the interconnection lines between microgrids (1.4851JPN calculated by gurobi)
- The imputation to the interconnection lines between the microgrids : the investment to the lines to increase the capacity by using this imputation, more efficient microgrid operation because If the profit of the grand coalition increases as the capacity of the line increases, the imputation to the line will increase.

Nomenclature

Suffix

i : The number of the players

Constant

c_i : The price coefficient C [JPY/p.u.]

b_i : The resource capacity [p.u.]

S_{line} : Line limit capacity [p.u.]

B_{line} : Line susceptance [p.u.]

Variable

y_i : The shadow price [JPY/p.u.]

θ_i : Phase angle of node i [rad]

y_{line_1} : The shadow price of the line at node 1[JPY/p.u.]

y_{line_2} : The shadow price of the line at node 2[JPY/p.u.]

4 Formulation

The profit maximization problem of each microgrid can be formulated as a bi-level mathematical programming problem with the leader as each microgrid and the follower as a transaction manager as follows.

$$\text{Maximize}_{\substack{b_{add}^{100} \\ i=1}} \sum_{i=1}^{100} (b_i + b_{add_i})y_i + y_{line_1} \left(\frac{S_{line}}{B_{line}} \right) + y_{line_2} \left(\frac{S_{line}}{B_{line}} \right) \quad (4.1)$$

subject to $b_{add_i} \geq 0 \quad \forall i \in N$ (4.2)

$$\text{Minimize}_{\substack{x_i, \theta_2}} -c_1x_1 - c_2x_2 - c_3x_3 - c_4x_4 \quad (4.3)$$

subject to $-x_1 \geq -(b_1 + b_{add_1}) \quad ; \quad (\lambda_1^+)$ (4.4)
 $-x_2 \geq -(b_2 + b_{add_2}) \quad ; \quad (\lambda_2^+)$ (4.5)
 $-x_3 \geq -(b_3 + b_{add_3}) \quad ; \quad (\lambda_3^+)$ (4.6)
 $-x_4 \geq -(b_4 + b_{add_4}) \quad ; \quad (\lambda_4^+)$ (4.7)

$$\theta_2 \geq \left(\frac{-S_{line}}{B_{line}} \right) \quad ; \quad y_{line_1} \quad (4.8)$$

$$\theta_2 \leq \left(\frac{S_{line}}{B_{line}} \right) \quad ; \quad y_{line_2} \quad (4.9)$$

is | | | |

$$x_1 - x_2 + B_{line}\theta_2 = 0 \quad ; \quad (\mu_1) \quad (4.10)$$

$$x_3 - x_4 - B_{line}\theta_2 = 0 \quad ; \quad (\mu_2) \quad (4.11)$$

$$x_1 \geq 0 \quad ; \quad (\beta_1) \quad (4.12)$$

$$x_2 \geq 0 \quad ; \quad (\beta_2) \quad (4.13)$$

$$x_3 \geq 0 \quad ; \quad (\beta_3) \quad (4.14)$$

$$x_4 \geq 0 \quad ; \quad (\beta_4) \quad (4.15)$$

The bi-level programming problem can be converted into equivalent constraints by applying KKT (Karush-Kuhn-Tucker) conditions to its lower level problem (4.3 - 4.15). The two-level programming problem can then be converted to a mixed integer linear programming problem.

(1)KKT condition

The KKT (Karush-Kuhn-Tucker) condition is the optimality condition that the optimal solution of various optimization problems should satisfy. The KKT condition is a necessary condition, and is a first-order condition, that is, using first-order derivative vectors and matrices. It is a condition to be formulated.

