

# Process Control: Part II- Model Predictive Control (EE6225, AY2019/20, S1)

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1. Answer the following questions about model predictive control (MPC)

(a) How to select the sampling time of MPC?  
(1 Marks)

(b) How to select the prediction horizon of MPC;  
(1 Marks)

(c) How to select the control horizon of MPC;  
(1 Marks)

(d) How to select the constrains of MPC;  
(1 Marks)

(e) How to select the weights of MPC;  
(1 Marks)

- **Sampling time:** Set the sampling time between 5% and 10% of the minimum desired closed-loop response time.
- **Prediction horizon:** If the setting time of the system is  $T_{set}$ , the prediction time should be larger than  $T_{set}$ .
- **Control horizon:** The control horizon  $m$  is selected between 0.1  $P$  and 0.2  $P$ , where  $P$  is the prediction horizon.
- **Constrains:** For the output variable, soft constrain is recommended, which allow to break the limits. In addition, hard constrains should be avoided to be given to the input variable and its variable rate at the same time.
- **Weights:** Larger weight for important cost item; less weight for unimportant cost item.

2. For a system, its CARIMA model is as follows

$$A(z)y_k = b(z)\Delta u_k \quad (1)$$

where

$$a(z) = 1 - 0.5z^{-1} \quad (2)$$

$$b(z) = 3z^{-1} + 2z^{-2} \quad (3)$$

The CARIMA model based MPC expression is as follows:

$$\underline{y}_{k+1} = C_A^{-1}C_b \cdot \Delta \underline{u}_k + \left( C_A^{-1}H_b \cdot \Delta \underline{u}_{k-1} - C_A^{-1}H_A \cdot \underline{y}_k \right) \quad (4)$$

The prediction horizon is 4, please answer the following questions:

$$A(z)y_k = b(z)\Delta u_k$$

$$a(z) = 1 - 0.5z^{-1}$$

$$b(z) = 3z^{-1} + 2z^{-2}$$

□ Using the definition of the prediction matrices and a horizon of 4.

$$\left. \begin{array}{l} a(z) = 1 - 0.5z^{-1} \\ \Delta = 1 - z^{-1} \end{array} \right\} \Rightarrow A(z) = a(z)\Delta = 1 - \overset{A_1}{\uparrow} 1.5z^{-1} + \overset{A_2}{\nwarrow} 0.5z^{-2} + \overset{A_3, 4, \dots}{\uparrow} 0$$

$$C_A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ A_1 & 1 & 0 & 0 \\ A_2 & A_1 & 1 & 0 \\ A_3 & A_2 & A_1 & 1 \end{bmatrix} \Rightarrow C_A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1.5 & 1 & 0 & 0 \\ 0.5 & -1.5 & 1 & 0 \\ 0 & 0.5 & -1.5 & 1 \end{bmatrix}$$

$$H_A = \begin{bmatrix} A_1 & A_2 \\ A_2 & A_3 \\ A_3 & A_4 \\ A_4 & A_5 \end{bmatrix} \Rightarrow H_A = \begin{bmatrix} -1.5 & 0.5 \\ 0.5 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$A(z)y_k = b(z)\Delta u_k$$

$$a(z) = 1 - 0.5z^{-1}$$

$$b(z) = 3z^{-1} + 2z^{-2}$$

□ Using the definition of the prediction matrices and a horizon of 4.

$$b(z) = 3z^{-1} + 2z^{-2} \Rightarrow b_1 = 3 \quad b_2 = 2 \quad b_{3,4,\dots} = 0$$

$$C_b = \begin{bmatrix} b_1 & 0 & 0 & 0 \\ b_2 & b_1 & 0 & 0 \\ b_3 & b_2 & b_1 & 0 \\ b_4 & b_3 & b_2 & b_1 \end{bmatrix} \Rightarrow C_b = \begin{bmatrix} 3 & 0 & 0 & 0 \\ 2 & 3 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ 0 & 0 & 2 & 3 \end{bmatrix}$$

$$H_b = \begin{bmatrix} b_2 \\ b_3 \\ b_4 \\ b_5 \end{bmatrix} \Rightarrow H_b = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$$A(z) \mathbf{y}_k = b(z) \Delta \mathbf{u}_k$$

$$a(z) = 1 - 0.5z^{-1}$$

$$b(z) = 3z^{-1} + 2z^{-2}$$

□ Using the definition of the prediction matrices and a horizon of 4.

$$\underline{\mathbf{y}}_{k+1} = \mathbf{H} \cdot \Delta \underline{\mathbf{u}}_k + \left( \mathbf{P} \cdot \Delta \underline{\mathbf{u}}_{k-1} - \mathbf{Q} \cdot \underline{\mathbf{y}}_k \right)$$

$$\mathbf{H} = \mathbf{C}_A^{-1} \mathbf{C}_b$$

$$\mathbf{P} = \mathbf{C}_A^{-1} \mathbf{H}_b$$

$$\mathbf{Q} = \mathbf{C}_A^{-1} \mathbf{H}_A$$

$$\mathbf{C}_A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ -1.5 & 1 & 0 & 0 \\ 0.5 & -1.5 & 1 & 0 \\ 0 & 0.5 & -1.5 & 1 \end{bmatrix}$$

$$\mathbf{C}_b = \begin{bmatrix} 3 & 0 & 0 & 0 \\ 2 & 3 & 0 & 0 \\ 0 & 2 & 3 & 0 \\ 0 & 0 & 2 & 3 \end{bmatrix}$$

$$\mathbf{H}_A = \begin{bmatrix} -1.5 & 0.5 \\ 0.5 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$\mathbf{H}_b = \begin{bmatrix} 2 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

# APPLICATION OF FINITE CONTROL SET MPC IN THREE PHASE INVERTER

[07/11/2019]

- Basic knowledge of power inverters
- Finite Control Set (FCS) MPC for power inverters
- Improved FCS-MPC for power inverters

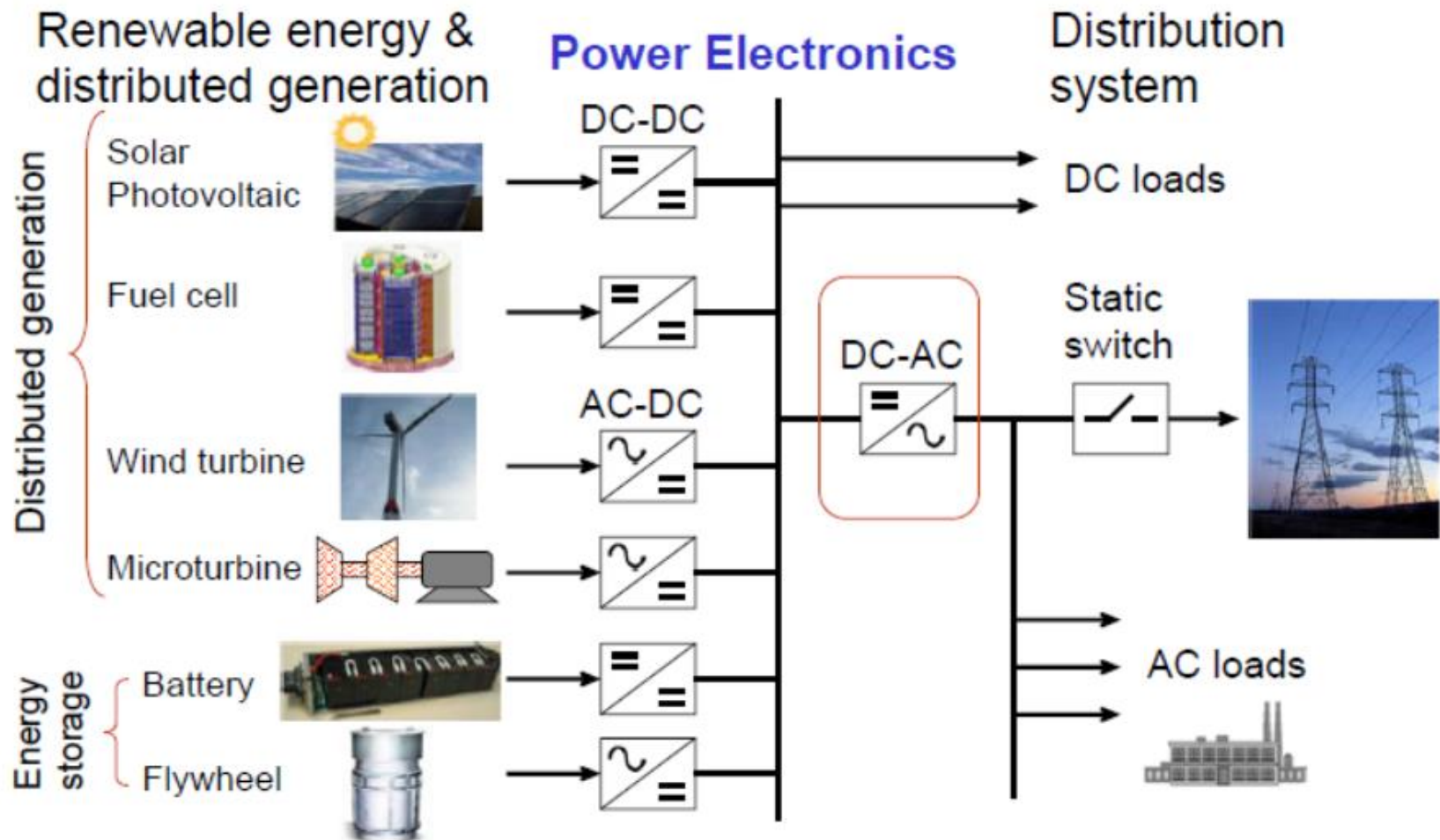


# APPLICATION OF FINITE CONTROL SET MPC IN THREE PHASE INVERTER

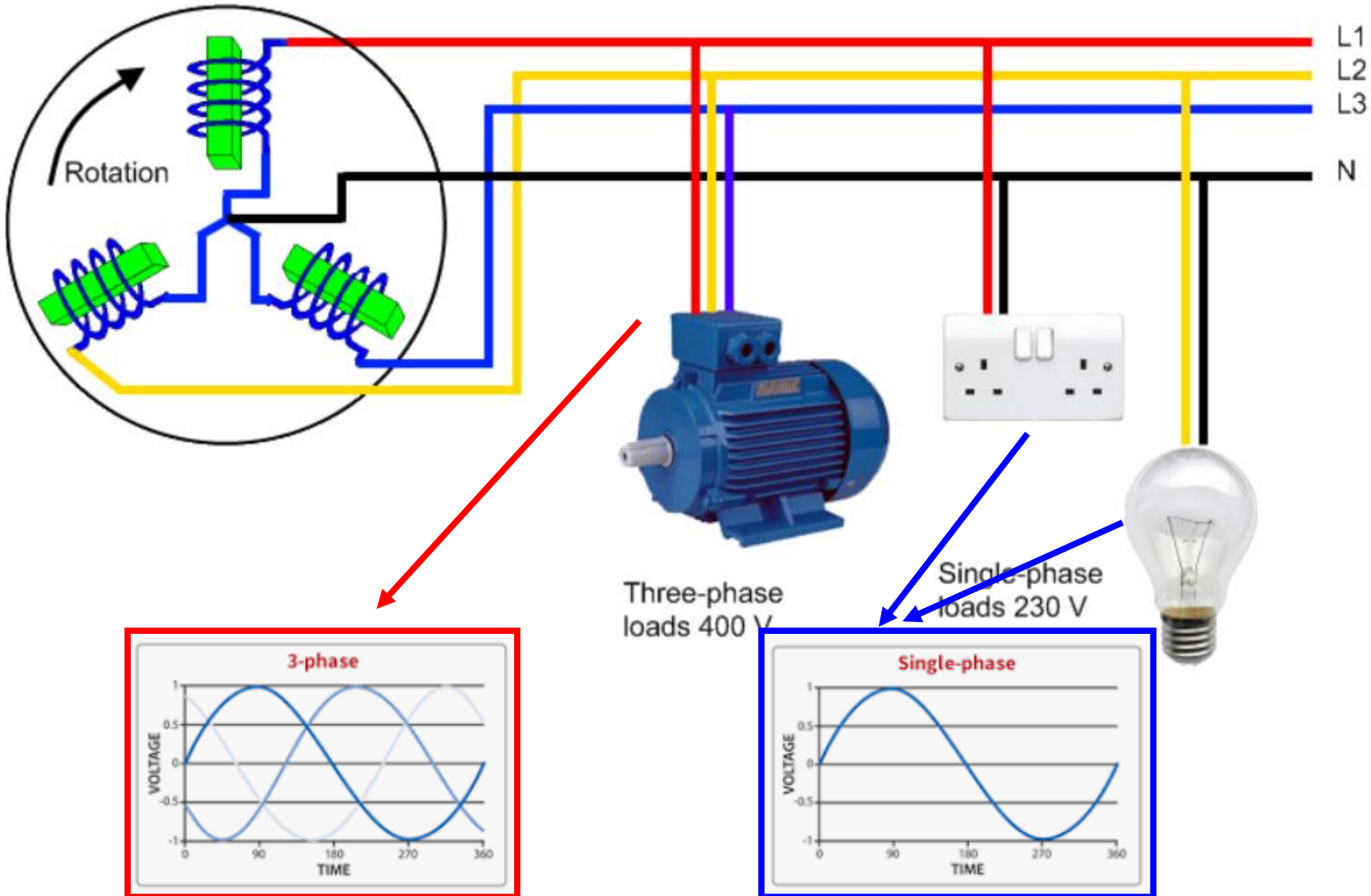
[07/11/2019]

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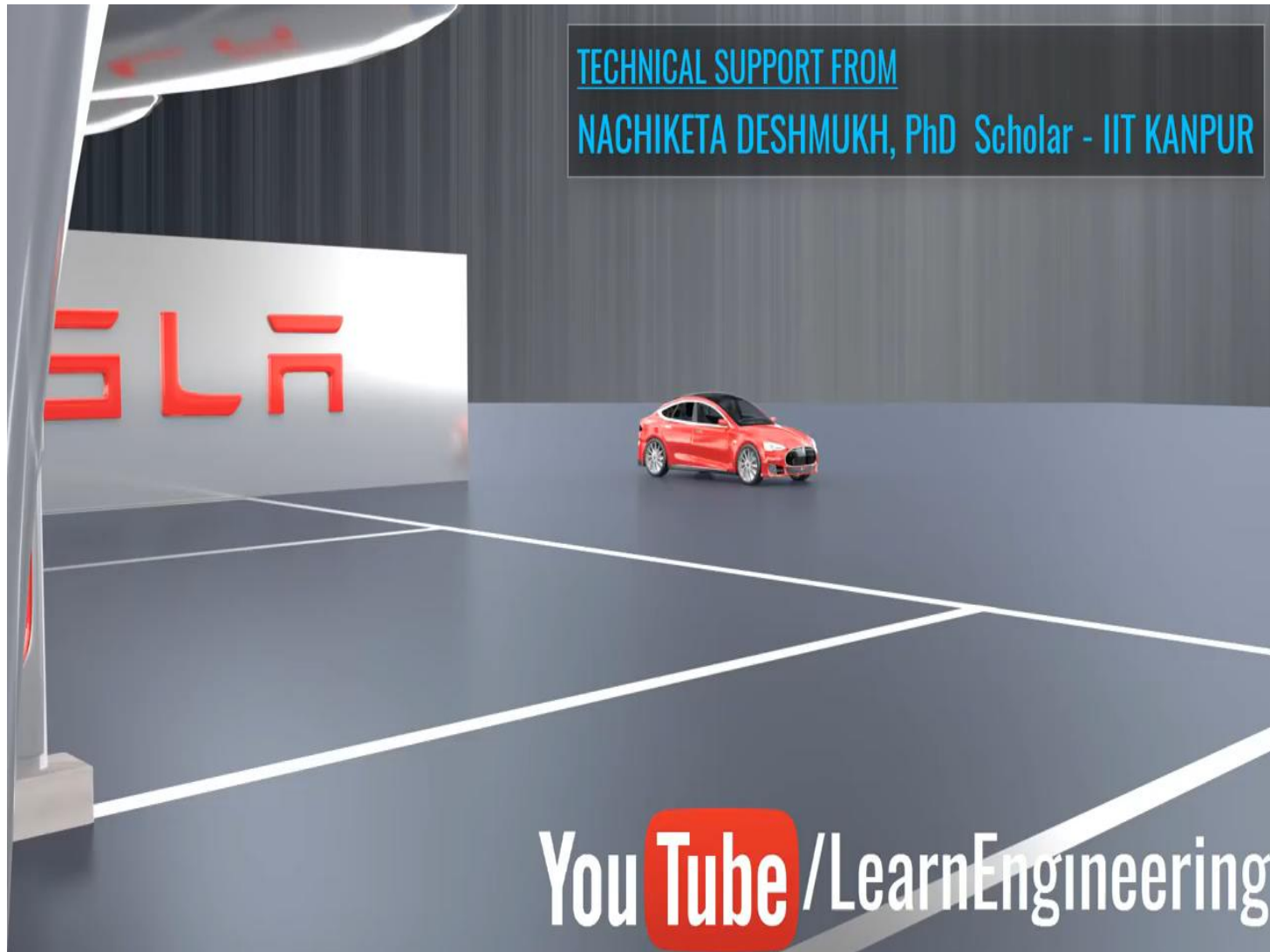
## Inverter as the Core Technology for Renewable Energy and Distributed Generation



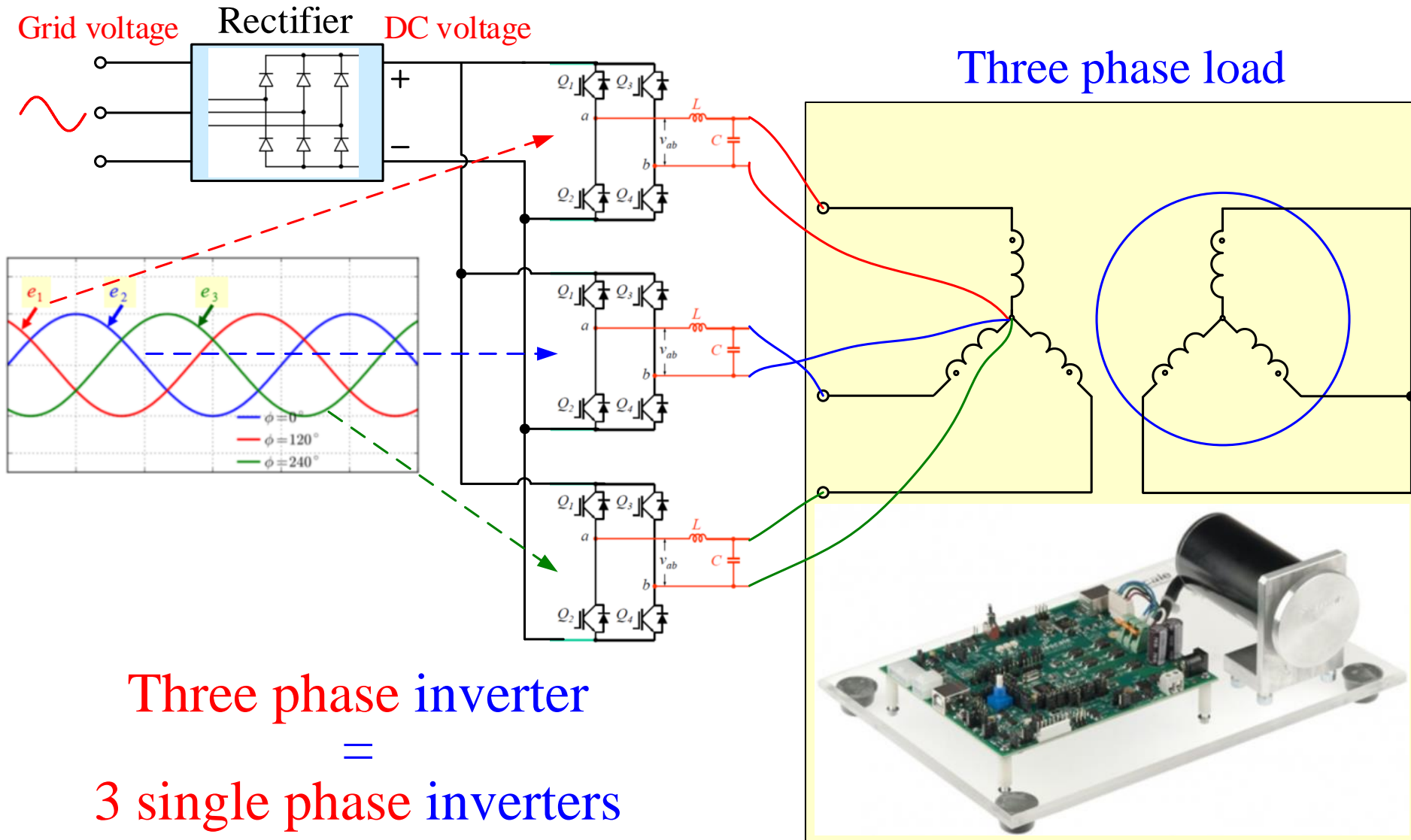
# Single-phase inverter and three-phase inverter



# Single phase inverter's working principle

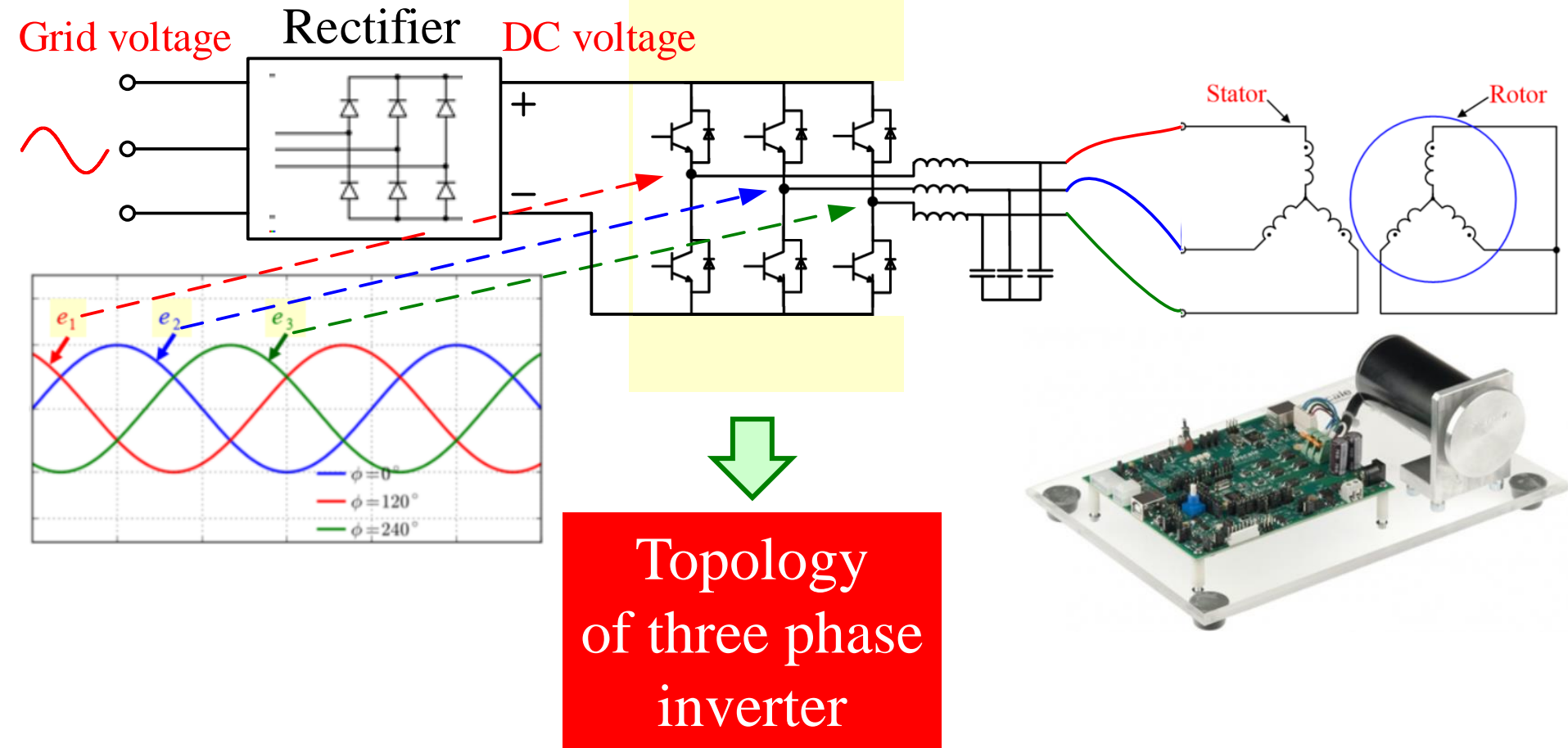


# Three phase inverter = 3 single phase inverters

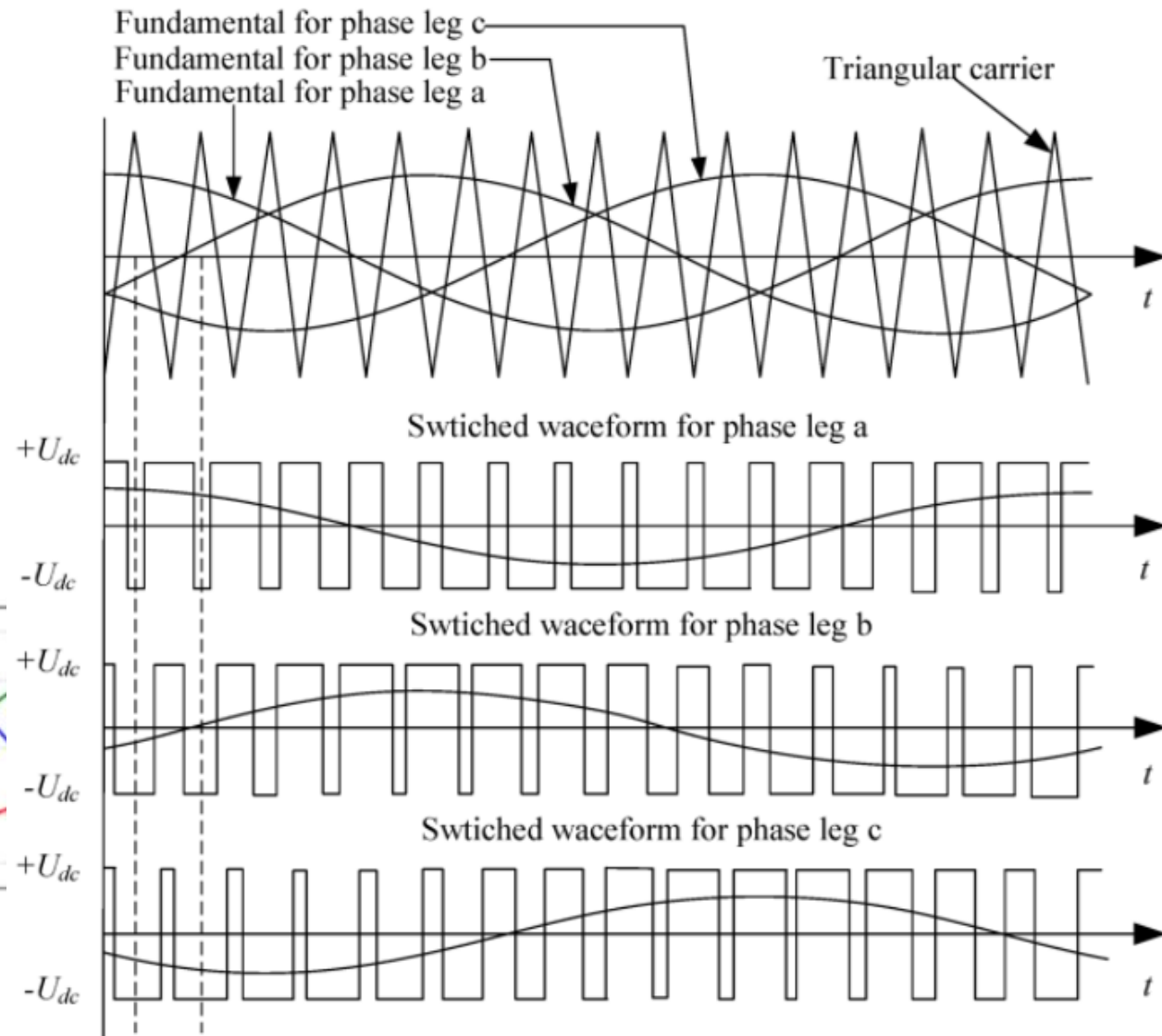
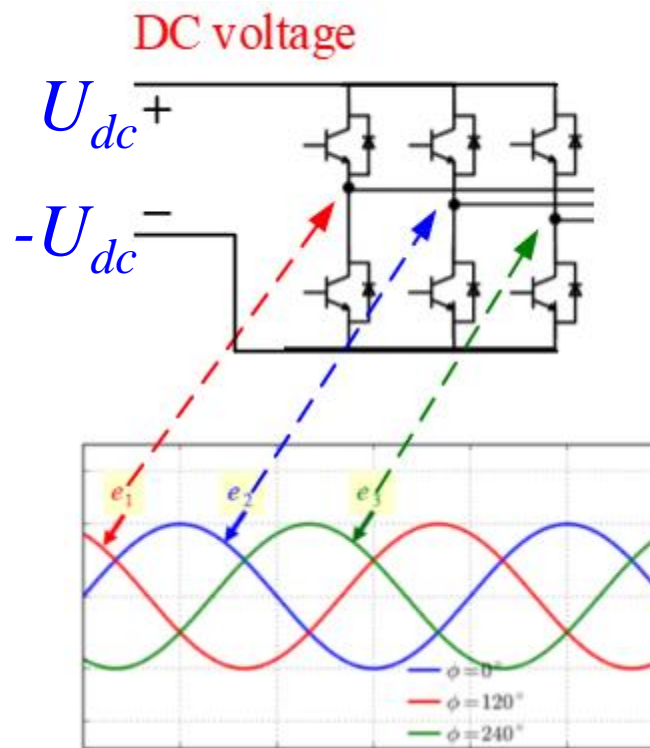




# Topology of three phase inverter

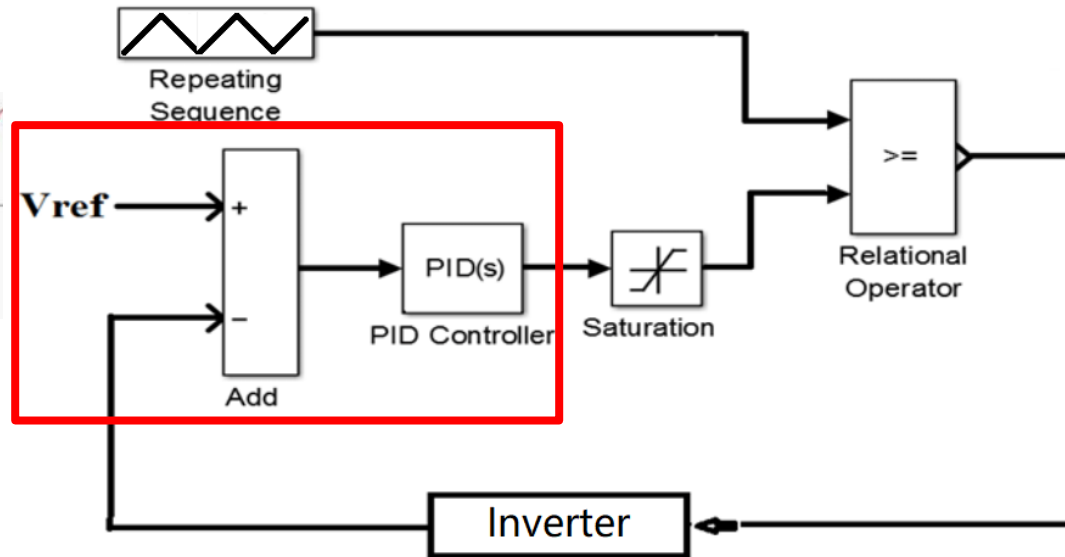
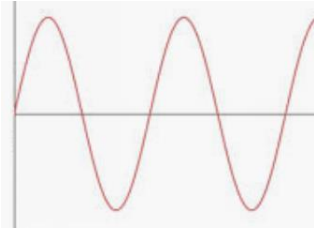


# Three phase inverter SPWM control

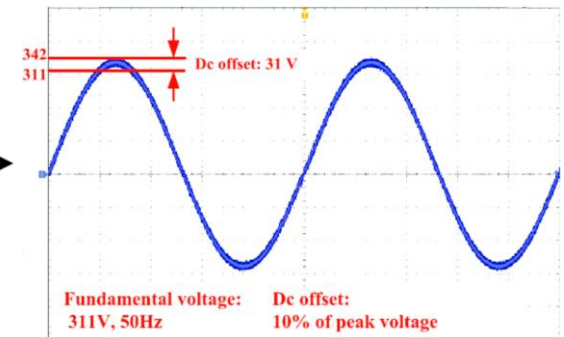
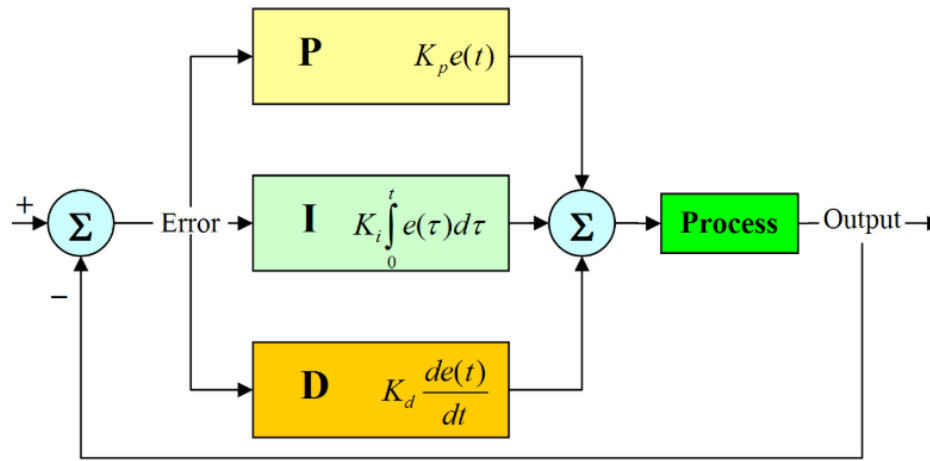




# Why SPWM control needs further improvement?



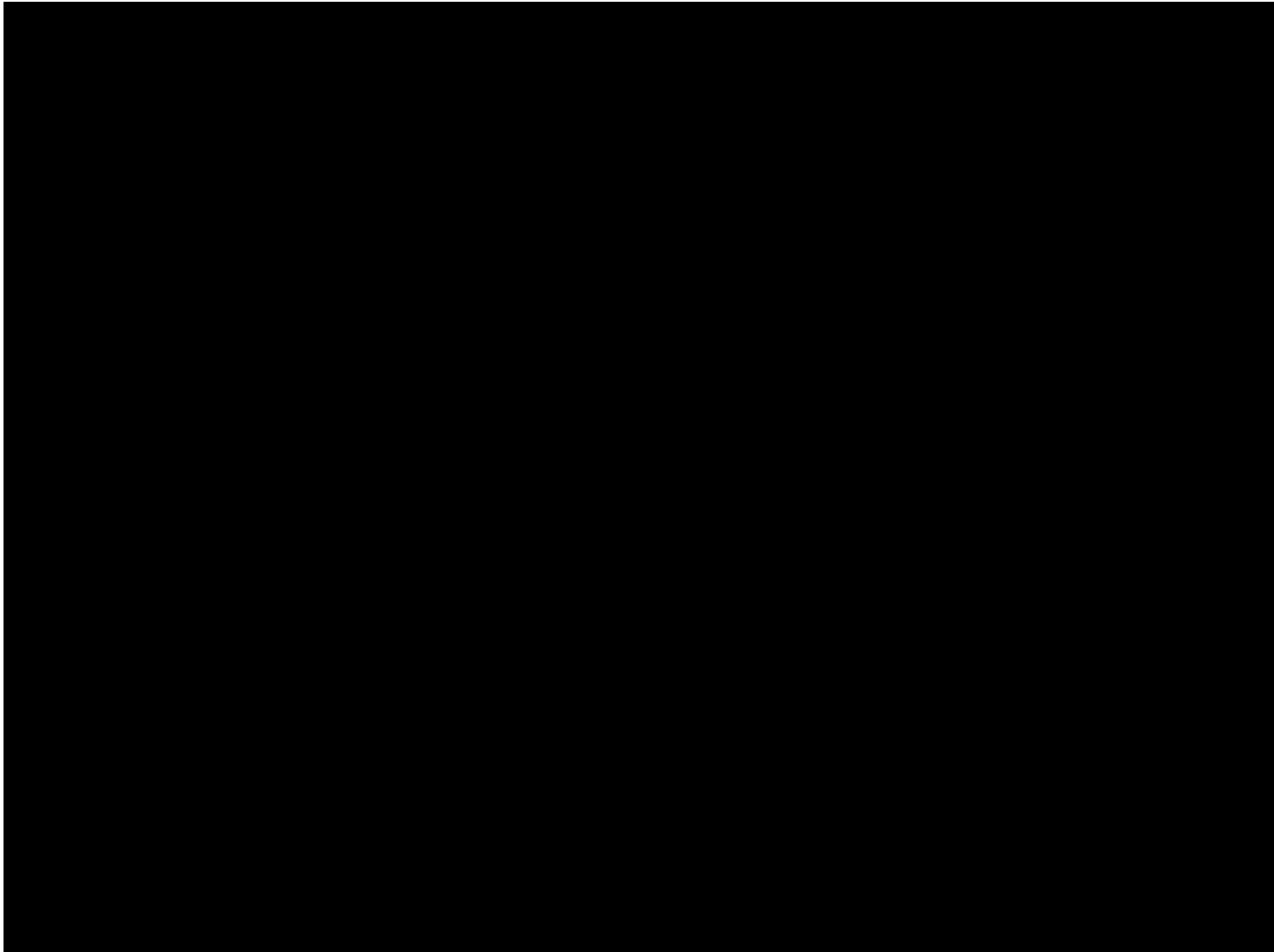
PID  
control is  
optimal to  
DC, but  
not AC





LET US TO TRY MPC CONTROLLER 😊

# Application of the three-phase inverter



# APPLICATION OF FINITE CONTROL SET MPC IN THREE PHASE INVERTER

[07/11/2019]

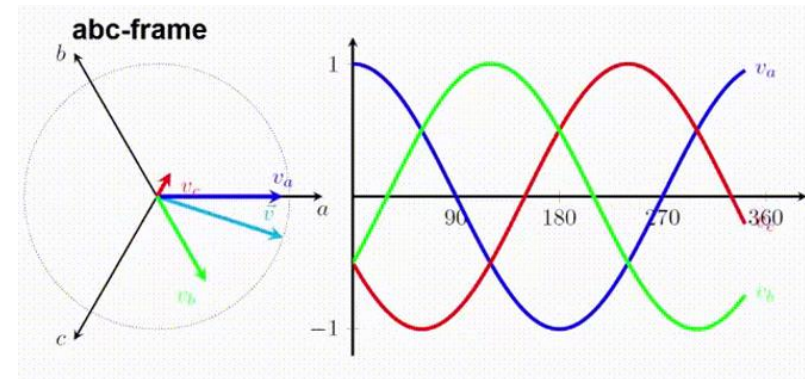
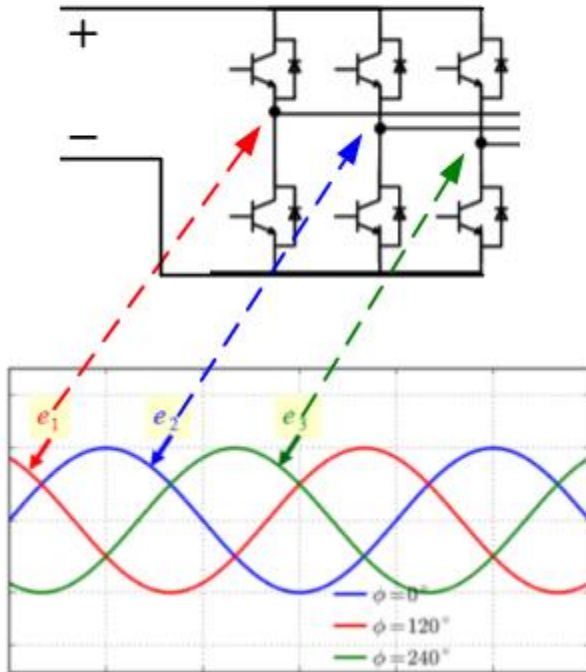
- Basic knowledge of power inverters
- Finite Control Set (FCS) MPC for power inverters
- Improved FCS-MPC for power inverters

## FCS-MPC FOR POWER INVERTERS - PRELIMINARY



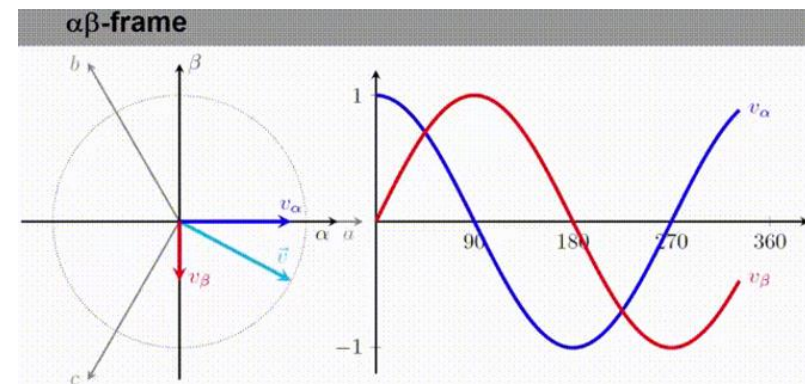
# 3 phase voltages can change to 2 phase voltages

DC voltage



$$\begin{bmatrix} u_\alpha \\ u_\beta \end{bmatrix} = \frac{2}{3} \begin{bmatrix} 1 & -\frac{1}{2} & -\frac{1}{2} \\ 0 & \frac{\sqrt{3}}{2} & -\frac{\sqrt{3}}{2} \end{bmatrix} \begin{bmatrix} u_a \\ u_b \\ u_c \end{bmatrix}$$

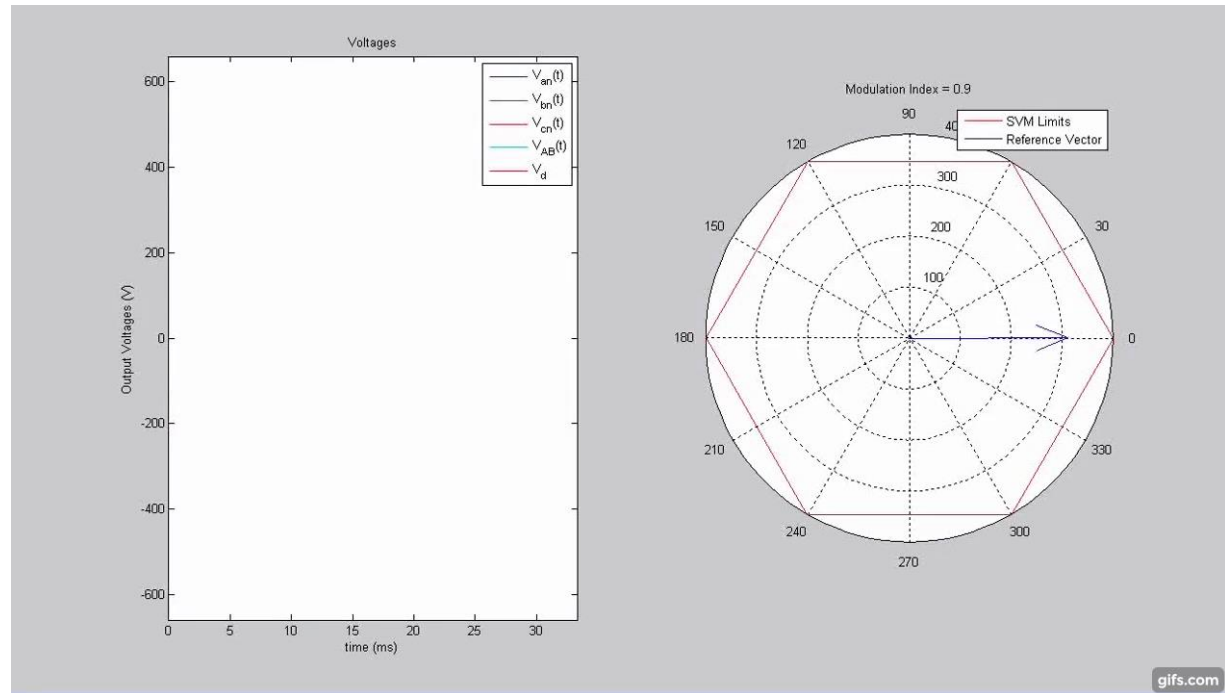
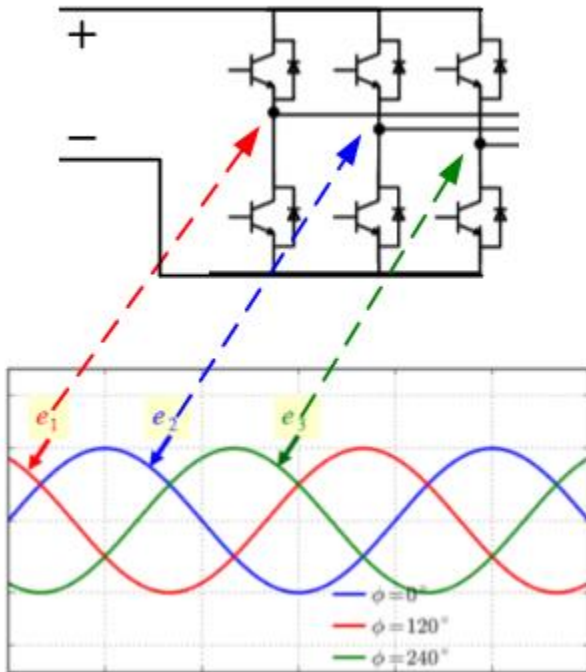
$$\begin{bmatrix} u_a \\ u_b \\ u_c \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -\frac{1}{2} & \frac{\sqrt{3}}{2} \\ \frac{1}{2} & -\frac{\sqrt{3}}{2} \end{bmatrix} \begin{bmatrix} u_\alpha \\ u_\beta \end{bmatrix}$$