Consider the following general mathematical programming problem.

$$\text{Minimize} \quad f(x) \quad (4.16)$$

$$\text{subject to} \quad g(x) \leq 0 \quad (4.17)$$

$$h(x) = 0 \quad (4.18)$$

The Lagrange function corresponding to this problem is

$$L(x, \lambda, \mu) = f(x) + \lambda g(x) + \mu h(x) \quad (4.19)$$

The KKT condition in question is

$$\nabla_x f(x) + \lambda \nabla_x g(x) = 0 \quad (4.20)$$

$$h(x) = 0 \quad (4.21)$$

$$g(x) \leq 0 \quad (4.22)$$

$$\mu g(x) = 0 \quad (4.23)$$

$$\mu \geq 0 \quad (4.24)$$

Where λ, μ is the Lagrange multiplier vector for equations and inequalities, and ∇ is the gradient for x . The constraint expression (4.22)-(4.24) is a complementarity condition,

$$0 \leq \mu \perp g(x) \geq 0 \quad (4.25)$$

Here, \perp indicates complementarity, that is, $\mu g(x) = 0$.

(2)Lagrange function of microgrid i

The Lagrange function of the expression (4.26) is

$$\begin{aligned}
 L_i = & (-c_1x_1 - c_2x_2 - c_3x_3 - c_4x_4) + \mu_1(x_1 - x_2 + B_{line}\theta_2) \\
 & + \mu_2(x_3 - x_4 - B_{line}\theta_2) + \lambda_1^+(-(b_1 + b_{add_1}) + x_1) + \lambda_2^+(-(b_2 + b_{add_2}) + x_2) \\
 & + \lambda_3^+(-(b_3 + b_{add_3}) + x_3) + \lambda_4^+(-(b_4 + b_{add_4}) + x_4) + y_{line_1}(-\left(\frac{S_{line}}{B_{line}}\right) - \theta_2) \\
 & + y_{line_2}(\theta_2 - \left(\frac{S_{line}}{B_{line}}\right)) + \beta_1(-x_1) + \beta_2(-x_2) + \beta_3(-x_3) + \beta_4(-x_4)
 \end{aligned} \tag{4.26}$$

Where $\mu_1, \mu_2, \lambda_1^+, \lambda_2^+, \lambda_3^+, \lambda_4^+, y_{line_1}, y_{line_2}, \beta_1, \beta_2, \beta_3$ and β_4 are Lagrange multiplier

(3) KKT condition of microgrid i

From the Lagrange function(4.26), the KKT condition is as follows

$$\frac{\partial L}{\partial \theta_2} = \mu_1(B_{line}) - \mu_2(B_{line}) - y_{line_1} + y_{line_2} = 0 \tag{4.27}$$

$$\frac{\partial L}{\partial x_1} = -c_1 + \mu_1 + \lambda_1^+ - \beta_1 = 0 \tag{4.28}$$

$$\frac{\partial L}{\partial x_2} = -c_2 - \mu_1 + \lambda_2^+ - \beta_2 = 0 \tag{4.29}$$

$$\frac{\partial L}{\partial x_3} = -c_3 + \mu_2 + \lambda_3^+ - \beta_3 = 0 \tag{4.30}$$

$$\frac{\partial L}{\partial x_4} = -c_4 - \mu_2 + \lambda_4^+ - \beta_4 = 0 \tag{4.31}$$

$$0 \leq -x_1 + (b_1 + b_{add_1}) \perp \lambda_1^+ \geq 0 \tag{4.32}$$

$$0 \leq -x_2 + (b_2 + b_{add_2}) \perp \lambda_2^+ \geq 0 \tag{4.33}$$

$$0 \leq -x_3 + (b_3 + b_{add_3}) \perp \lambda_3^+ \geq 0 \tag{4.34}$$

$$0 \leq -x_4 + (b_4 + b_{add_4}) \perp \lambda_4^+ \geq 0 \tag{4.35}$$

$$0 \leq \theta_2 + \left(\frac{S_{line}}{B_{line}}\right) \perp y_{line_1} \geq 0 \tag{4.36}$$

$$0 \leq \left(\frac{S_{line}}{B_{line}}\right) - \theta_2 \perp y_{line_2} \geq 0 \tag{4.37}$$