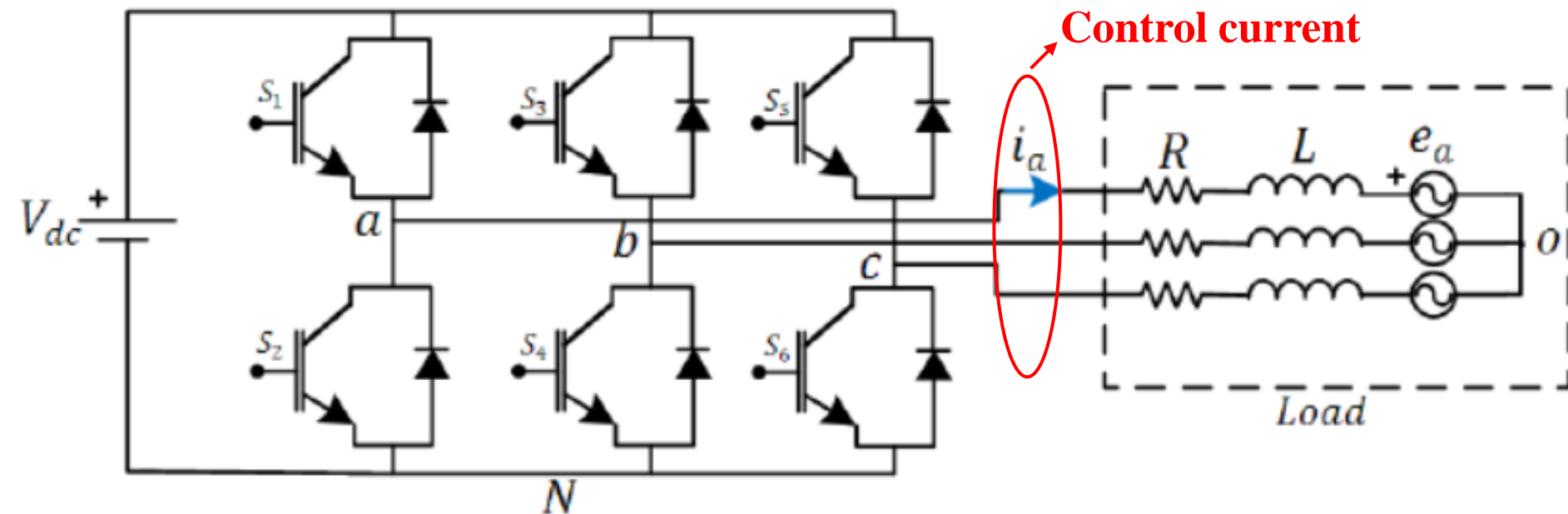


# 3 phase voltages can change to 1 voltage vector

DC voltage

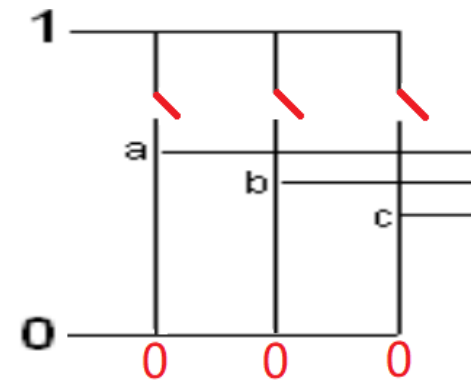


$$\vec{v} = \vec{u}_a + \vec{u}_b + \vec{u}_c = \vec{u}_\alpha + \vec{u}_\beta$$

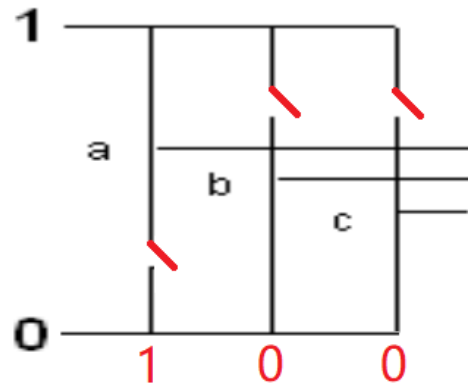


- $S_1$  &  $S_2, S_3$  &  $S_4, S_5$  &  $S_6$
- $S_{1/2/3/4/5/6}$ : 1 (Turn on); 0 (Turn off)
- Switching leg: 1 ( upper switch turn on); 0 (Upper switch turn off)

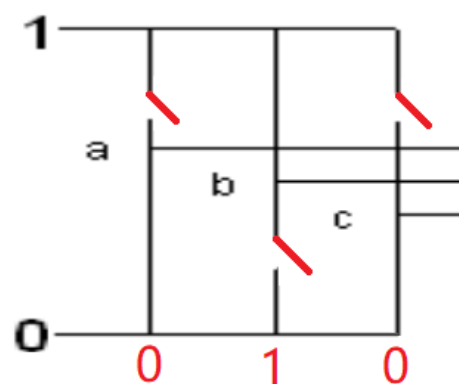
# 8 switching states of power inverters



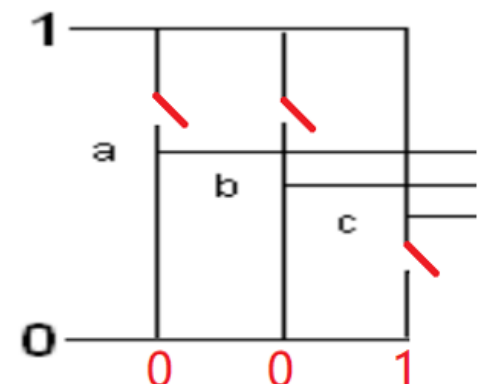
State 1



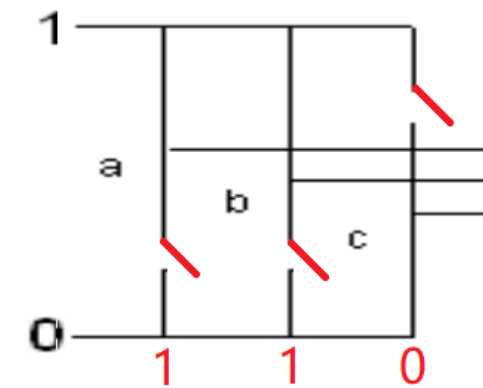
State 2



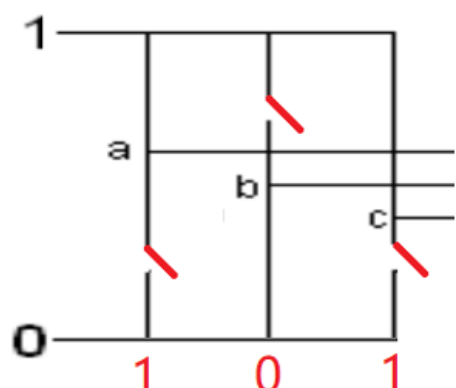
State 3



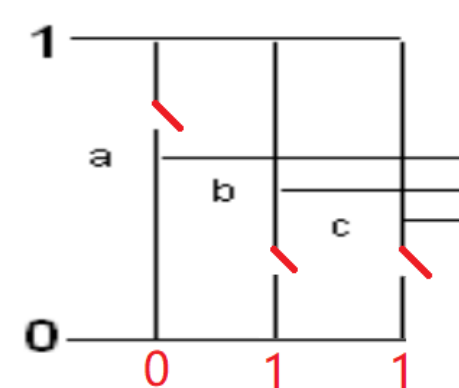
State 4



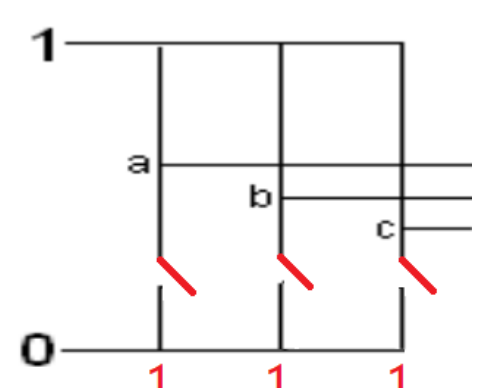
State 5



State 6



State 7

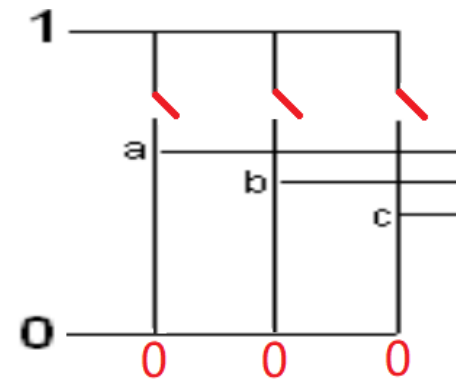
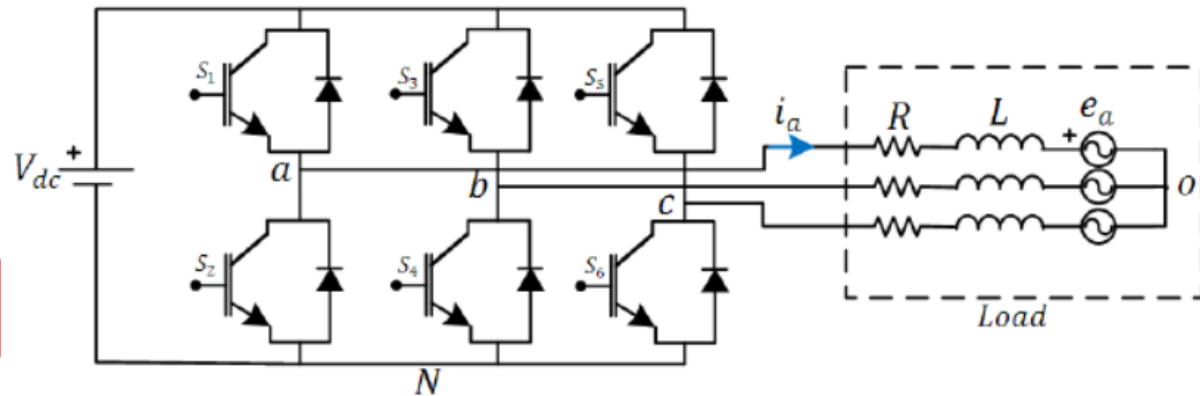


State 8



# Switch state 1 of power inverters

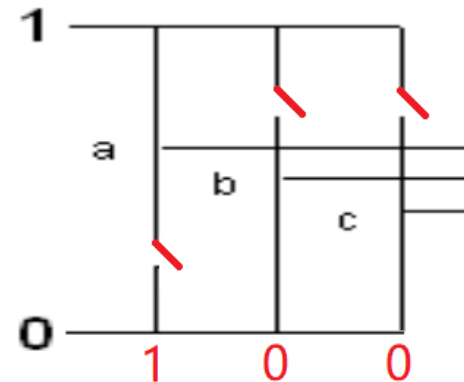
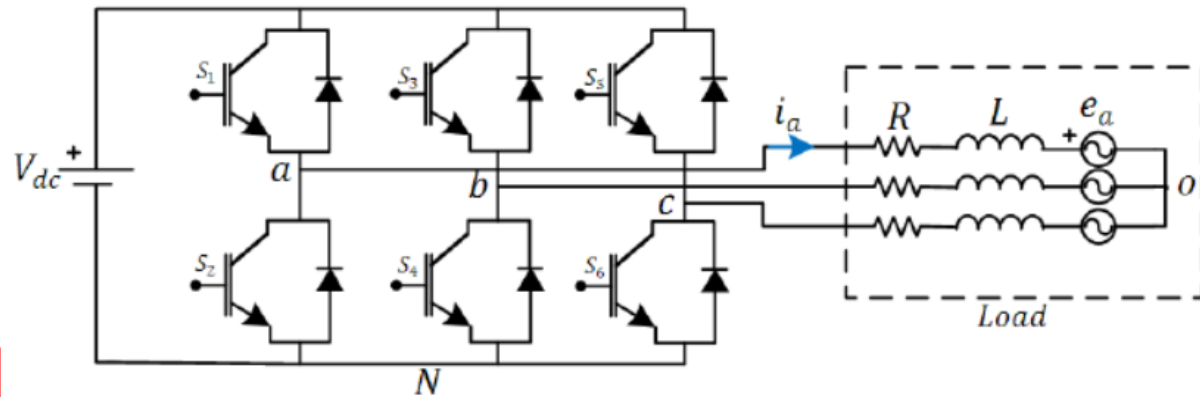
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 1

# Switch state 2 of power inverters

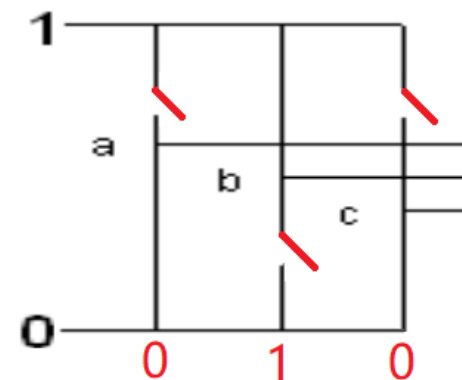
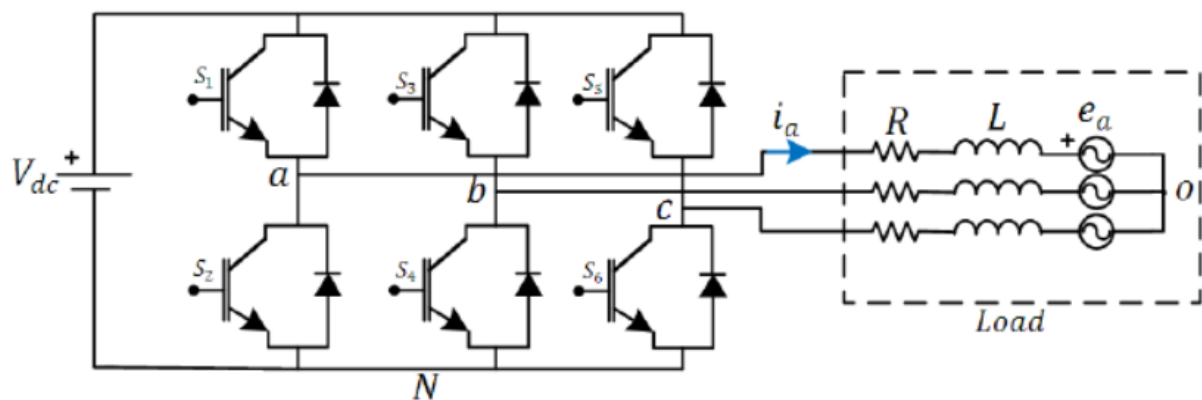
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 2

# Switch state 3 of power inverters

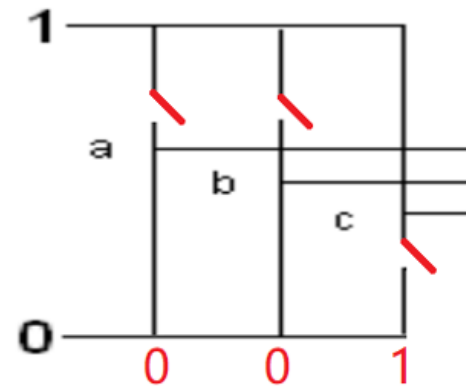
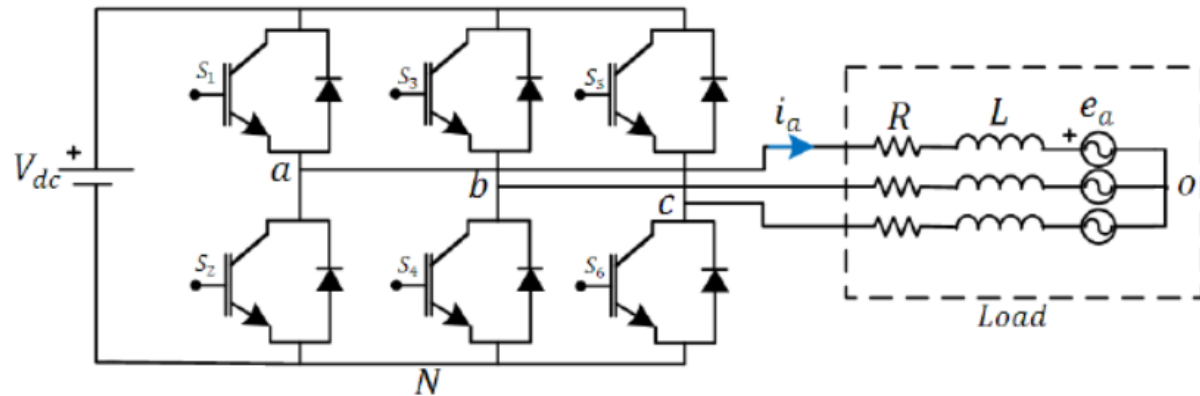
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 3

# Switch state 4 of power inverters

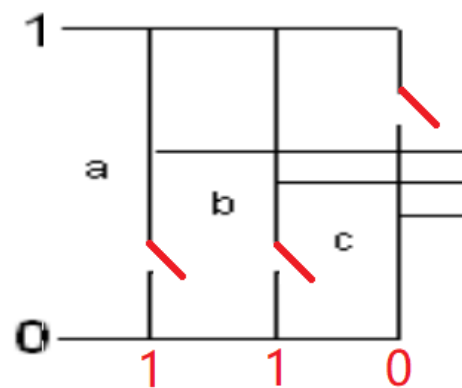
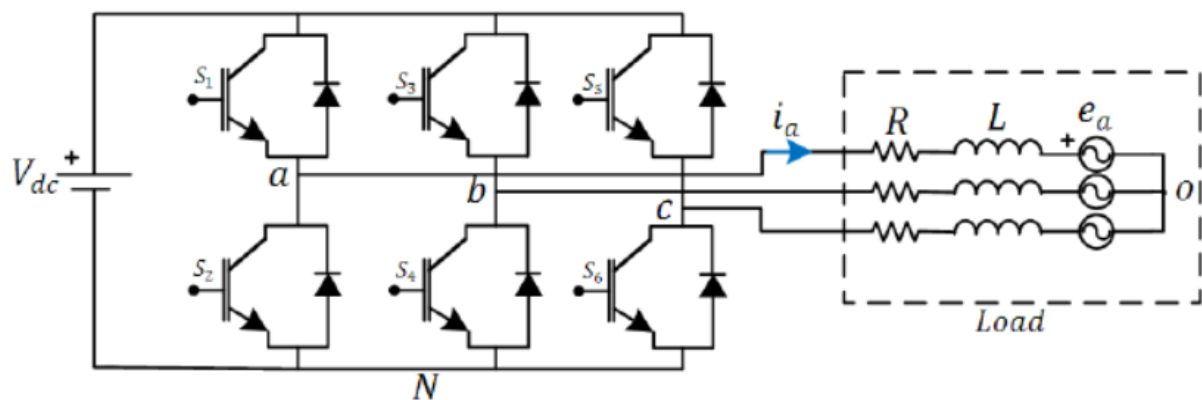
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 4

# Switch state 5 of power inverters

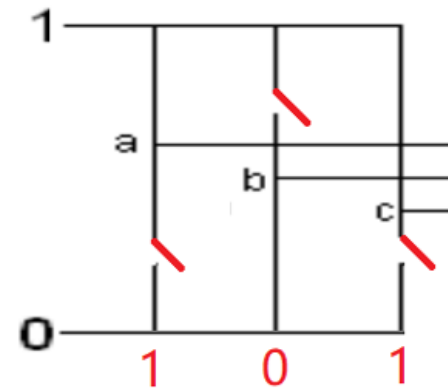
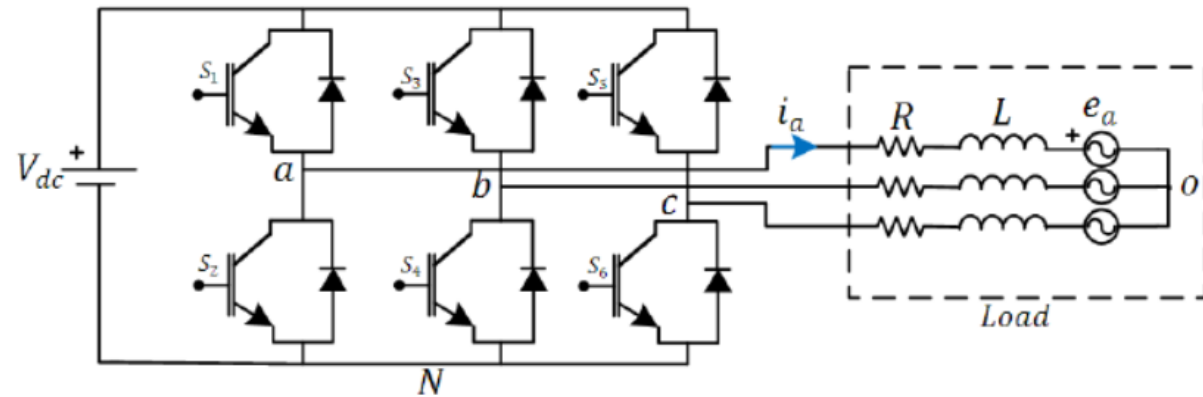
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 5

# Switch state 6 of power inverters

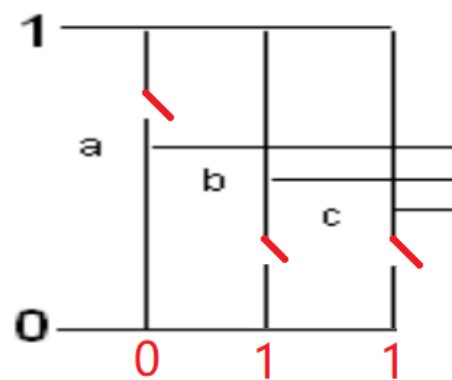
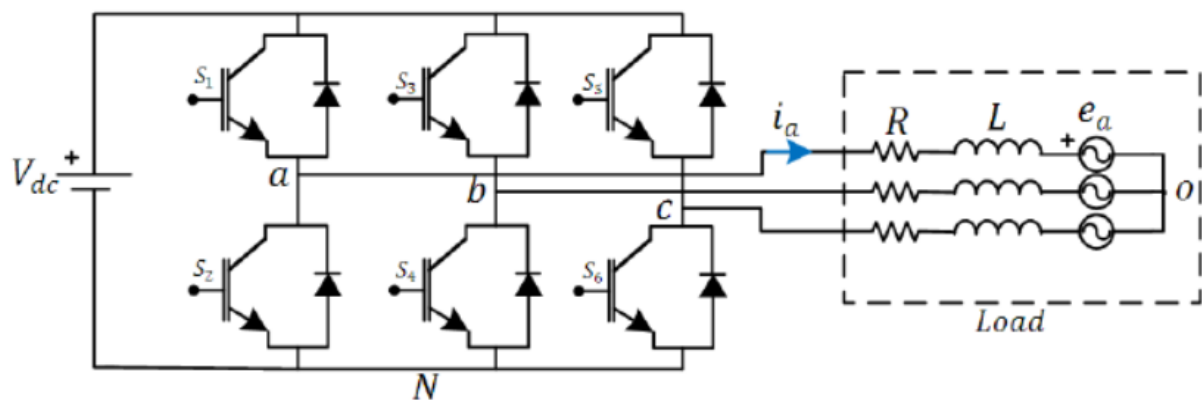
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 6

# Switch state 7 of power inverters

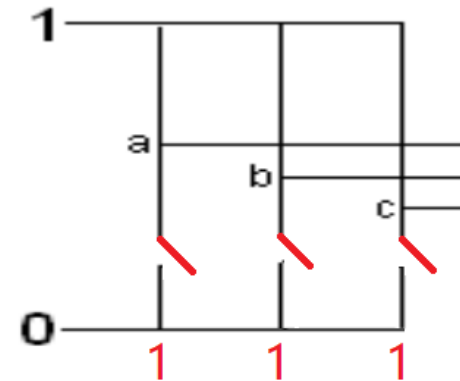
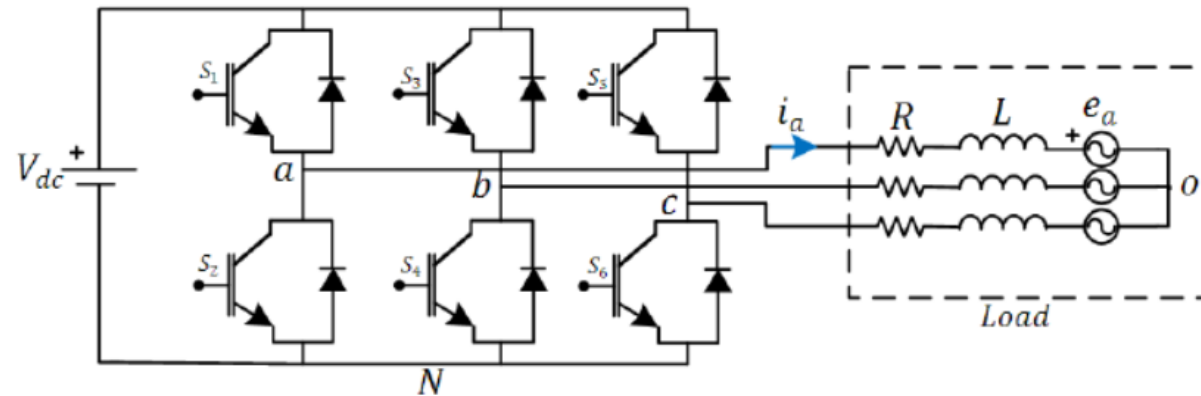
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 7

# Switch state 8 of power inverters

$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0



State 8



# Summary: 8 switch state of power inverters

State 1

State 2

State 3

State 4

State 5

State 6

State 7

State 8

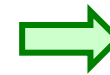
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0

## FCS-MPC FOR POWER INVERTERS - OVERVIEW

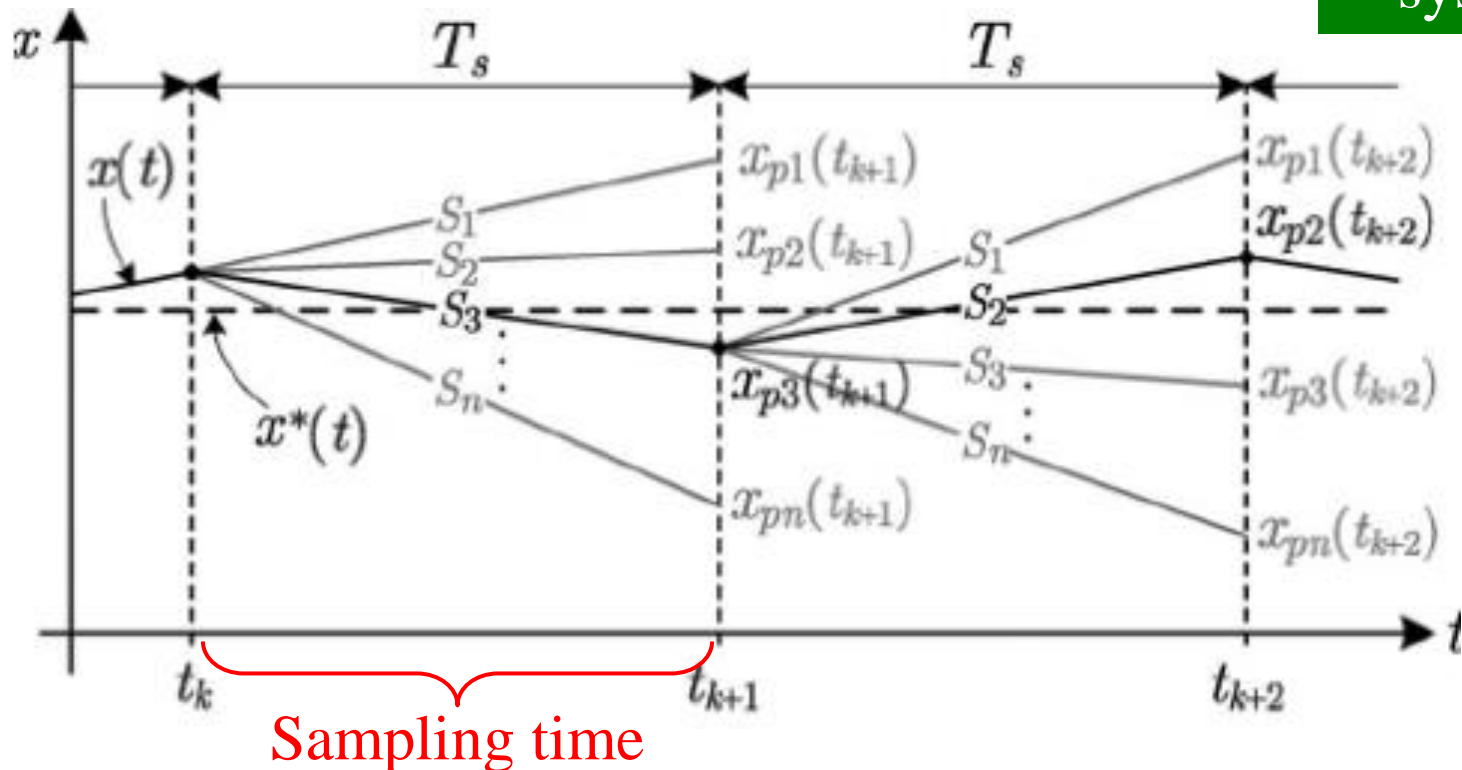


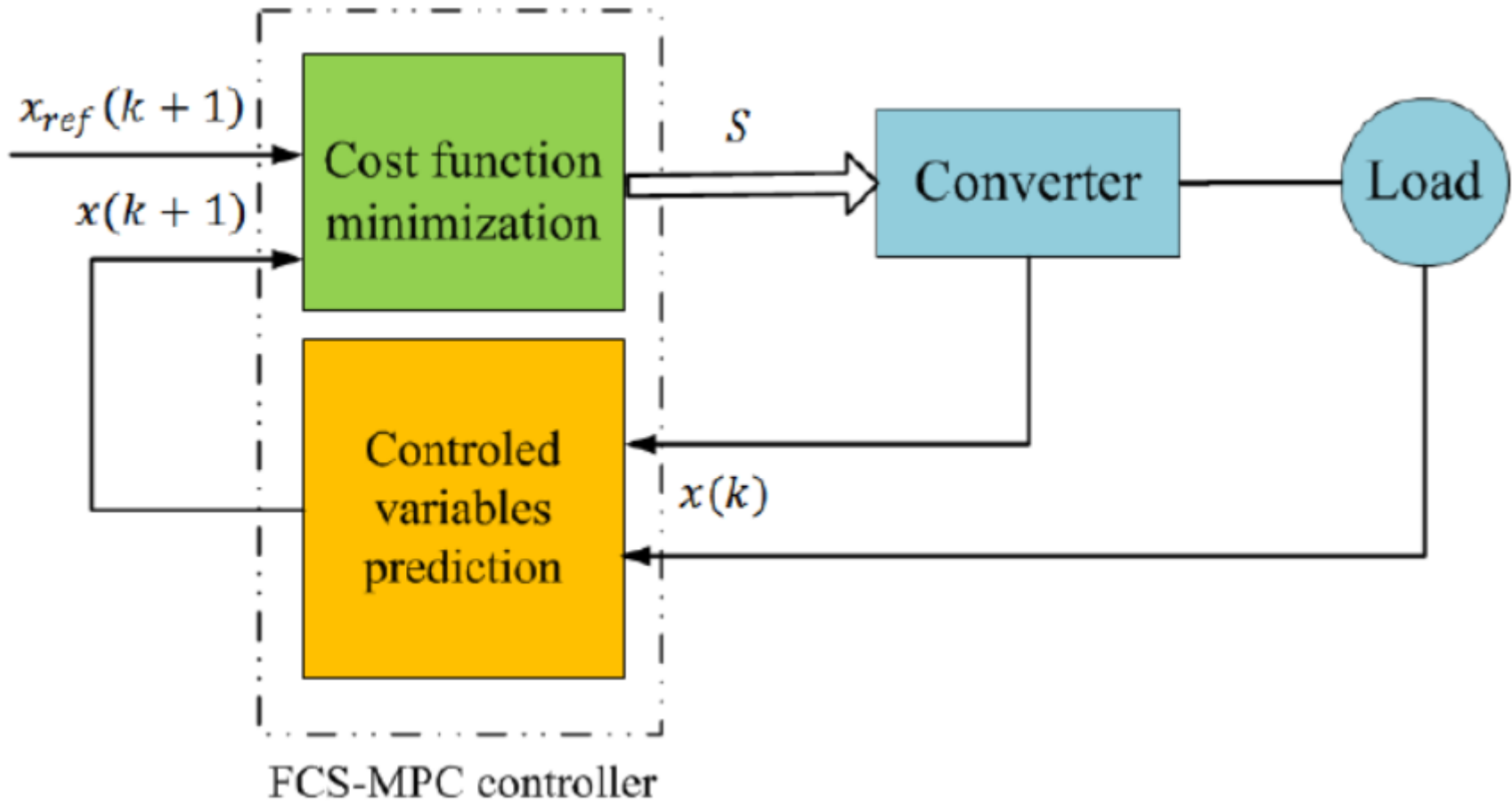
➤ Finite Control Set (FCS) MPC can reduce processing time.

- Finite number of switching states in a inverter
  - Prediction limited only to these states.
  - Select state via cost function minimization.

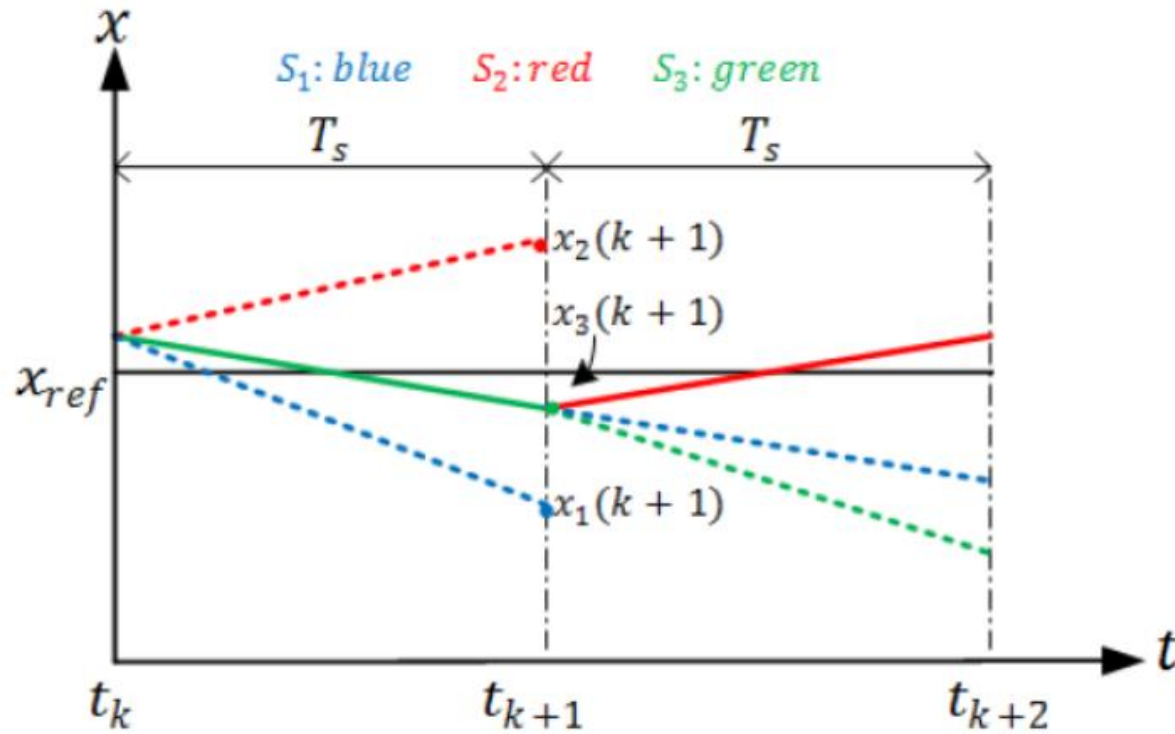


Suitable  
for discrete  
system





- $x(k)$ : Controlled variable in current state;
- $x(k+1)$ : Controlled variable in next sampling time state.



**Primary:** Identifying all the possible switching states.

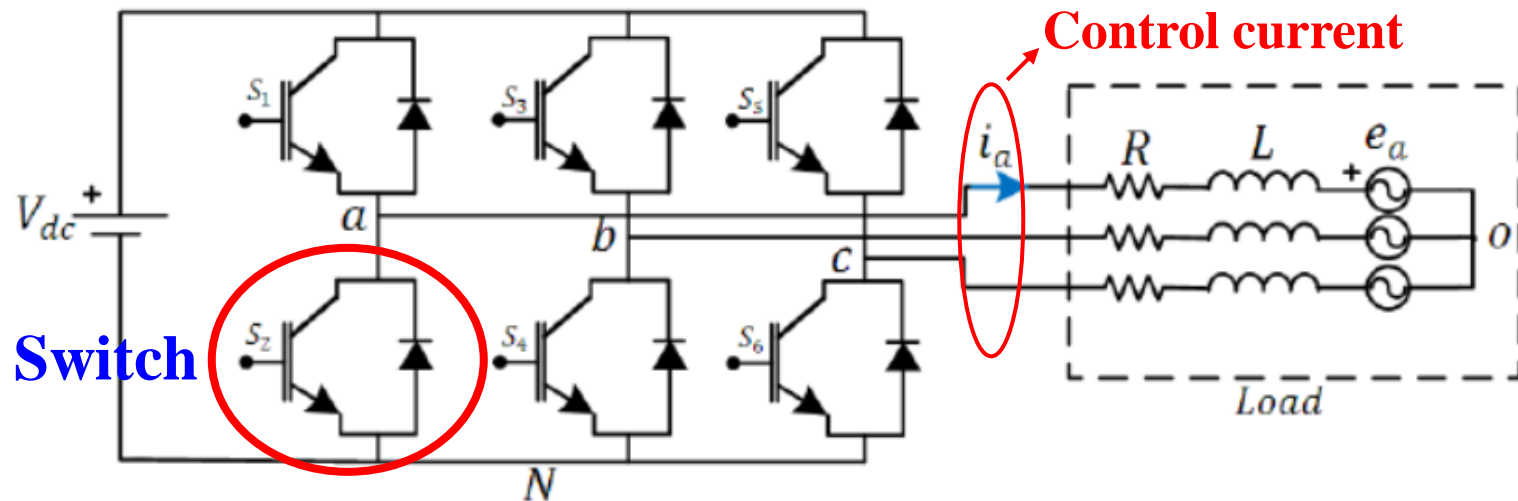
**Step 1:** Obtaining the discrete model

**Step 2:** Defining a cost function.

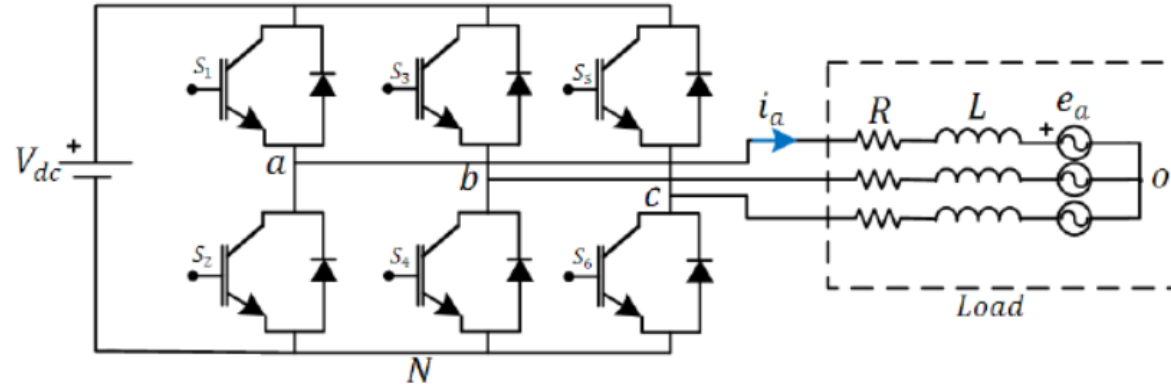
**Step 3:** Select the optimal switching states



# Inverter topology and assumptions

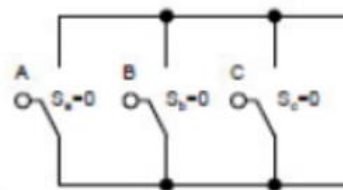


- Only two possible states for Switch: ON or OFF.
- These switching states are not acceptable:
  - Both up & down switches are ON (short circuit).
  - Both switches in each phase are OFF (no power transfer).

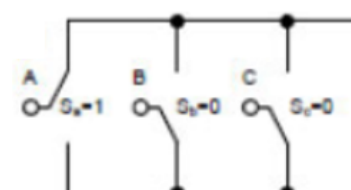


**8 switch states**

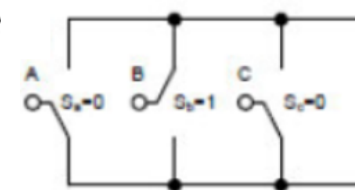
	$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
<b>State 1</b>	0	1	0	1	0	1
<b>State 2</b>	1	0	0	1	0	1
<b>State 3</b>	0	1	1	0	0	1
<b>State 4</b>	0	1	0	1	1	0
<b>State 5</b>	1	0	1	0	0	1
<b>State 6</b>	1	0	0	1	1	0
<b>State 7</b>	0	1	1	0	1	0
<b>State 8</b>	1	0	1	0	1	0



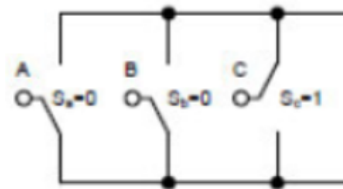
**State 1**



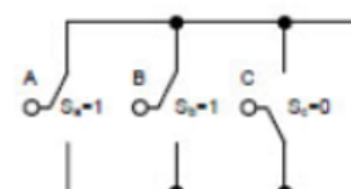
**State 2**



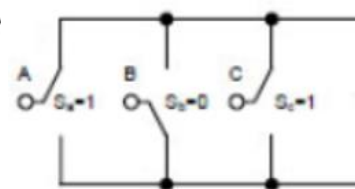
**State 3**



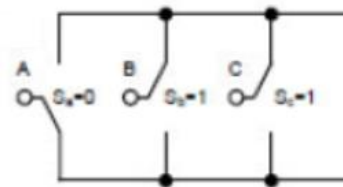
**State 4**



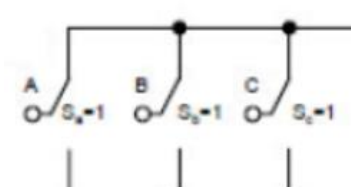
**State 5**



**State 6**



**State 7**

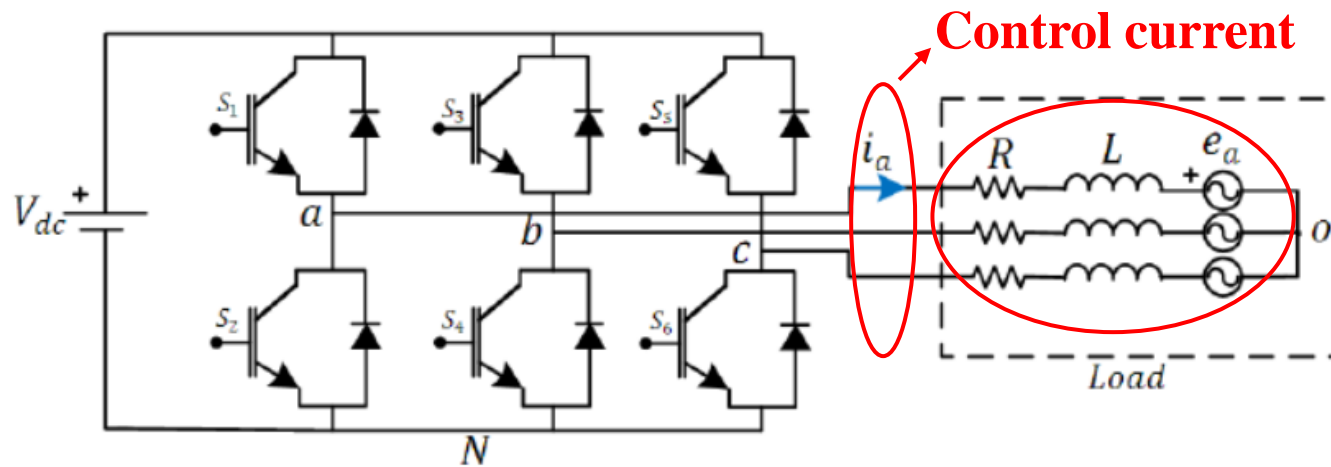


**State 8**

## FCS-MPC FOR POWER INVERTERS

### - STEP 1: MODEL





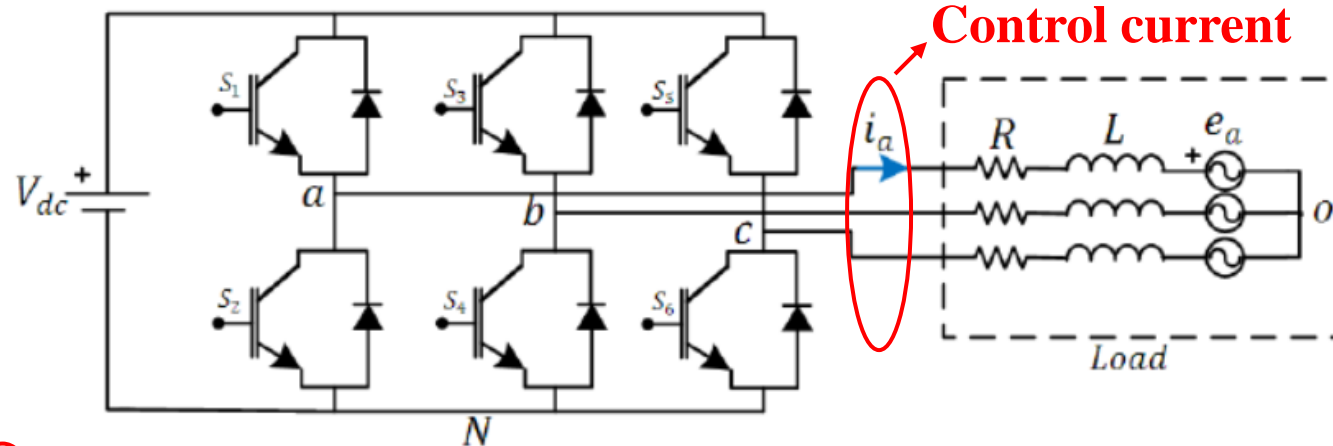
$$i = \frac{2}{3} (i_a + a i_b + a^2 i_c)$$

$$v = L \frac{di}{dt} + Ri + e \rightarrow e = \frac{2}{3} (e_a + a e_b + a^2 e_c) \rightarrow a = e^{j\frac{2\pi}{3}} = -\frac{1}{2} + j\frac{\sqrt{3}}{2}$$

$$v = \frac{2}{3} (v_{aN} + a v_{bN} + a^2 v_{cN})$$



# Inverter discrete model1: Forward Euler method



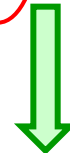
$$\frac{dx}{dt} = \frac{x(k+1) - x(k)}{T_s}$$



$$x(k+1) = x(k) + T_s f(x(k), u(k))$$

Forward Euler method

$$v = L \frac{di}{dt} + Ri + e$$

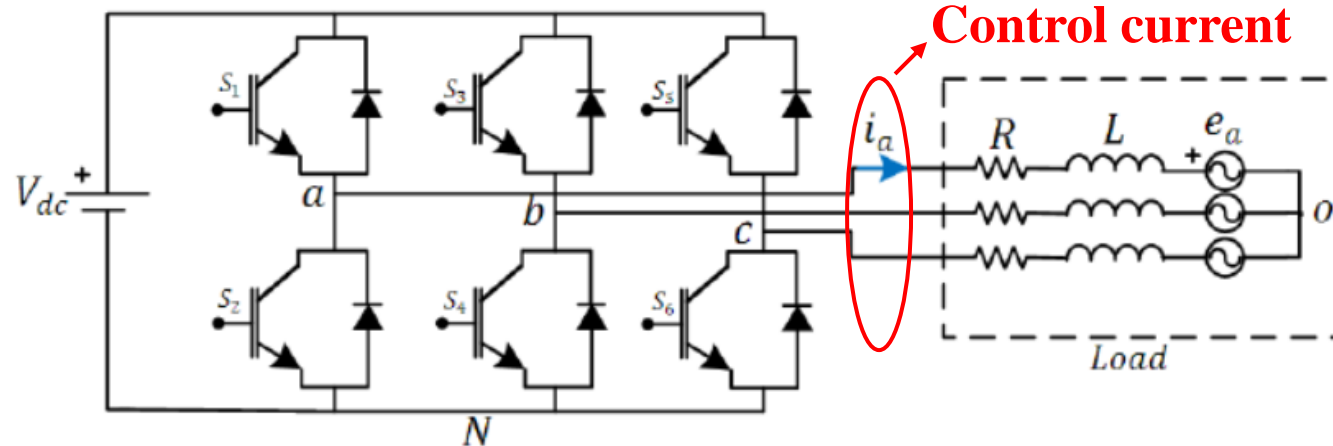


Forward Euler method

$$i(k+1) = \left(1 - \frac{T_s R}{L}\right) i(k) + \frac{T_s}{L} (v(k) - e(k))$$



# Inverter discrete model2: Backward Euler method



$$\frac{dx}{dt} = \frac{x(k+1) - x(k)}{T_s} \quad \xrightarrow{\text{Backward Euler method}} \quad \underline{x(k+1) = x(k) + T_s f(x(k+1), u(k+1))}$$