$$0 \leq x_1 \perp \beta_1 \geq 0 \tag{4.38}$$

$$0 \leq x_2 \perp \beta_2 \geq 0 \tag{4.39}$$

$$0 \leq x_3 \perp \beta_3 \geq 0 \tag{4.40}$$

$$0 \leq x_4 \perp \beta_4 \geq 0 \tag{4.41}$$

$$x_1 - x_2 + B_{line}\theta_2 = 0 \tag{4.42}$$

$$x_3 - x_4 - B_{line}\theta_2 = 0 \tag{4.43}$$

The equation (4.32 - 4.41) are complementarity constraints which are non - linear equation . By replace them with linear constraint , we need to use Big - M method to solve the problem (4.44 - 4.45)

$$0 \leq \mu \perp g(x) \geq 0 \iff \begin{cases} 0 \leq \mu \leq Mu \\ 0 \leq g(x) \leq M(1-u) \end{cases} \tag{4.44}$$

$$(4.45)$$

The complementarity conditions (4.32 - 4.41) are replaced with the following linear constraints(4.46 - 4.65)

$$0 \leq \lambda_1^+ \leq Mu_{\lambda_1^+} \quad (4.46)$$

$$0 \leq -x_1 + (b_1 + b_{add_1}) \leq M(1 - u_{\lambda_1^+}) \quad (4.47)$$

$$0 \leq \lambda_2^+ \leq Mu_{\lambda_2^+} \quad (4.48)$$

$$0 \leq -x_2 + (b_2 + b_{add_2}) \leq M(1 - u_{\lambda_2^+}) \quad (4.49)$$

$$0 \leq \lambda_3^+ \leq Mu_{\lambda_3^+} \quad (4.50)$$

$$0 \leq -x_3 + (b_3 + b_{add_3}) \leq M(1 - u_{\lambda_3^+}) \quad (4.51)$$

$$0 \leq \lambda_4^+ \leq Mu_{\lambda_4^+} \quad (4.52)$$

$$0 \leq -x_4 + (b_4 + b_{add_4}) \leq M(1 - u_{\lambda_4^+}) \quad (4.53)$$

$$0 \leq y_{line_1} \leq Mu_{y_{line_1}} \quad (4.54)$$

$$0 \leq \theta_2 + \left(\frac{S_{line}}{B_{line}} \right) \leq M(1 - u_{y_{line_1}}) \quad (4.55)$$

$$0 \leq y_{line_2} \leq Mu_{y_{line_2}} \quad (4.56)$$

$$0 \leq \left(\frac{S_{line}}{B_{line}} \right) - \theta_2 \leq M(1 - u_{y_{line_2}}) \quad (4.57)$$

$$0 \leq \beta_1 \leq Mu_{\beta_1} \quad (4.58)$$

$$0 \leq x_1 \leq M(1 - u_{\beta_1}) \quad (4.59)$$

$$0 \leq \beta_2 \leq Mu_{\beta_2} \quad (4.60)$$

$$0 \leq x_2 \leq M(1 - u_{\beta_2}) \quad (4.61)$$

$$0 \leq \beta_3 \leq Mu_{\beta_3} \quad (4.62)$$

$$0 \leq x_3 \leq M(1 - u_{\beta_3}) \quad (4.63)$$

$$0 \leq \beta_4 \leq Mu_{\beta_4} \quad (4.64)$$

$$0 \leq x_4 \leq M(1 - u_{\beta_4}) \quad (4.65)$$

If both the primal and dual problems are feasible, and the KKT condition is satisfied, the strong duality theorem shows ;

$$\begin{aligned} -c_1x_1 - c_2x_2 - c_3x_3 - c_4x_4 = & -\lambda_1^+(b_1 + b_{add_1}) - \lambda_2^+(b_2 + b_{add_2}) - \lambda_3^+(b_3 + b_{add_3}) - \lambda_4^+(b_4 + b_{add_4}) \\ & - y_{line_1} \left(\frac{S_{line}}{B_{line}} \right) - y_{line_2} \left(\frac{S_{line}}{B_{line}} \right) \end{aligned} \quad (4.66)$$