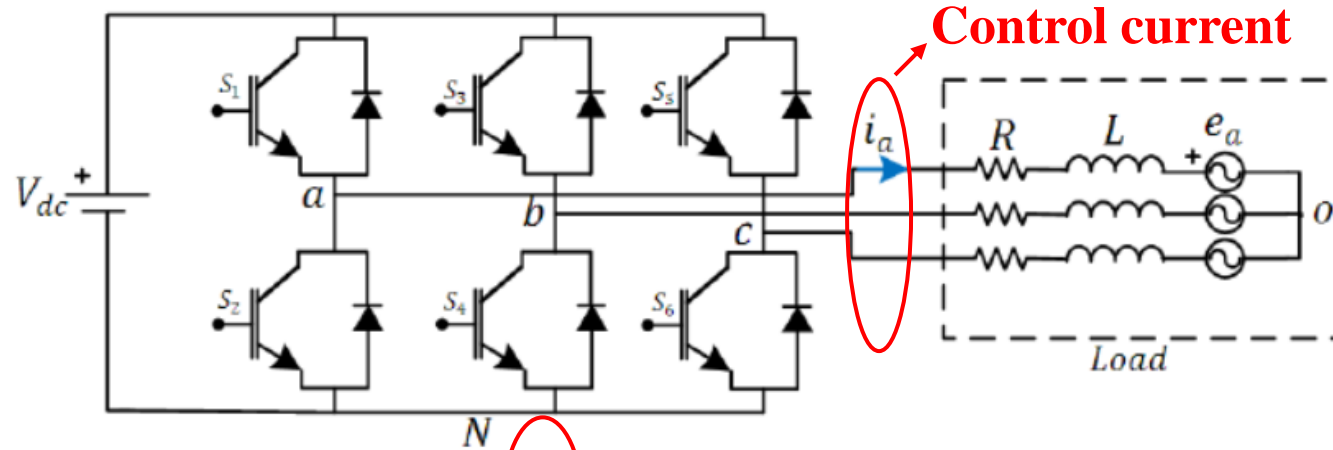
Sampling time

$$v = L \frac{di}{dt} + Ri + e$$

$$\underline{i(k+1)} = \frac{T_s}{L + RT_s} [v(k+1) - e(k+1)] + \frac{L}{L + RT_s} i(k)$$



# Inverter discrete model3: Midpoint Euler method



$$\frac{dx}{dt} = \frac{x(k+1) - x(k)}{T_s}$$



Midpoint Euler method

$$x(k+1) = x(k) + \frac{T_s}{2} [f(x(k), u(k)) + f(x(k+1), u(k+1))]$$

$$v = L \frac{di}{dt} + Ri + e$$

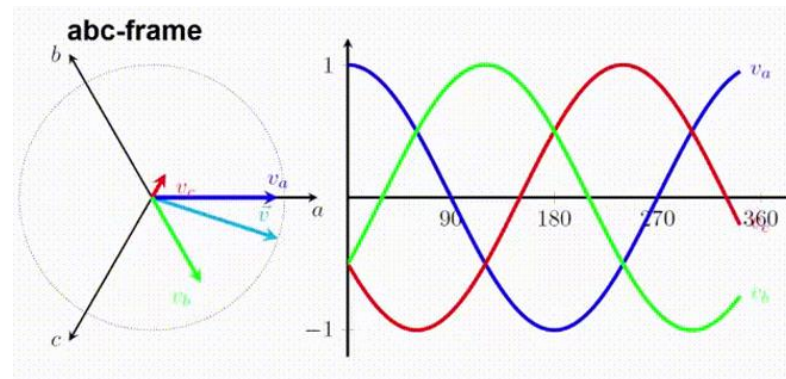
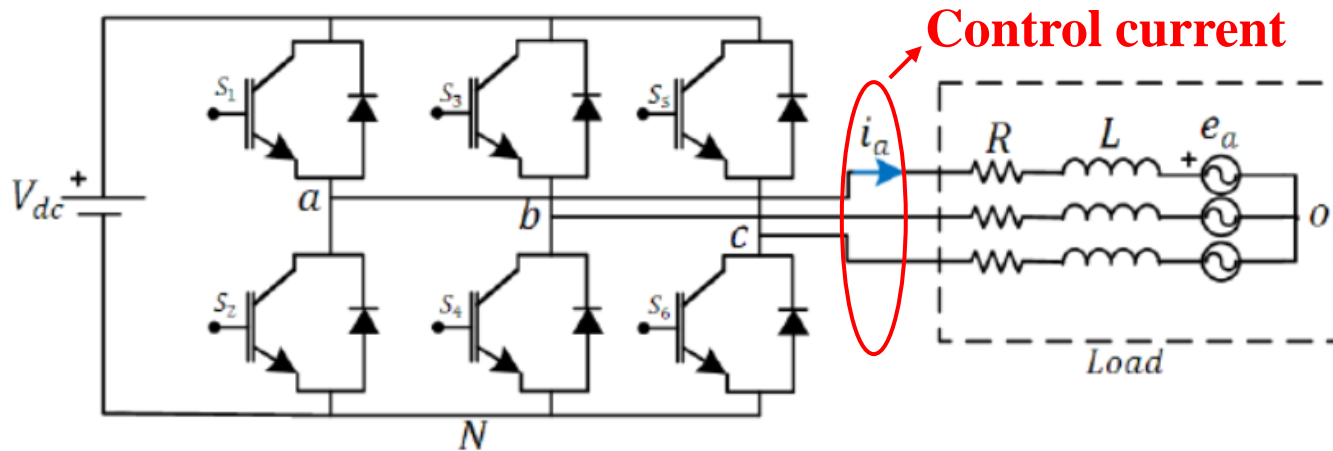


Midpoint Euler method

$$i(k+1) = \frac{2 - RT_s}{2 + RT_s} i(k) + \frac{T_s}{2 + RT_s} [v(k) + v(k+1) - e(k) - e(k+1)]$$

## FCS-MPC FOR POWER INVERTERS

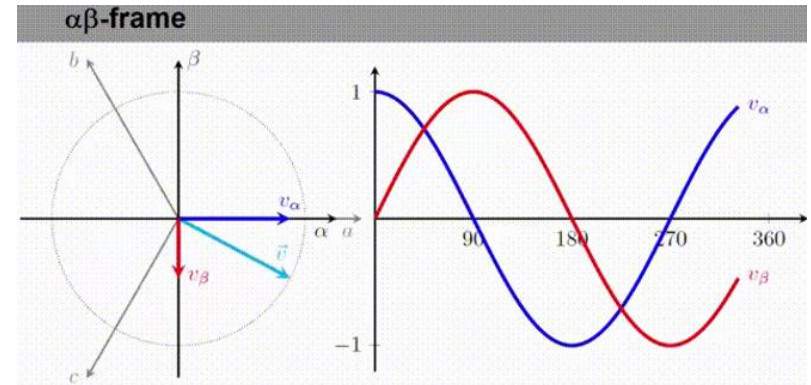
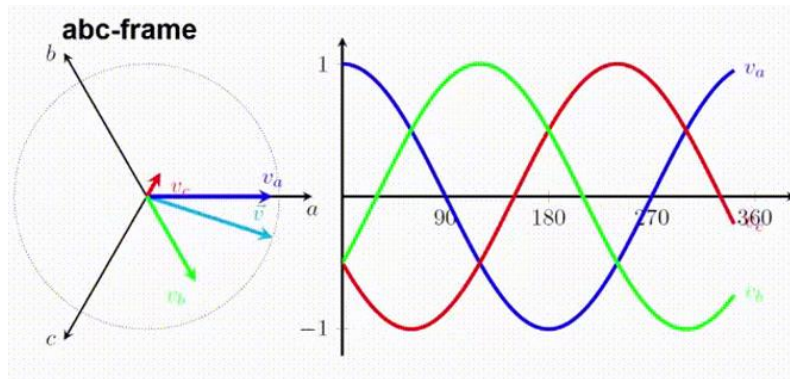
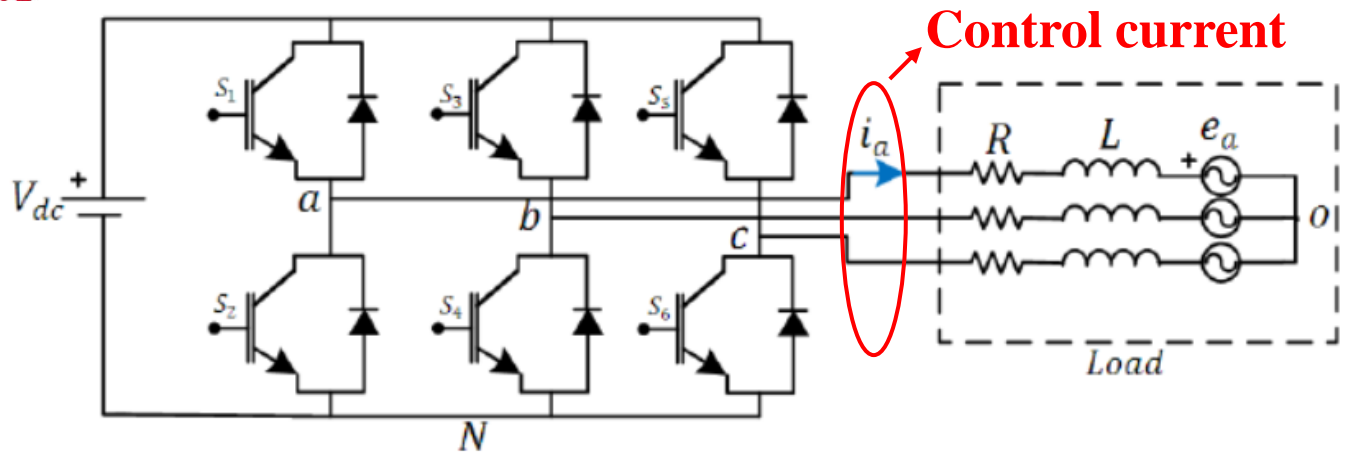
### - STEP 2: COST FUNCTION



➤ Basic cost function:  $i_{a/b/c}$  is the real current,  $i_{a/b/c}^*$  is the reference

$$J = \left| i_a^*(k+1) - i_a(k+1) \right| + \left| i_b^*(k+1) - i_b(k+1) \right| + \left| i_c^*(k+1) - i_c(k+1) \right|$$

# Improved cost function: reduce items

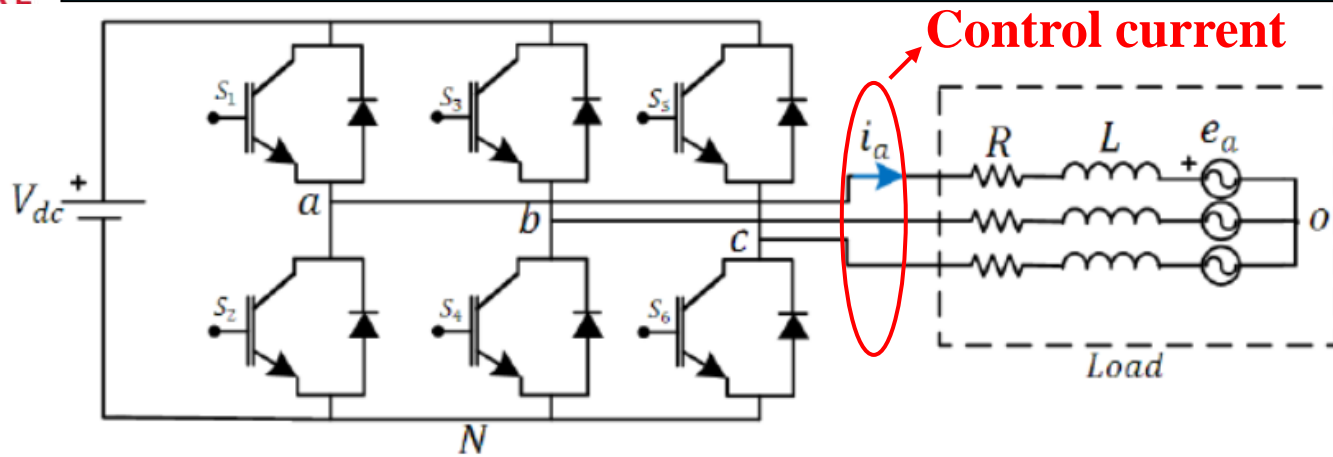


➤ Improved cost function:

Reduce from  $i_{a/b/c}$  to  $i_{\alpha/\beta}$

$$J = |i_{\alpha}^*(k+1) - i_{\alpha}(k+1)| + |i_{\beta}^*(k+1) - i_{\beta}(k+1)|$$





➤ Ideal cost function:

$$J = |i_{\alpha}^*(k+1) - i_{\alpha}(k+1)| + |i_{\beta}^*(k+1) - i_{\beta}(k+1)|$$

➤ If high sampling frequency

- Current is approximated to be constant in one step time,
- Practical cost function:

$$J = |i_{\alpha}^*(k) - i_{\alpha}(k+1)| + |i_{\beta}^*(k) - i_{\beta}(k+1)|$$



\* ADD CONSTRAINS TO THE COST  
FUNCTION OF THREE PHASE INVERTERS

- ✓ Switching frequency minimization
- ✓ Voltage and current ripple minimization
- ✓ Defining maximum allowed current and voltage

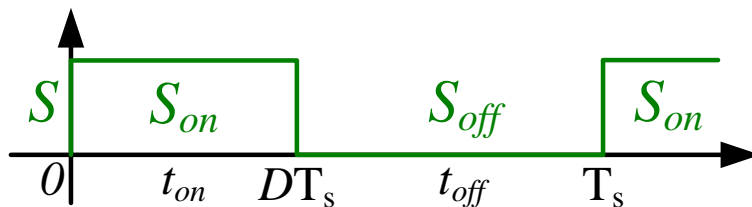


➤ Switch states amount changed at each sampling time:

Weight factor: (0~1)

$$J = \left| i_{\alpha}^*(k+1) - i_{\alpha}(k+1) \right| + \left| i_{\beta}^*(k+1) - i_{\beta}(k+1) \right| + \boxed{\lambda \cdot n}$$

Switching frequency



$$n = \sum_{i=1}^N |S_i(k+1) - S_i(k)|$$

- $S_i(k)$ : switch state  $i$  at the current state;
- $S_i(k+1)$ : switch state  $i$  at the next sampling period;

- General form of adding **voltage ripple constrain**:

$$J = \|x_{ref} - x_{prediction}\| + \lambda \cdot \|v(k+1) - v(k)\|$$



**Voltage ripple constrain**

Weight  
factor:  
(0~1)

- General form of adding **current ripple constrain**:

$$J = \|x_{ref} - x_{prediction}\| + \lambda \cdot \|i(k+1) - i(k)\|$$



**Current ripple constrain**

## Defining allowed maximum current & voltage

- Cost function considering allowed maximum current:

$$J = \|x_{ref} - x_{prediction}\| + \boxed{f_{lim}(i_{prediction})} \quad \text{Allowed max. } I$$

Weight  
factor:  
(0~1)

$$\boxed{f_{lim}(i_{prediction}) = \begin{cases} \infty & \text{if } |i_{prediction}| > I_{max} \\ 0 & \text{if } |i_{prediction}| \leq I_{max} \end{cases}}$$

- Cost function considering allowed maximum voltage:

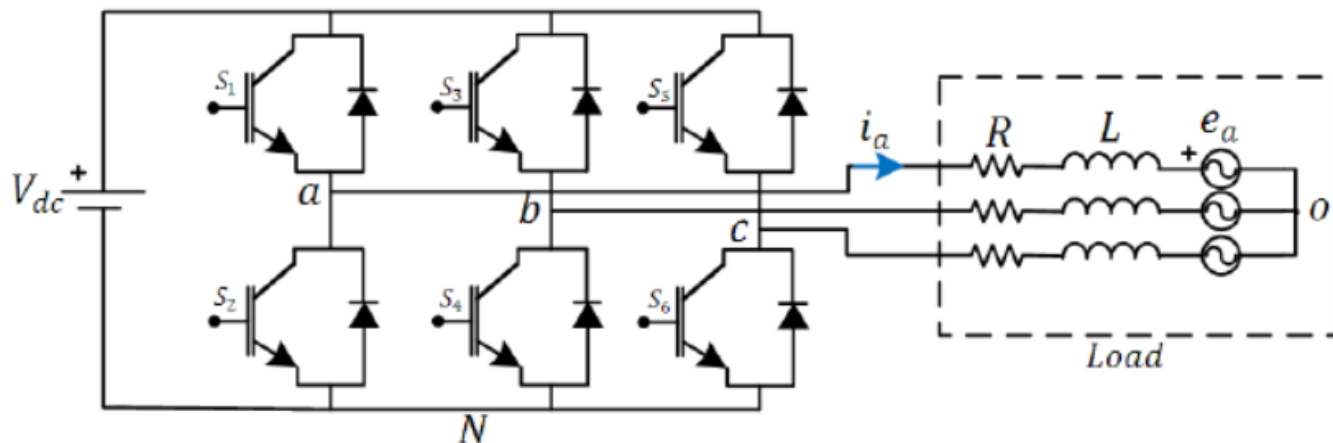
$$J = \|x_{ref} - x_{prediction}\| + \boxed{f_{lim}(v_{prediction})} \quad \text{Allowed max. } V$$

$$\boxed{f_{lim}(v_{prediction}) = \begin{cases} \infty & \text{if } |v_{prediction}| > V_{max} \\ 0 & \text{if } |v_{prediction}| \leq V_{max} \end{cases}}$$

## FCS-MPC FOR POWER INVERTERS

### - STEP 3: SELECT THE OPTIMAL SWITCH STATES

# Select the suitable switching state



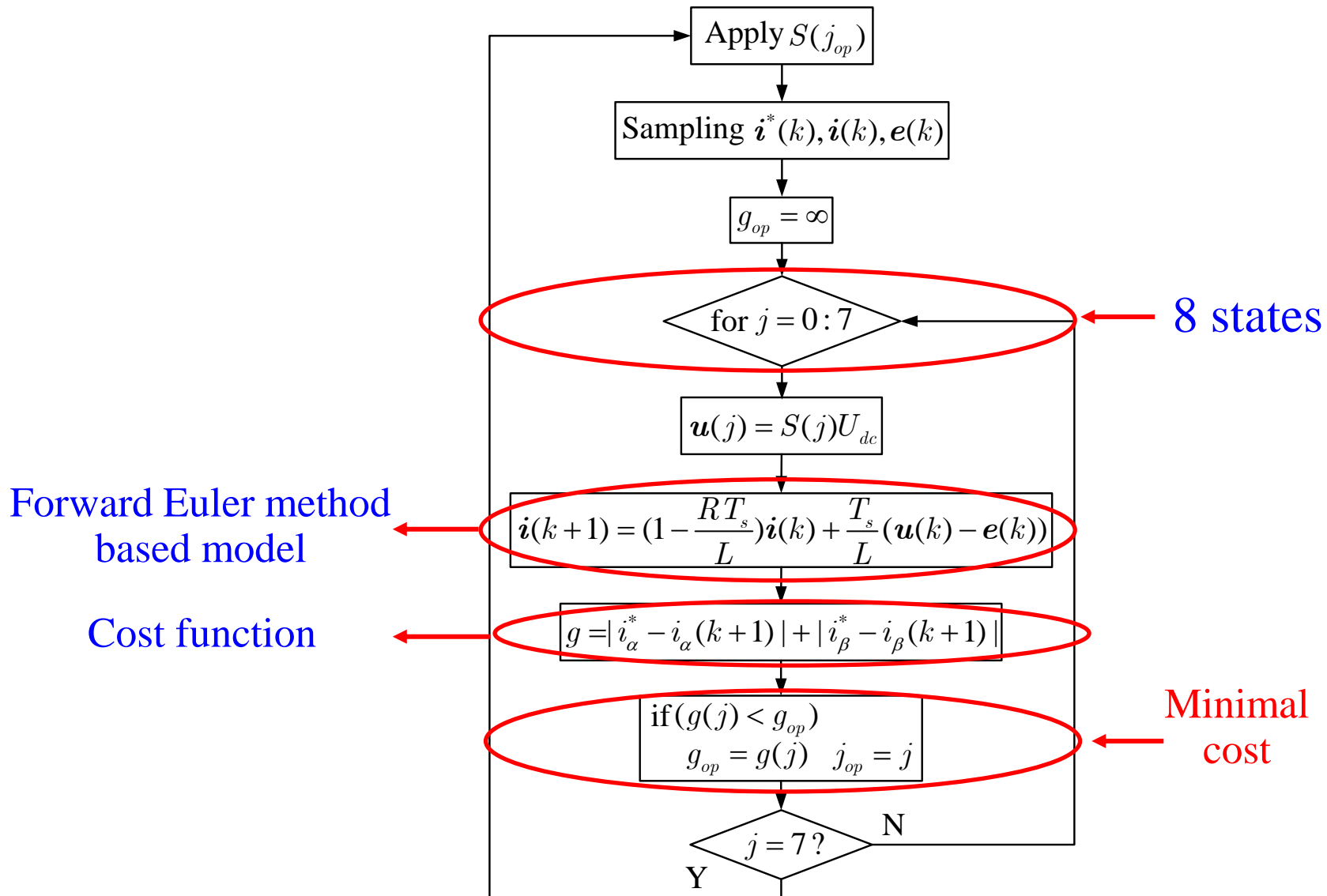
$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$
0	1	0	1	0	1
1	0	0	1	0	1
0	1	1	0	0	1
0	1	0	1	1	0
1	0	1	0	0	1
1	0	0	1	1	0
0	1	1	0	1	0
1	0	1	0	1	0

➤ Practical cost function:

$$J = |i_{\alpha}^*(k) - i_{\alpha}(k+1)| + |i_{\beta}^*(k) - i_{\beta}(k+1)|$$



Select the suitable switching state to get the minimal value





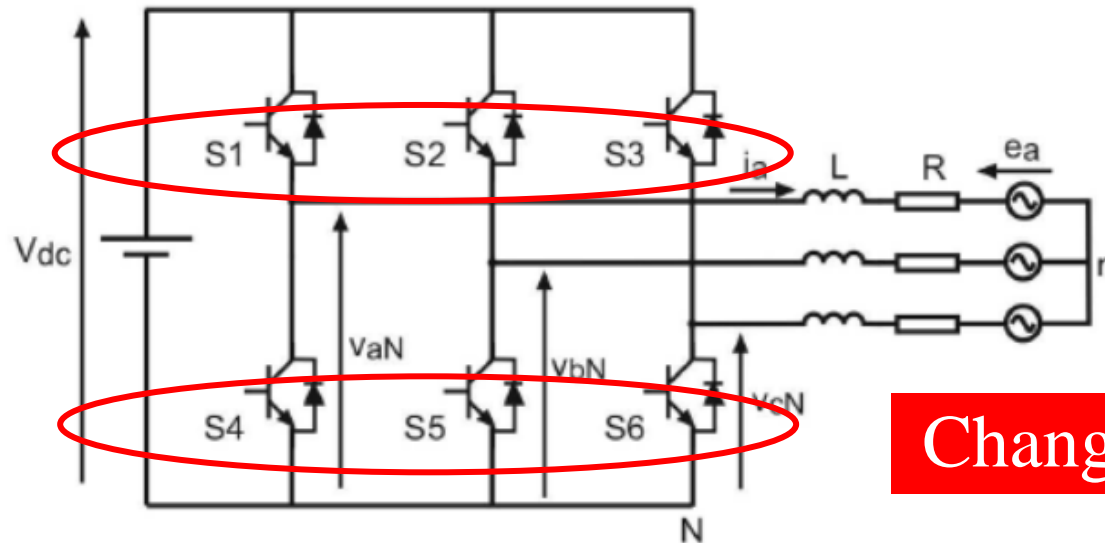


# FCS-MPC FOR POWER INVERTERS

## - EXAMPLE



# Parameters of the three-phase inverter



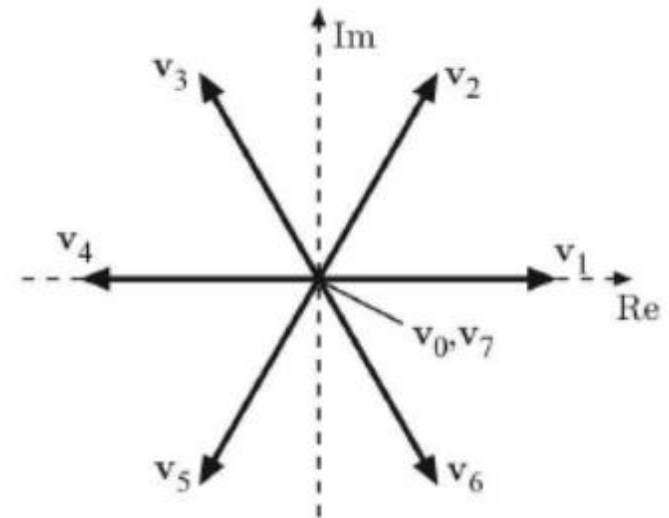
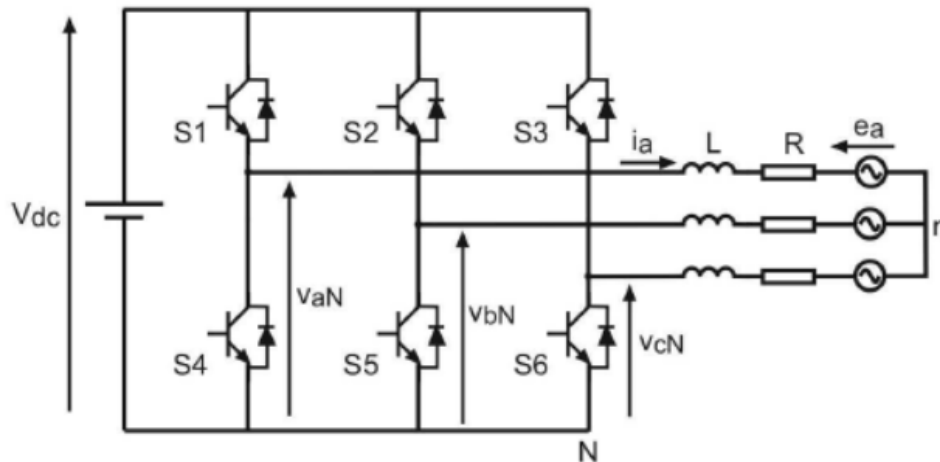
Change orders

$V_{dc}$	6.6 kV	$R$	$0.3 \Omega$
$e_{\text{line-line}}$	3.3 kV (rms)	$L$	2.5 mH
$F$	50 Hz	$T_s$	100 $\mu\text{s}$
$I_{\text{nominal}}$	3.5 kA (peak)	$f_s$	10 kHz



# Switching states of the example inverter

- Model of a three-phase inverter
  - Only 8 possible switching states
  - 7 different voltage vectors



$$S_a = \begin{cases} 1 & \text{if } S_1 \text{ on and } S_4 \text{ off} \\ 0 & \text{if } S_1 \text{ off and } S_4 \text{ on} \end{cases}$$
$$S_b = \begin{cases} 1 & \text{if } S_2 \text{ on and } S_5 \text{ off} \\ 0 & \text{if } S_2 \text{ off and } S_5 \text{ on} \end{cases}$$
$$S_c = \begin{cases} 1 & \text{if } S_3 \text{ on and } S_6 \text{ off} \\ 0 & \text{if } S_3 \text{ off and } S_6 \text{ on} \end{cases}$$

- Space vectors

$$\mathbf{S} = \frac{2}{3}(S_a + \mathbf{a}S_b + \mathbf{a}^2S_c)$$

$$\mathbf{v} = \frac{2}{3}(v_{aN} + \mathbf{a}v_{bN} + \mathbf{a}^2v_{cN})$$

$$\mathbf{v} = V_{dc}\mathbf{S}$$



# Detailed switching states of example inverter

$$\mathbf{v} = \frac{2}{3}(v_{aN} + \mathbf{a}v_{bN} + \mathbf{a}^2v_{cN}) \quad a = e^{j\frac{2\pi}{3}} = -\frac{1}{2} + j\frac{\sqrt{3}}{2}$$

$S_1$	$S_2$	$S_3$	$S_4$	$S_5$	$S_6$	Inverter terminal voltage space vector $v$
0	1	0	1	0	1	$v_0 = 0$
1	0	0	1	0	1	$v_1 = \frac{2}{3} V_{dc}$
0	1	1	0	0	1	$v_2 = \frac{1}{3}(-1 + j\sqrt{3})V_{dc}$
0	1	0	1	1	0	$v_3 = \frac{1}{3}(-1 - j\sqrt{3})V_{dc}$
1	0	1	0	0	1	$v_4 = \frac{1}{3}(1 + j\sqrt{3})V_{dc}$
1	0	0	1	1	0	$v_5 = \frac{1}{3}(1 - j\sqrt{3})V_{dc}$
0	1	1	0	1	0	$v_6 = -\frac{2}{3} V_{dc}$
1	0	1	0	1	0	$v_7 = 0$

- Load model

- Vector equation for the load current dynamics

$$\mathbf{v} = R\mathbf{i} + L\frac{d\mathbf{i}}{dt} + \mathbf{e}$$

where

$$\mathbf{v} = \frac{2}{3}(v_{aN} + \mathbf{a}v_{bN} + \mathbf{a}^2v_{cN})$$

$$\mathbf{i} = \frac{2}{3}(i_a + \mathbf{a}i_b + \mathbf{a}^2i_c)$$

$$\mathbf{e} = \frac{2}{3}(e_a + \mathbf{a}e_b + \mathbf{a}^2e_c)$$

- Discrete-time equations

$$\hat{\mathbf{i}}(k+1) = \left(1 - \frac{RT_s}{L}\right) \mathbf{i}(k) + \frac{T_s}{L} (\mathbf{v}(k) - \hat{\mathbf{e}}(k))$$

$$\hat{\mathbf{e}}(k+1) = \mathbf{v}(k+1) - \frac{L}{T_s} \mathbf{i}(k+1) + \left(R - \frac{L}{T_s}\right) \mathbf{i}(k)$$

$$\frac{d\mathbf{i}}{dt} \approx \frac{\mathbf{i}(k+1) - \mathbf{i}(k)}{T_s}$$

Forward Euler  
method



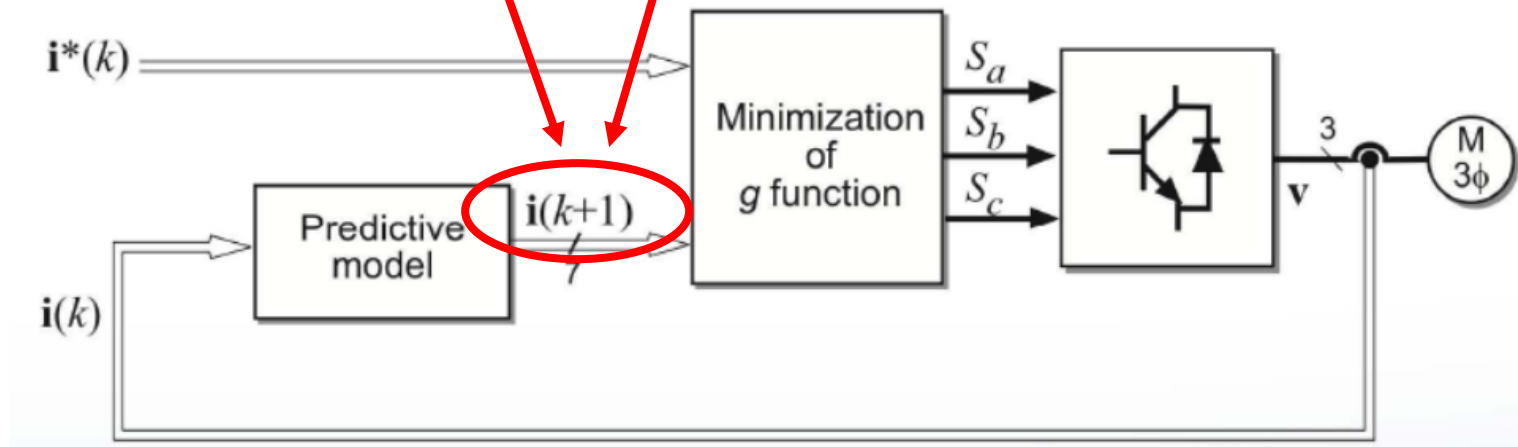
# Cost function of the example inverter

## Cost function:

$$g = |i_{\alpha}^* - i_{\alpha}^p| + |i_{\beta}^* - i_{\beta}^p|$$

$i_{\alpha}^*, i_{\beta}^*$  : reference values

$i_{\alpha}^p, i_{\beta}^p$  : predicted values

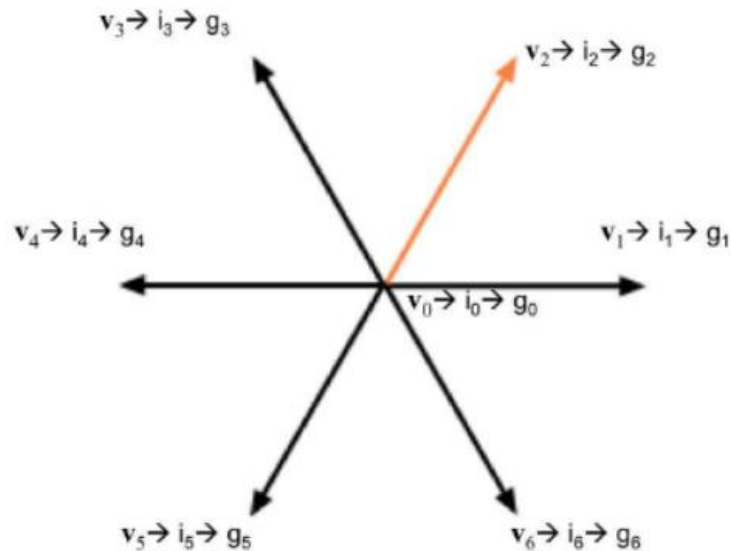


- No need for linear controllers !!
- No need for modulator (PWM or SVM) !!



# Select the optimal state for the example inverter

## Cost function minimization



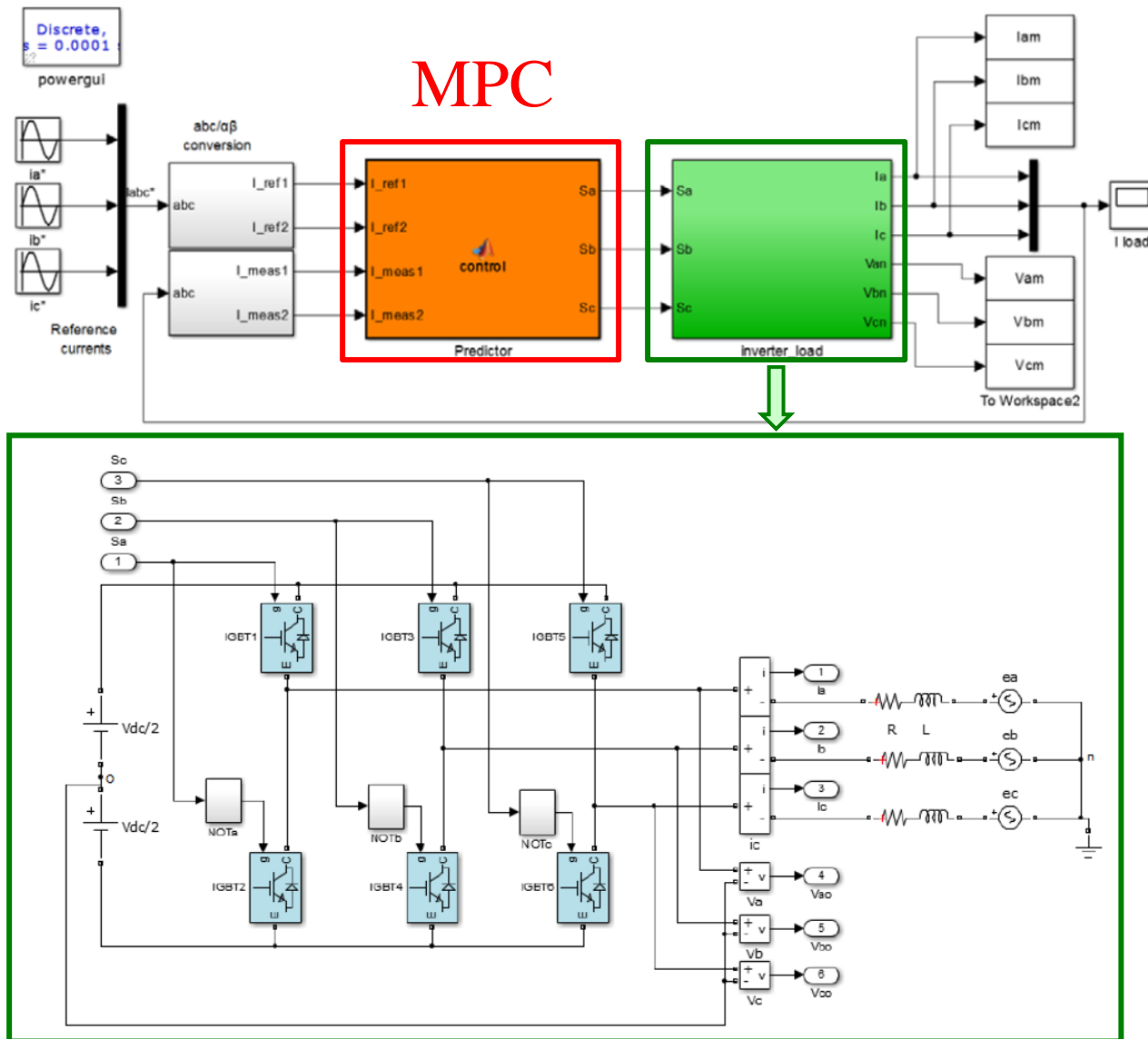
$\mathbf{v}_0$	$\mathbf{g}_0$	0.60
$\mathbf{v}_1$	$\mathbf{g}_1$	0.82
$\mathbf{v}_2$	$\mathbf{g}_2$	0.24
$\mathbf{v}_3$	$\mathbf{g}_3$	0.42
$\mathbf{v}_4$	$\mathbf{g}_4$	0.96
$\mathbf{v}_5$	$\mathbf{g}_5$	1.24
$\mathbf{v}_6$	$\mathbf{g}_6$	1.19

←  $g_{\min}$

- Voltage vector  $\mathbf{v}_0$  is used to predict  $i_0$  and to calculate cost function (error)  $g_0$ .
- Voltage vector  $\mathbf{v}_1$  is used to predict  $i_1$  and to calculate cost function (error)  $g_1$ .
- ...
- $g_{\min} = g_2$ .
- Voltage vector  $\mathbf{v}_2$  is selected and will be applied during the next sampling interval.