Therefore , the two-level mathematical programming problem is reduced to the following mixed integer linear

$$\textcircled{1} \quad \left[\begin{array}{|c|c|c|c|c|c|c|c|c|c|c|c|} \hline 0 & 0 & 0 & 0 & B_{line} & -B_{line} & 0 & 0 & -1 & 1 & 1 & 0_{202} \\ \hline \end{array} \right]$$

\downarrow

zeros($1 \geq \text{len}(R)$)

$-C$

$\theta_{1-100}, \theta_2, \mu_1, \mu_2, \beta_{100}, \lambda_{1-100}, y_{line_1}, y_{line_2}$
 programming problem as follows ;

$b_{add_{1-100}}, \theta_2, \mu_1, \mu_2, \beta_{100}, \lambda_{1-100}, y_{line_1}, y_{line_2}$
 eq cons x_{20} $\leq b_{add}$ y_{line}

$[\$][b_{add}] \geq 0$ (4.67)

Maximize $-c_1x_1 - c_2x_2 - c_3x_3 - c_4x_4$
 subject to $b_{add_i} \geq 0 \quad \forall i \in N$ no big-M (4.68)

$\mu_1(B_{line}) - \mu_2(B_{line}) - y_{line_1} + y_{line_2} = 0 [B_{line} - B_{line} \rightarrow 1] M$ (4.69)

$-c_1 + \mu_1 + \lambda_1^+ - \beta_1 = 0$ (4.70)

$-c_2 + \mu_1 + \lambda_2^+ - \beta_2 = 0$ (4.71)

$-c_3 + \mu_2 + \lambda_3^+ - \beta_3 = 0$ (4.72)

$-c_4 + \mu_2 + \lambda_4^+ - \beta_4 = 0$ (4.73)

$x_1 - x_2 + B_{line}\theta_2 = 0$ (4.74)

$x_3 - x_4 - B_{line}\theta_2 = 0$ (4.75)

$\theta \leq b_{add}$

$0 \leq \lambda_1^+ \leq Mu_{\lambda_1^+} \Rightarrow \lambda \leq Mu$ (4.76)

$0 \leq -x_1 + (b_1 + b_{add_1}) \leq M(1 - u_{\lambda_1^+})$ (4.77)

$0 \leq \lambda_2^+ \leq Mu_{\lambda_2^+}$ (4.78)

$0 \leq -x_2 + (b_2 + b_{add_2}) \leq M(1 - u_{\lambda_2^+})$ (4.79)

$0 \leq \lambda_3^+ \leq Mu_{\lambda_3^+}$ (4.80)

$0 \leq -x_3 + (b_3 + b_{add_3}) \leq M(1 - u_{\lambda_3^+})$ (4.81)

$0 \leq \lambda_4^+ \leq Mu_{\lambda_4^+}$ (4.82)

$0 \leq -x_4 + (b_4 + b_{add_4}) \leq M(1 - u_{\lambda_4^+})$ (4.83)

$0 \leq y_{line_1} \leq Mu_{y_{line_1}}$ (4.84)

$0 \leq \theta_2 + \left(\frac{S_{line}}{B_{line}}\right) \leq M(1 - u_{y_{line_1}})$ (4.85)

$0 \leq y_{line_2} \leq Mu_{y_{line_2}}$ (4.86)

$0 \leq \left(\frac{S_{line}}{B_{line}}\right) - \theta_2 \leq M(1 - u_{y_{line_2}})$ (4.87)

$0 \leq \beta_1 \leq Mu_{\beta_1}$ (4.88)

$0 \leq x_1 \leq M(1 - u_{\beta_1})$ (4.89)

$0 \leq \beta_2 \leq Mu_{\beta_2}$ (4.90)

$0 \leq x_2 \leq M(1 - u_{\beta_2})$ (4.91)

$0 \leq \beta_3 \leq Mu_{\beta_3}$ (4.92)