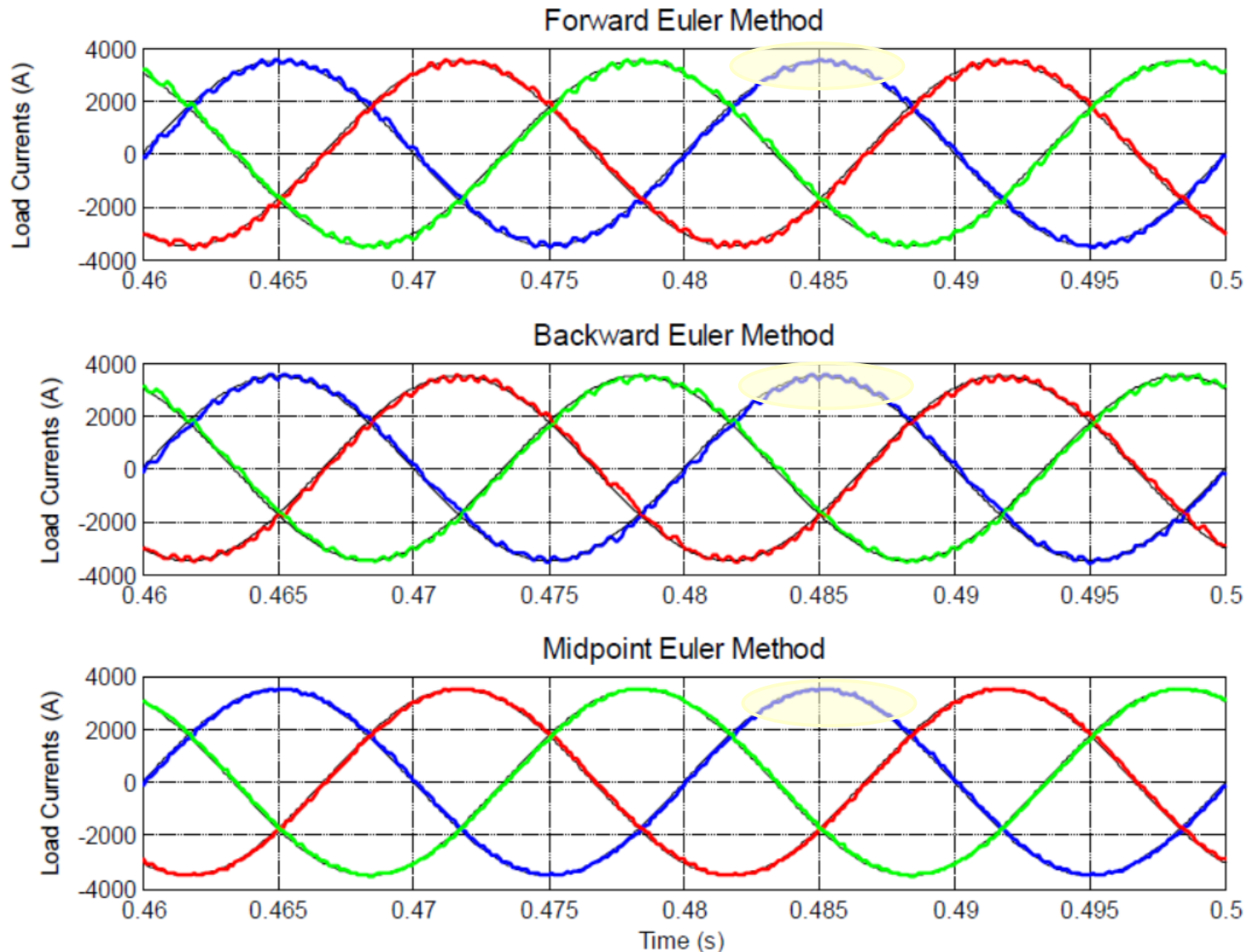
# Simulation of the example inverter





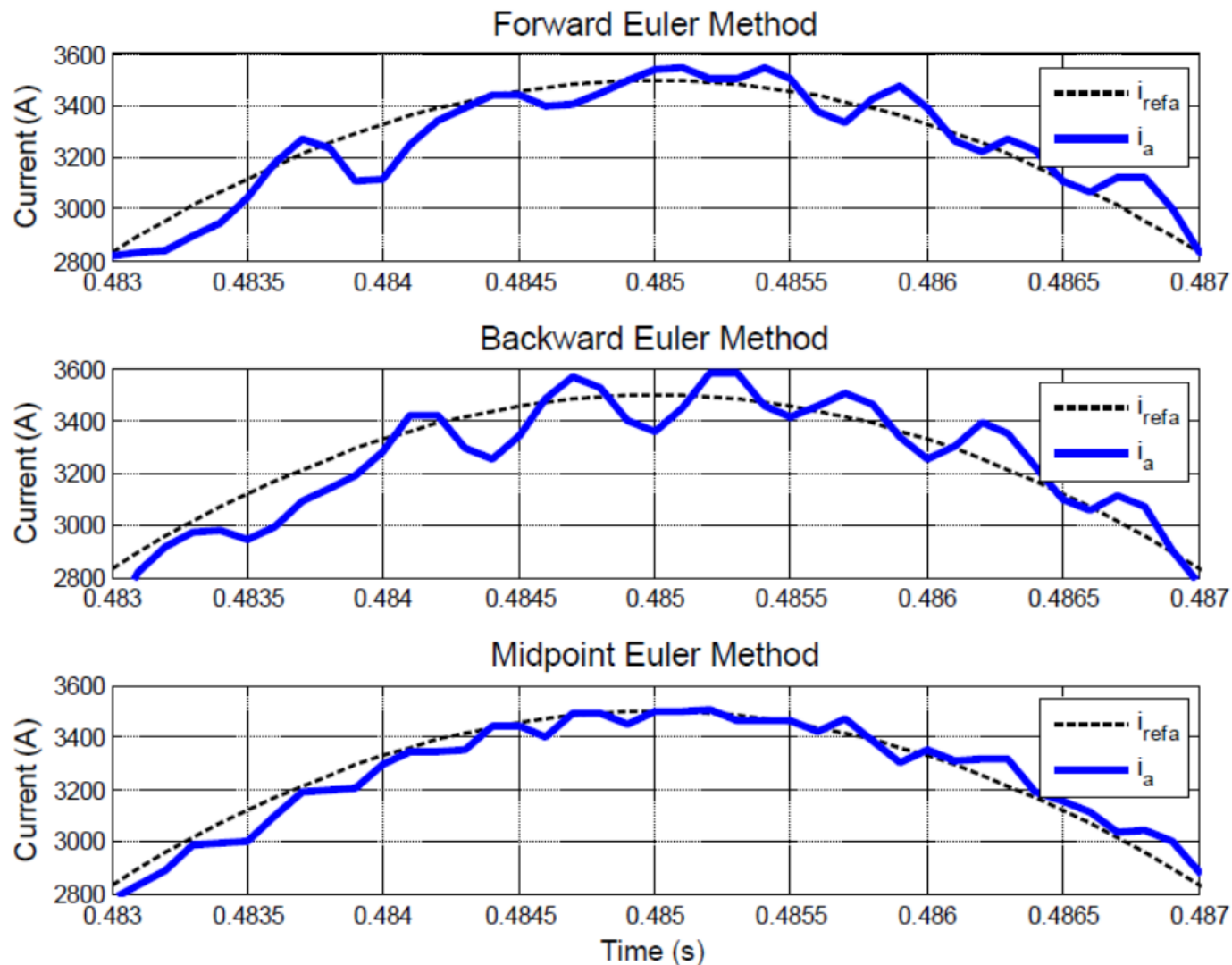


# Simulation results with different discrete models

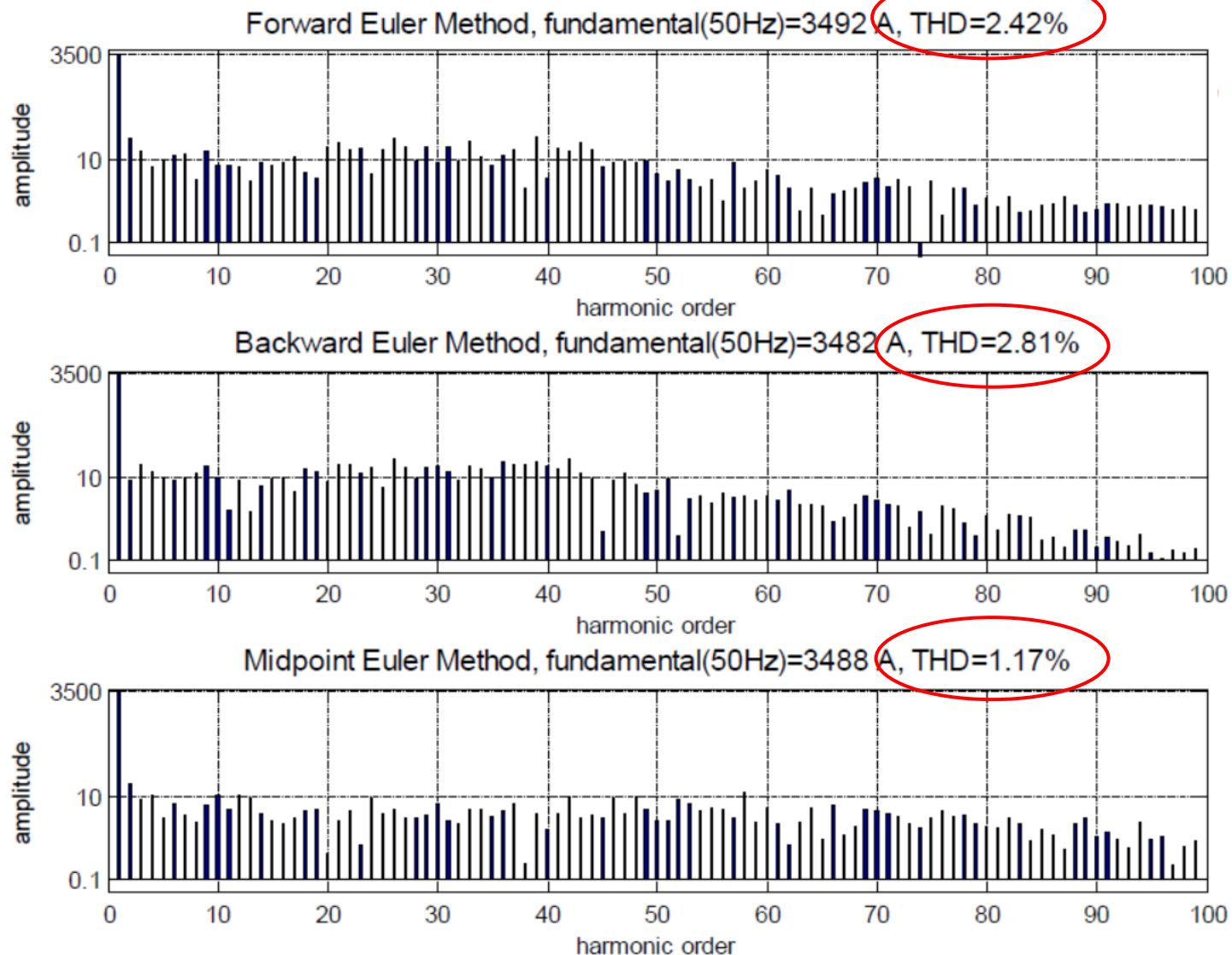




# Simulation : ripple waveforms (comparisons)

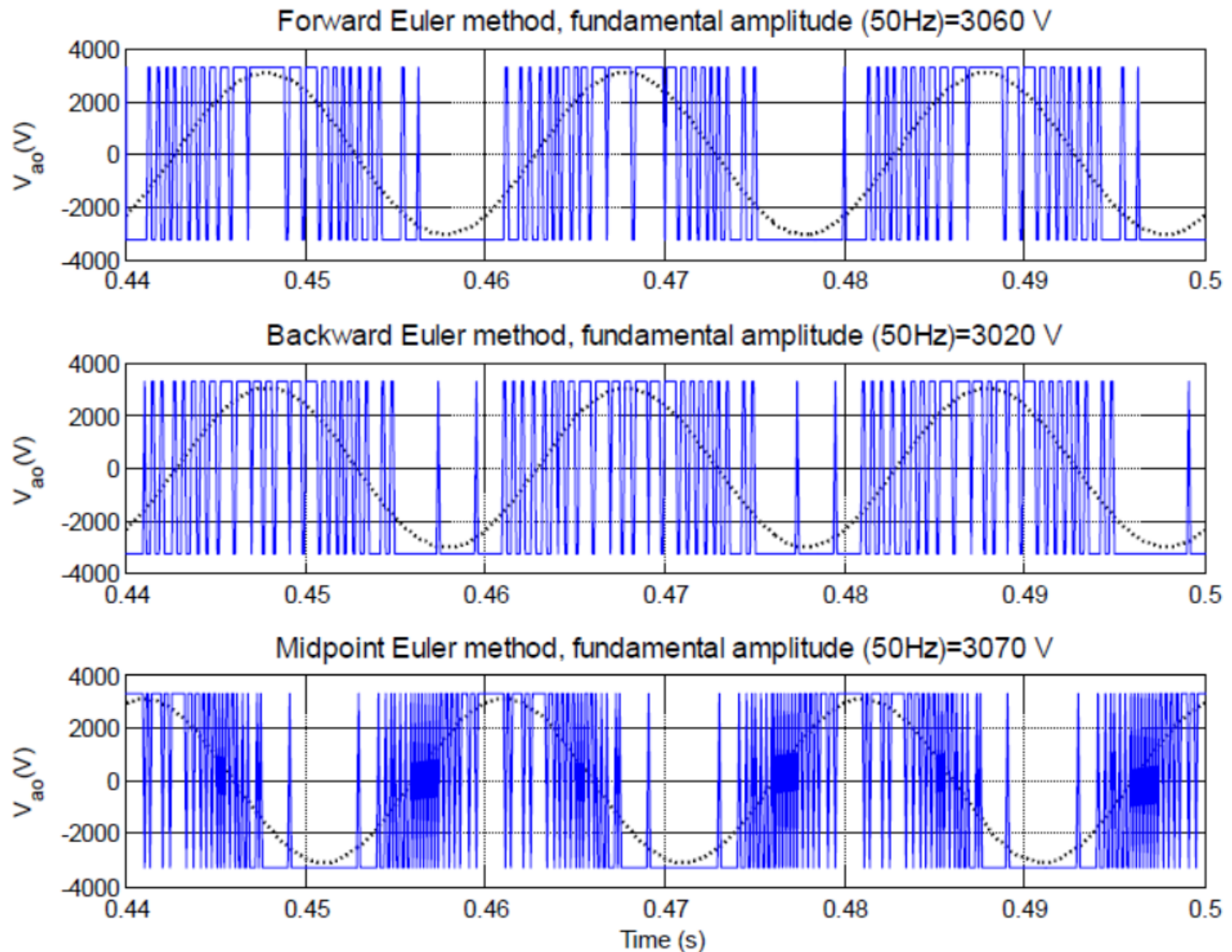


# Simulation: THD (comparisons)





# Simulation: switching waveforms (comparisons)



- The backward & forward Euler based models give smaller switching frequency.
  - Less accurate
  - Less Power loss
  
- The Midpoint Euler based models give bigger switching frequency.
  - More accurate
  - More Power loss

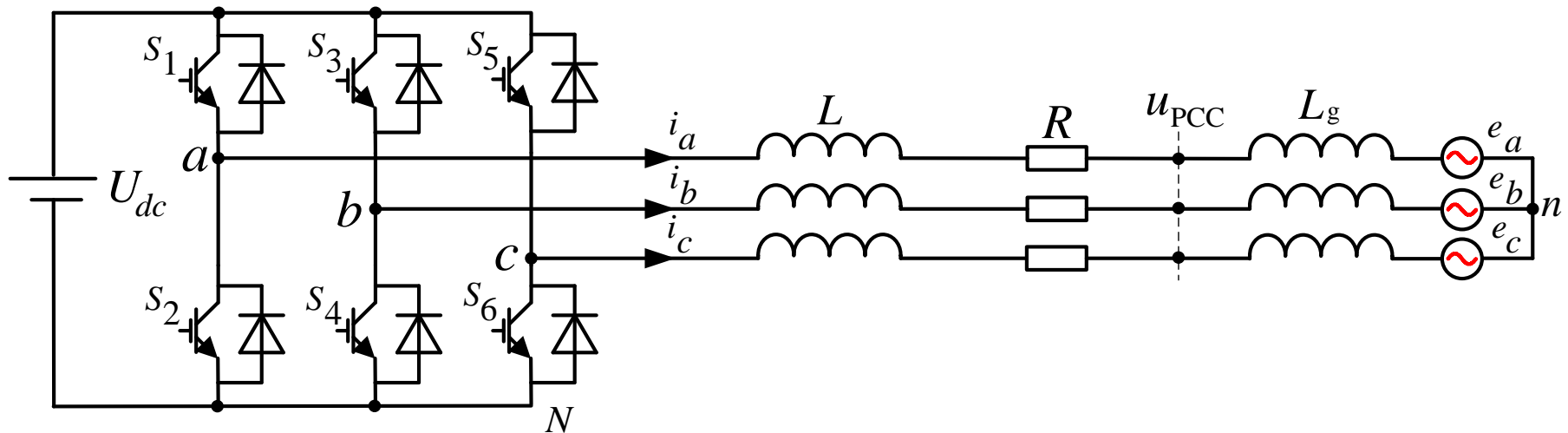
# APPLICATION OF FINITE CONTROL SET MPC IN THREE PHASE INVERTER

[07/11/2019]

- Basic knowledge of power inverters
- Finite Control Set (FCS) MPC for power inverters
- Improved FCS-MPC for power inverters



- 2 level 3 phase grid-connected inverter
- Control in  $\alpha\beta$  coordinates
- Differential equation of the load:  $\underline{v} = R \cdot \underline{i} + L \frac{d\underline{i}}{dt} + \underline{e}$
- $\underline{v}$ : Applied voltage vector







- Time-discrete load model:

Euler-forwards approximation:

$$\frac{d\underline{i}}{dt} \approx \frac{\underline{i}(k+1) - \underline{i}(k)}{T_s}$$

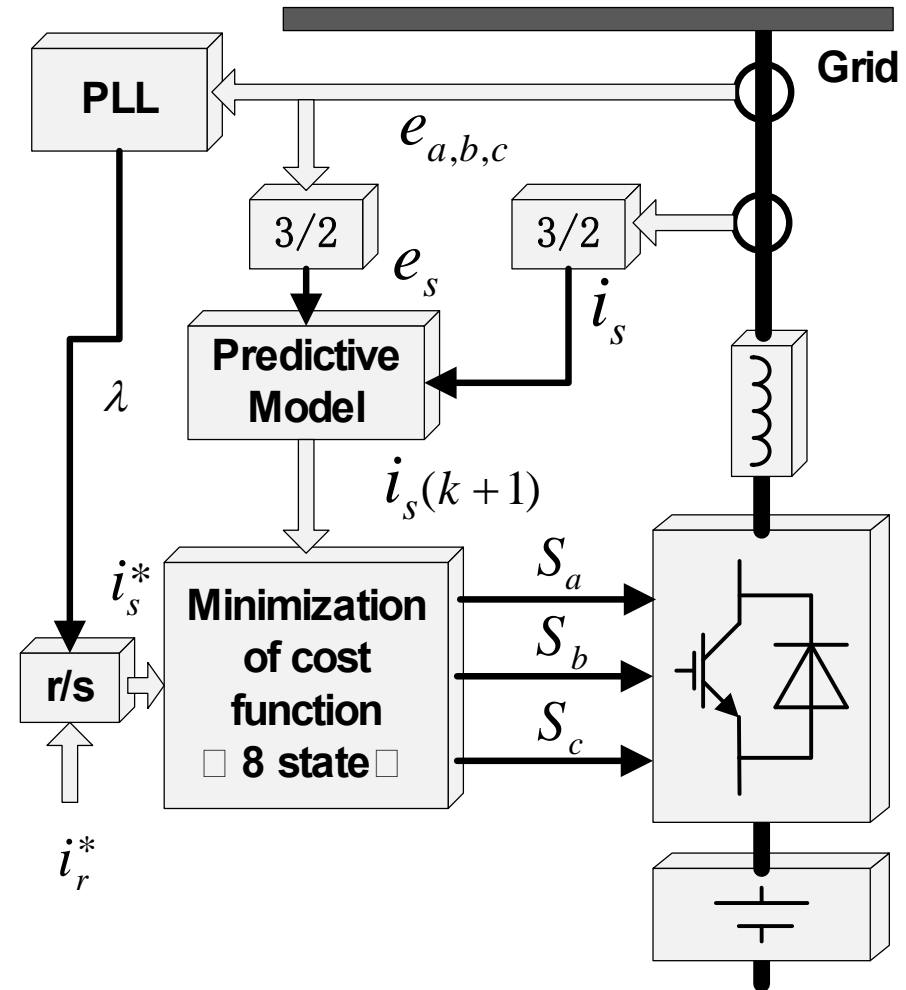
- Current prediction:

$$\underline{i}(k+1) = \left(1 - \frac{RT_s}{L}\right) \underline{i}(k) + \frac{T_s}{L} (\underline{v}(k) - \underline{e}(k))$$

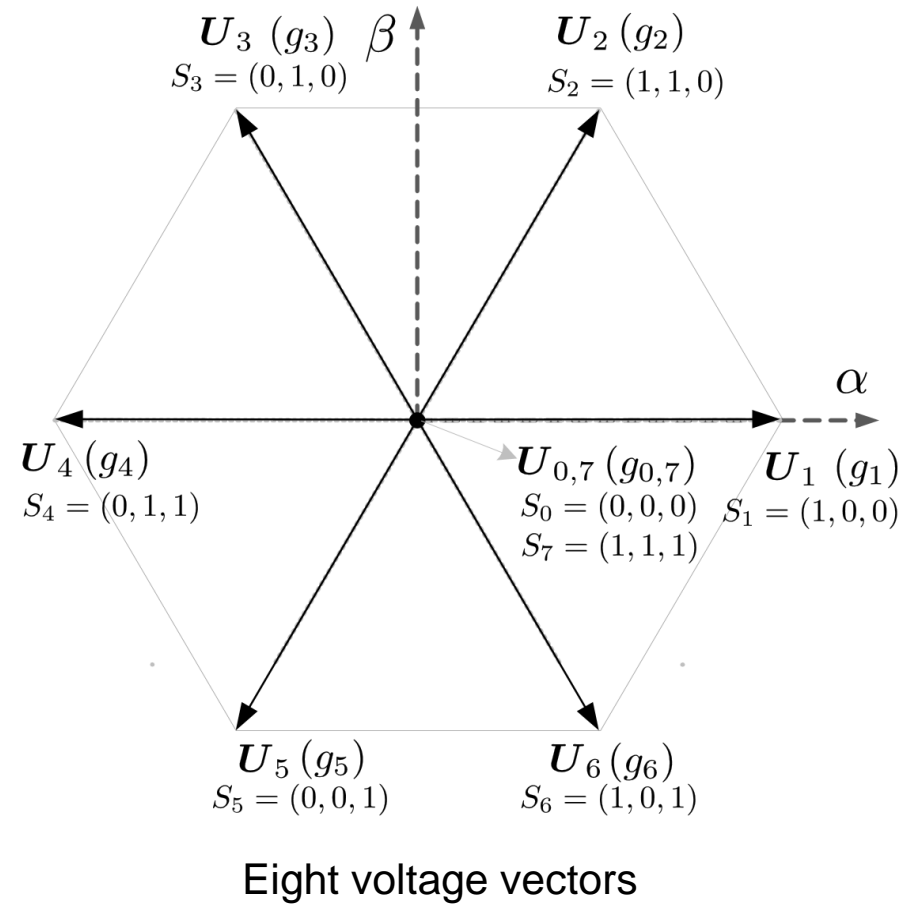
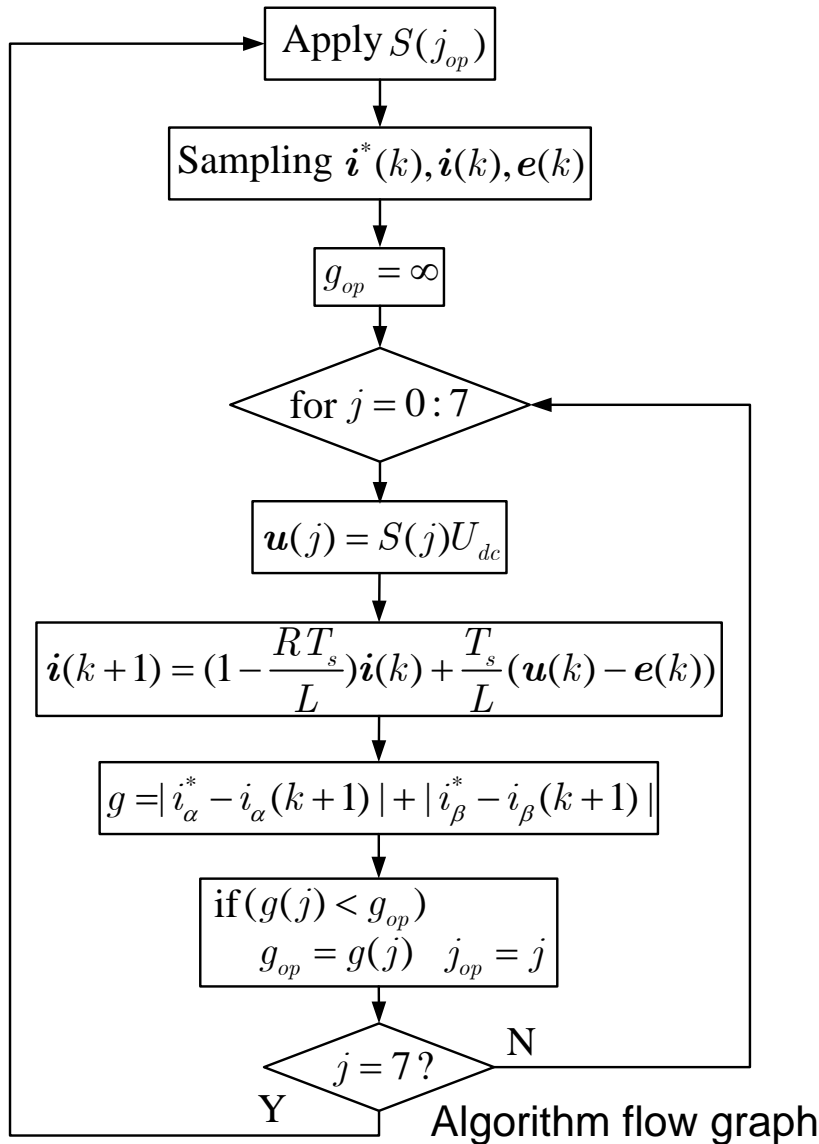
- Cost function:

$$G =$$

$$|i_{\alpha}^* - i_{\alpha}(k+1)| + |i_{\beta}^* - i_{\beta}(k+1)|$$



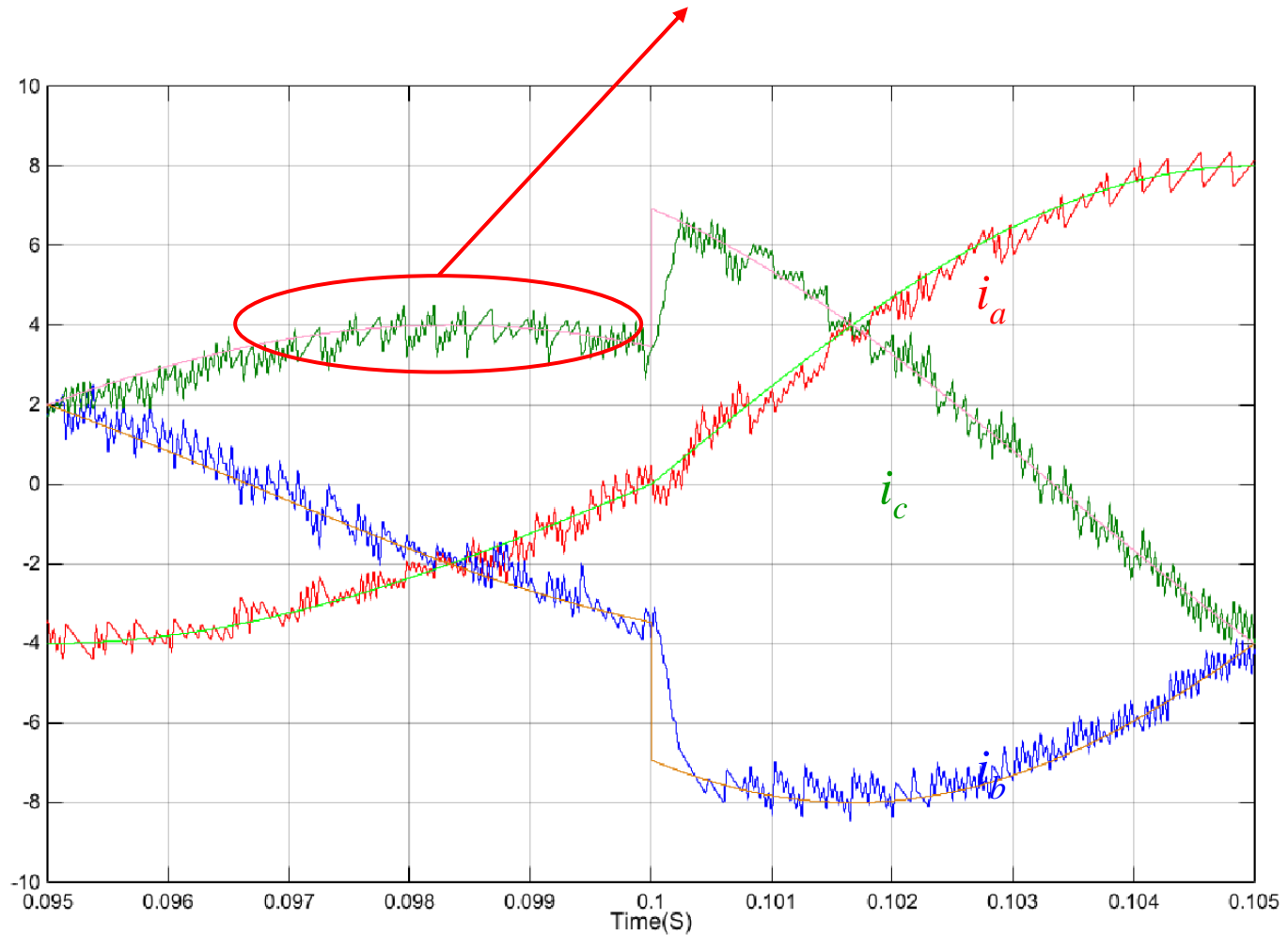






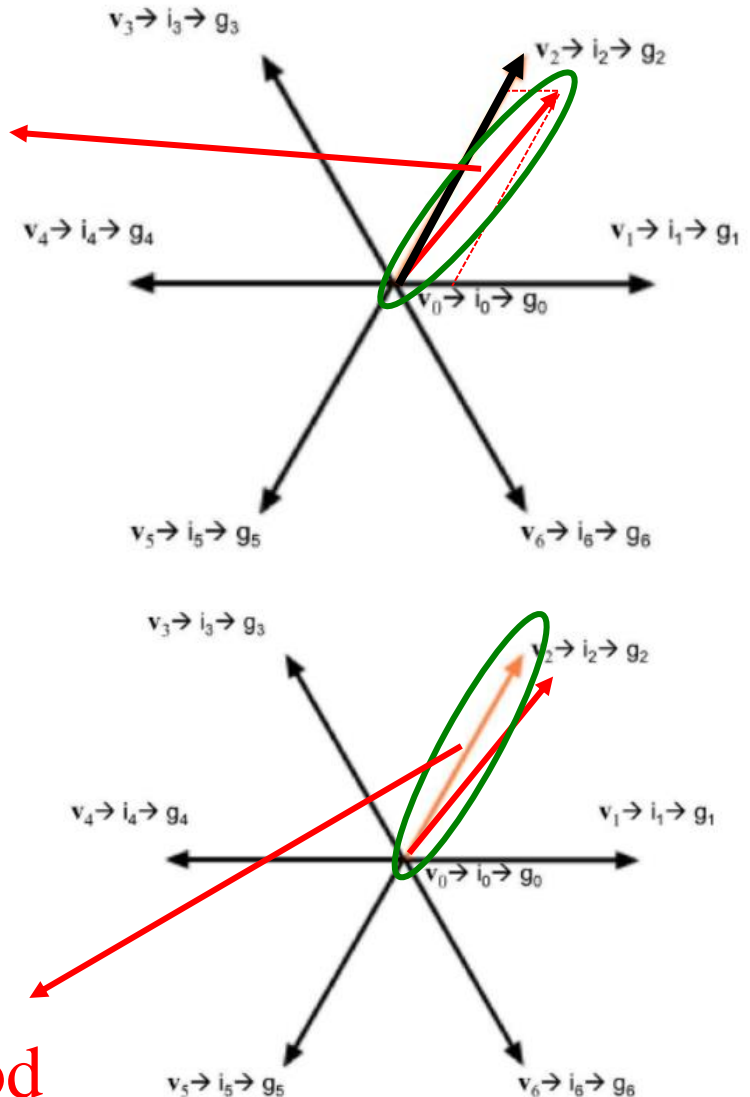
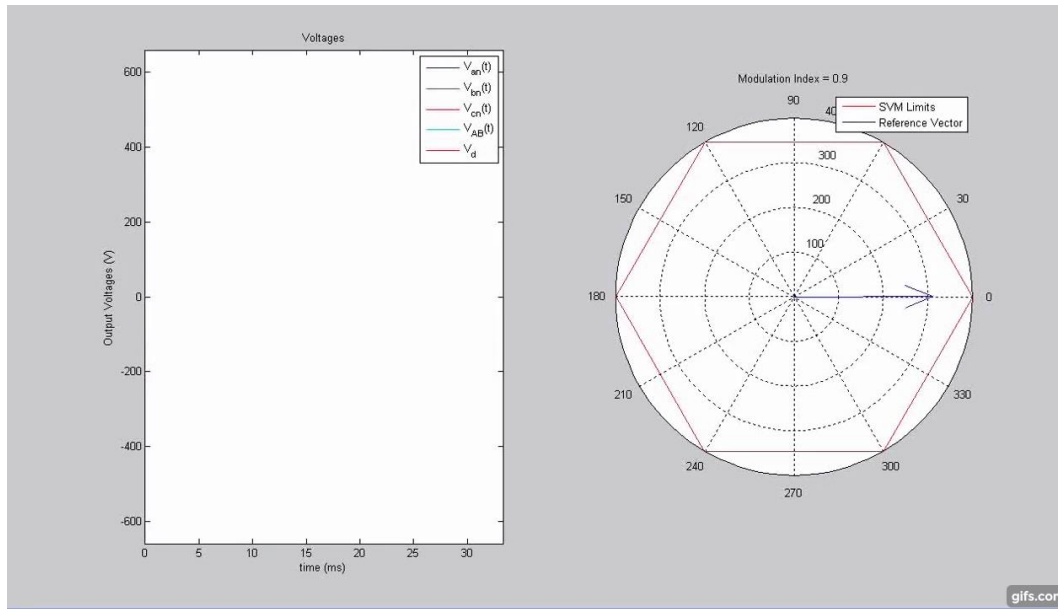
# Problem: variable switching frequency