$0 \leq x_3 \leq M(1 - u_{\beta_3})$ (4.93)

$0 \leq \beta_4 \leq Mu_{\beta_4}$ (4.94)

$0 \leq x_4 \leq M(1 - u_{\beta_4})$ (4.95)

$\theta \leq M - Mu$
 $\theta \times Mu \leq M - S$

$-\theta_2 + \frac{S}{B} \leq M - Mu$
 $-\theta_2 + Mu \leq M - S$

$-\theta_2 + \frac{S}{B} \leq M - Mu$
 $-\theta_2 + Mu \leq M - S$

- Now , I am in the process of making the program by using gurobi

5 Future tasks

Continue making the program by using gurobi optimization

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A hand-drawn diagram of a vertical column labeled "Varia" with various points marked along its length. The points are:

- β
- L_{100}
- β_{odd}
- L_{100}
- β_2
- M_1
- M_2
- β
- L_{100}
- γ_{line_1}
- γ_{line_2}
- W
- W_p
- M_{line}

$$\begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & M \\ M & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & M \\ M & 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} = \begin{bmatrix} S(B) \\ S(B) \\ M-S \\ M-S \\ B \end{bmatrix}$$

$$\begin{array}{c}
 \text{R} \quad \text{back} \quad \theta_2 \quad M_1 \quad M_2 \quad \beta_{1-100} \quad \gamma_{100} \quad \gamma_{\text{line}} \quad u_x \quad u_y \quad u_{y\text{line}} \\
 \text{①} \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & -B_{\text{line}} & B_{\text{line}} & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\
 \hline
 \begin{pmatrix} 0_{100 \times 100} & 0_{100 \times 100} & 0_{100 \times 1} & I_{25 \times 1} & 0_{25 \times 1} & -I_{100} & I_{100} & 0_{100 \times 1} & 0_{100 \times 1} & 0_{100 \times 100} & 0_{100 \times 100} & 0_{100 \times 1} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \\
 \hline
 \begin{pmatrix} -1_{1 \times 25} & 0_{1 \times 25} & 0_{1 \times 25} & 0_{1 \times 25} & 0_{2 \times 100} & B_{\text{line}} & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0_{1 \times 25} & 0_{1 \times 25} & 0_{1 \times 25} & -1_{1 \times 25} & -B_{\text{line}} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}
 \end{array}$$

$$\begin{array}{c|c|c|c|c|c|c|c|c} \textbf{0}_{(100 \times 1)} & -\textbf{I}_{(100)} & \textbf{0}_{(100 \times 1)} & \textbf{0}_{(100 \times 1)} & \textbf{0}_{(100 \times 1)} \\ \textbf{0}_{(100 \times 1)} & -\textbf{I}_{(100)} & \textbf{0}_{(100 \times 1)} & \textbf{0}_{(100 \times 1)} & \textbf{0}_{(100 \times 1)} \\ \hline \end{array}$$

$$I_{(00)} - MI$$

$$\begin{array}{ccccccc} R & bdd & & & U & \subseteq & [0] \\ I_{(0,0)} & -I_{(0,0)} & & & O & \subseteq & M.I \\ -I_{(0,0)} & I_{(0,0)} & & & M.I & \subseteq & \mathbb{N} \end{array}$$

$$B_2 \geq 0 \quad \Leftrightarrow \quad -M I \leq 0$$

$$\mathcal{H} \rightarrow \mathbb{I}_{(00)} \quad \text{and} \quad \mathbb{I}_{(00)} \rightarrow \mathcal{O}$$

The diagram illustrates a chemical mechanism:

- $\text{Ylme} + \text{I}_2 \rightarrow \text{Iy}$
- $\text{Iy} + \text{M}-\text{I}_2 \rightarrow \text{MIy}$
- $\text{MIy} + \text{I}_2 \rightarrow \text{MIyI}$
- $\text{MIyI} + \text{SIB} \rightarrow \text{M}-\text{SIB} + \text{I}_2$