Variable switching frequency





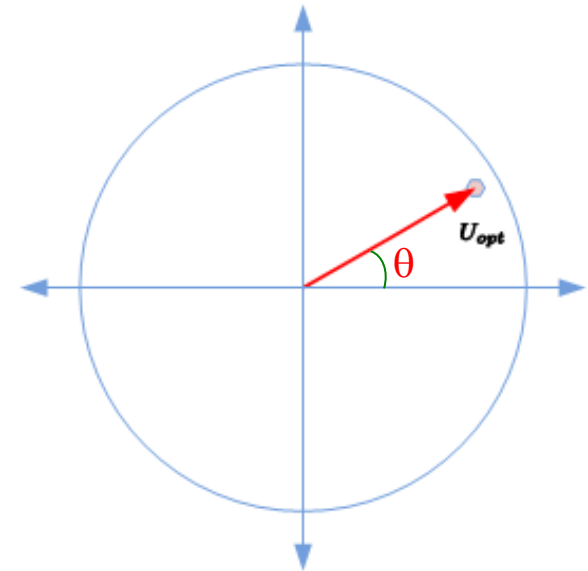
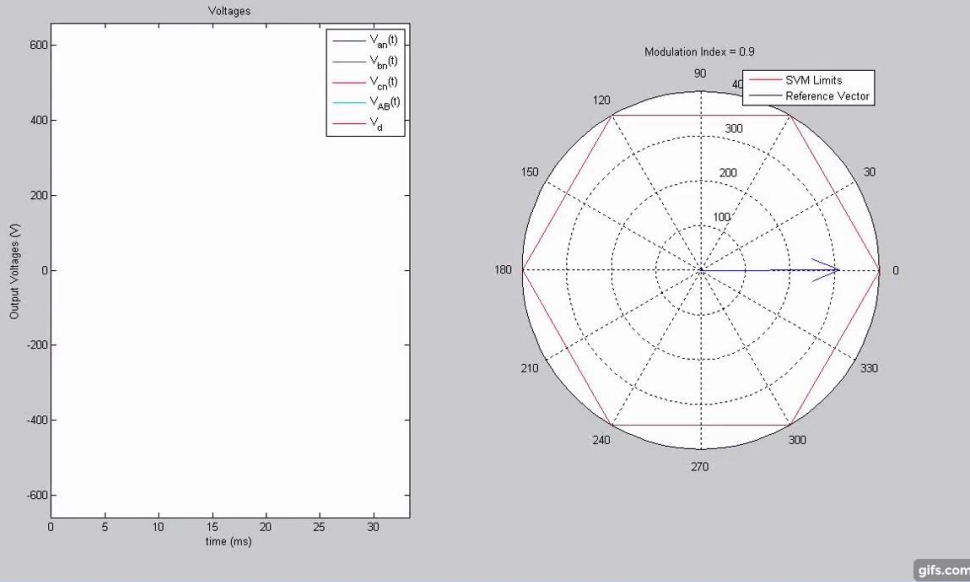
Best vector



Not the targeted vector,  
But we select it via FCS MPC method



# Improved direction: Directly determine the amplitude and angle



- Need to confirm the **amplitude** of the phasor
- Need to confirm the **angle** of the phasor

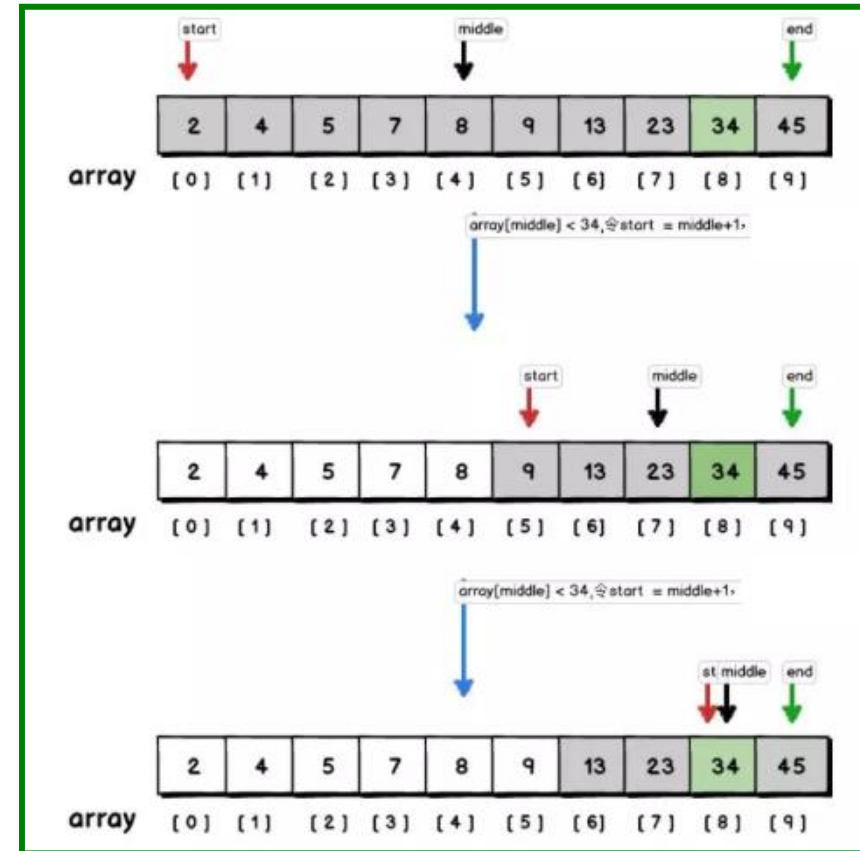
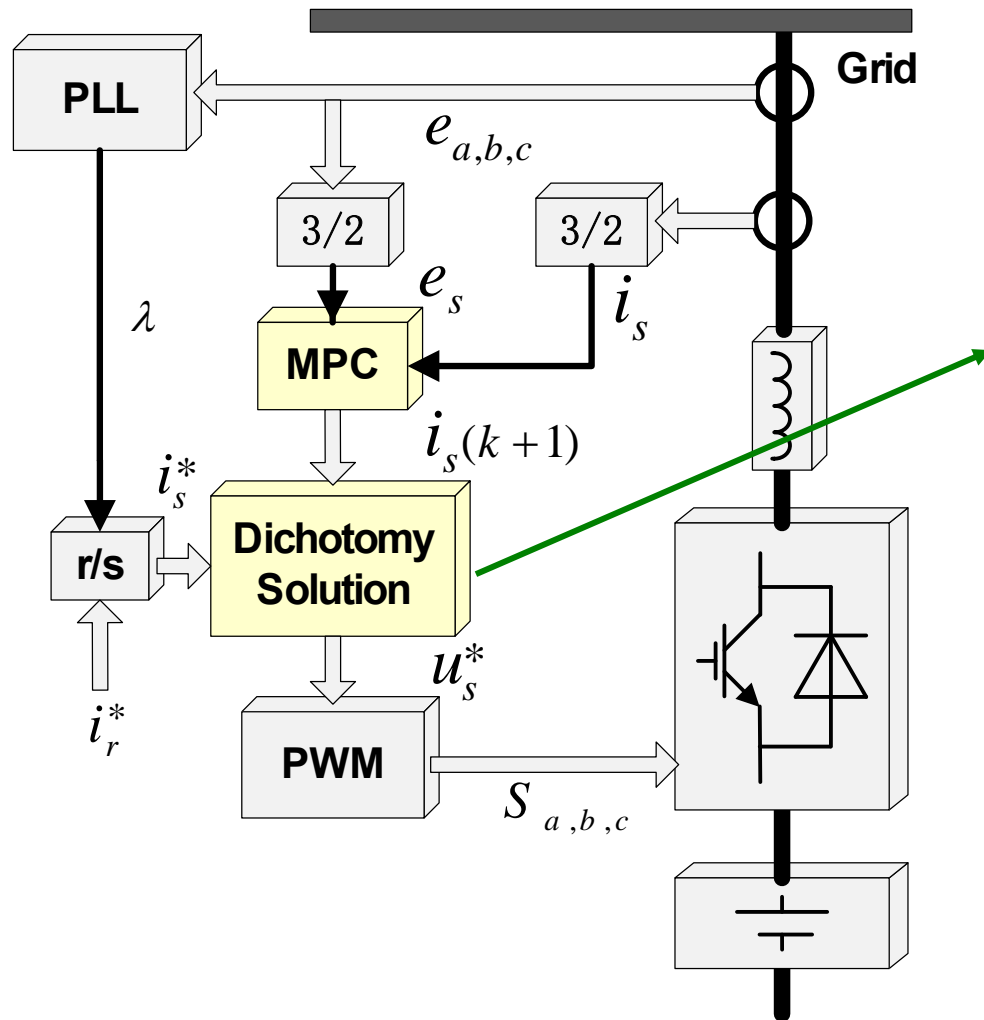


Constant switching frequency (CSF)

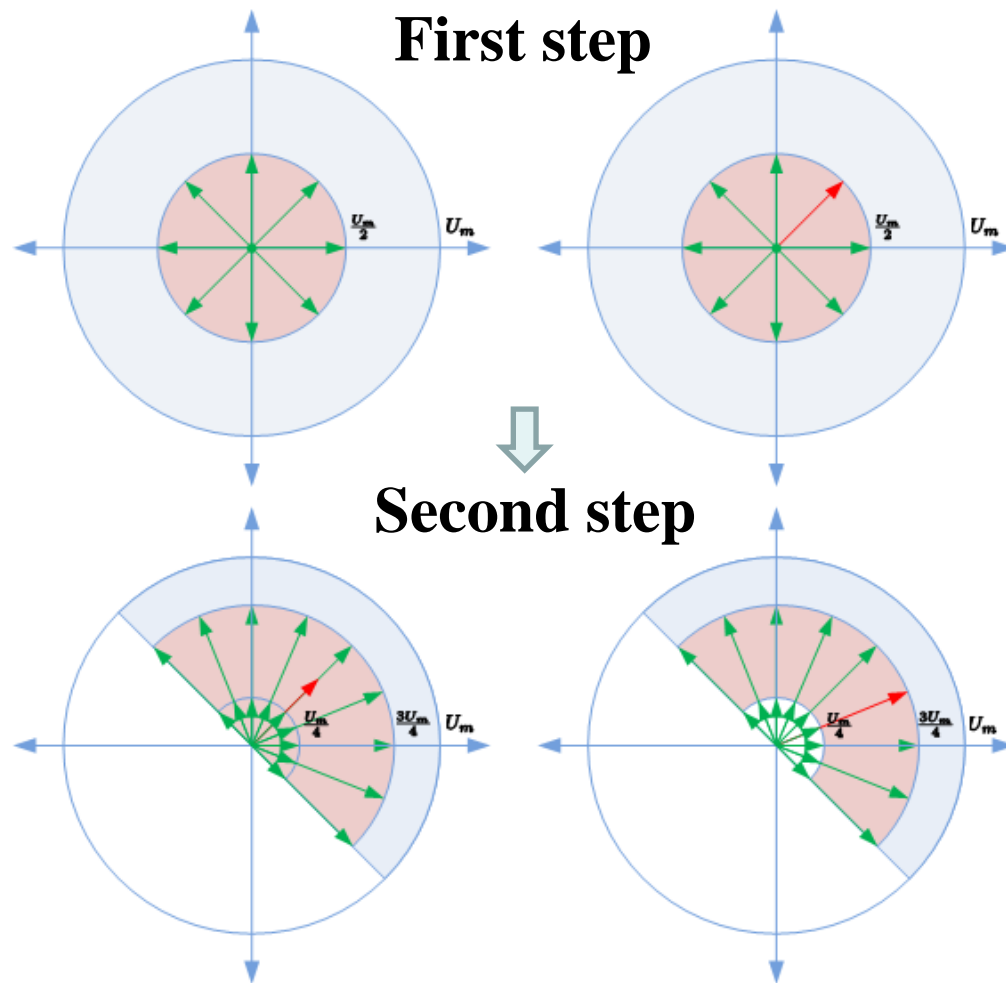


# Dichotomy solution to find the amplitude & angle

## Proposed CSF-MPC

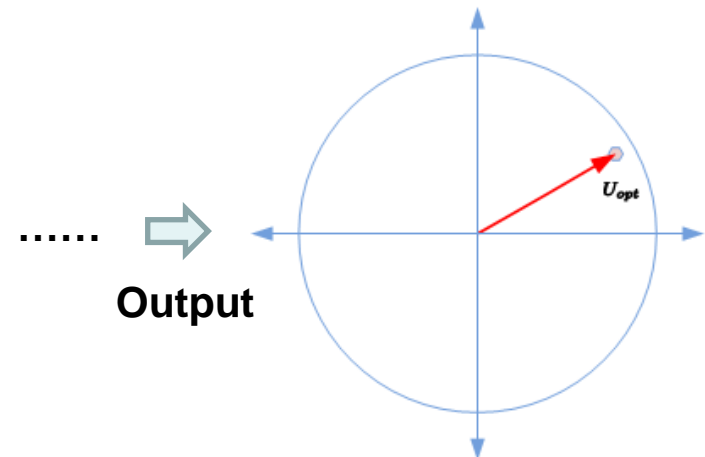


# Improved MPC: Schematic diagram



Key idea: **Dichotomy**

- voltage vector search area decrease to half of the former steps.
- The convergence of voltage vector selection is fast.



## Flow chart:

**N:**  
Resolution of  
the output  
voltage vector

**n:**  
Phase search  
resolution

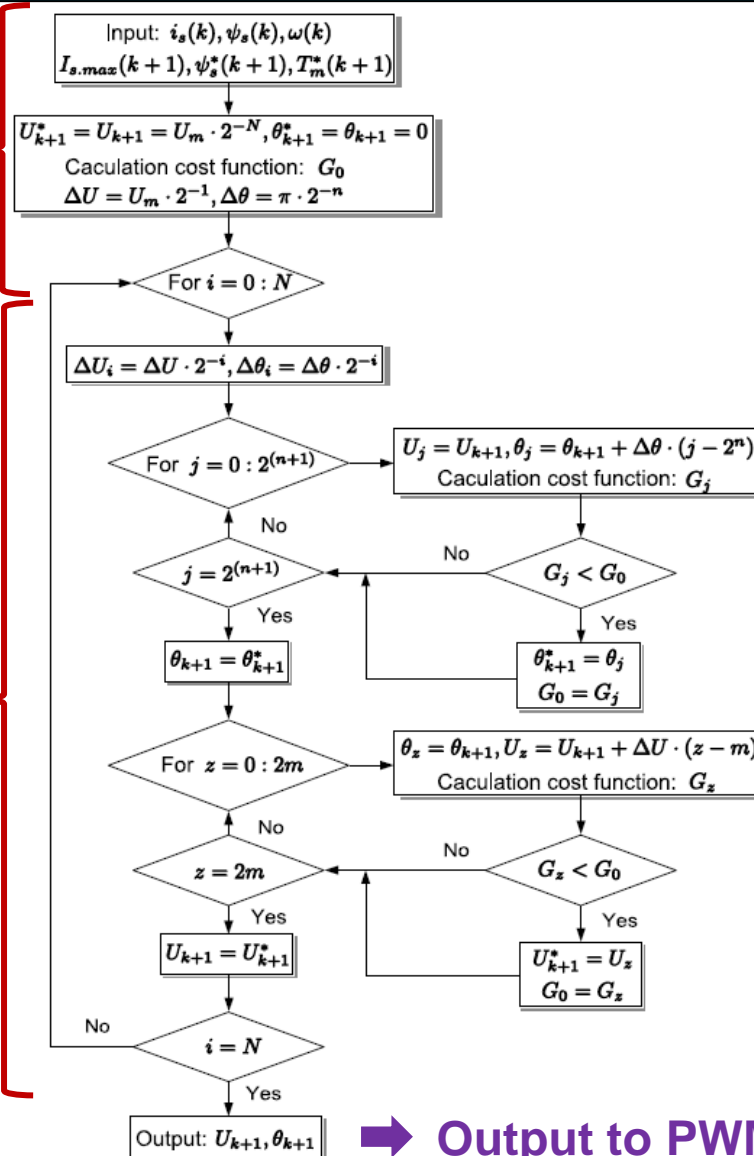
**2m:**  
Amplitude  
search  
resolution

**Initialization**

**Search  
Resolution**

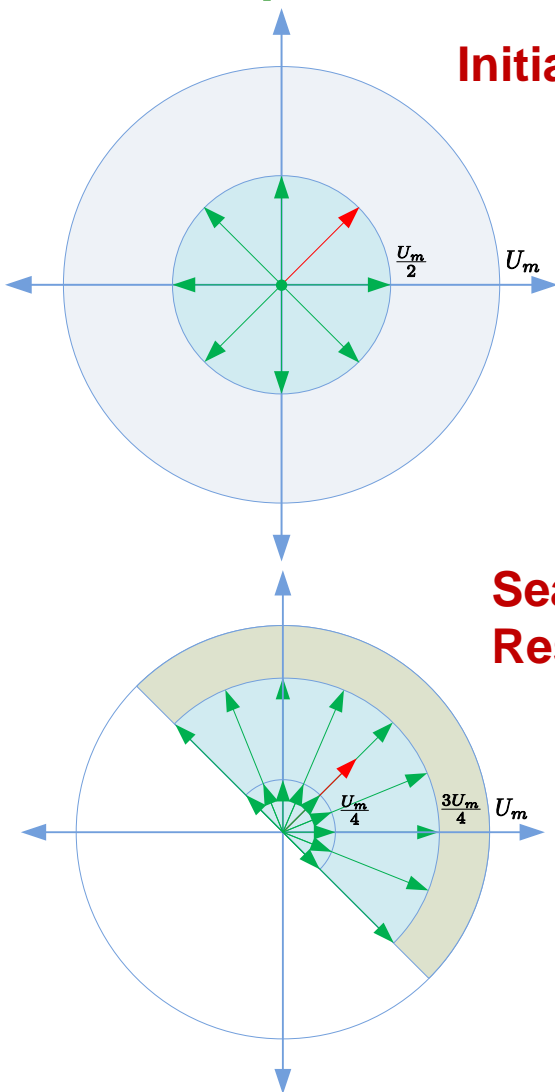
**Phase  
Calculation  
Loop**

**Amplitude  
Calculation  
Loop**



➡ Output to PWM modulator

One example:  $N=14$ ,  $n=2$ ,  $m=2$



**Initialization**

**Search Resolution**

Input:  $i_s(k), \psi_s(k), \omega(k)$   
 $I_{s,max}(k+1), \psi_s^*(k+1), T_m^*(k+1)$   
 $U_{k+1}^* = U_{k+1} = U_m \cdot 2^{-N}, \theta_{k+1}^* = \theta_{k+1} = 0$   
 Calculation cost function:  $G_0$   
 $\Delta U = U_m \cdot 2^{-1}, \Delta \theta = \pi \cdot 2^{-n}$

For  $i = 0 : N$

$\Delta U_i = \Delta U \cdot 2^{-i}, \Delta \theta_i = \Delta \theta \cdot 2^{-i}$

For  $j = 0 : 2^{(n+1)}$

$U_j = U_{k+1}, \theta_j = \theta_{k+1} + \Delta \theta \cdot (j - 2^n)$   
 Calculation cost function:  $G_j$

$j = 2^{(n+1)}$

$G_j < G_0$

$\theta_{k+1}^* = \theta_j$   
 $G_0 = G_j$

$\theta_{k+1} = \theta_{k+1}^*$

For  $z = 0 : 2m$

$\theta_z = \theta_{k+1}, U_z = U_{k+1} + \Delta U \cdot (z - m)$   
 Calculation cost function:  $G_z$

$z = 2m$

$G_z < G_0$

$U_{k+1}^* = U_z$   
 $G_0 = G_z$

$U_{k+1} = U_{k+1}^*$

$i = N$

Output:  $U_{k+1}, \theta_{k+1}$

**Phase Calculation Loop**

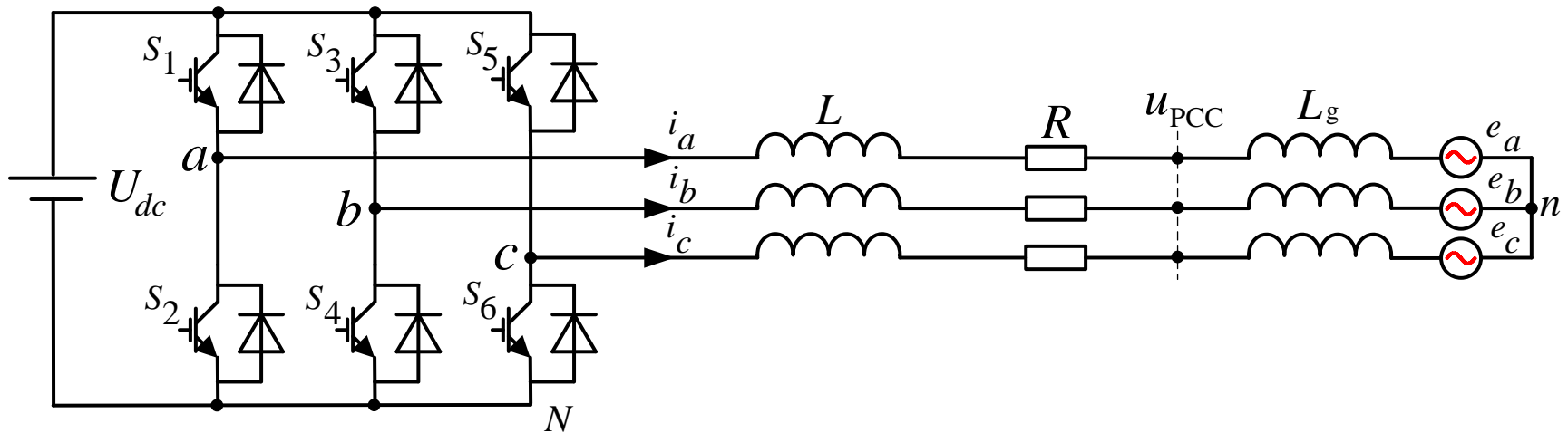
**Amplitude Calculation Loop**

**Output to PWM modulator**





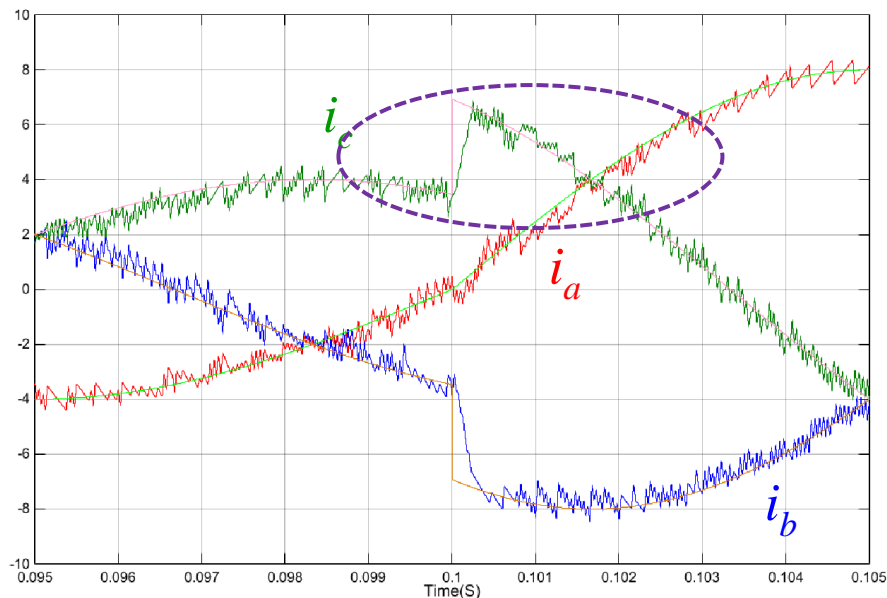
## Example of the improved FCS-MPC method



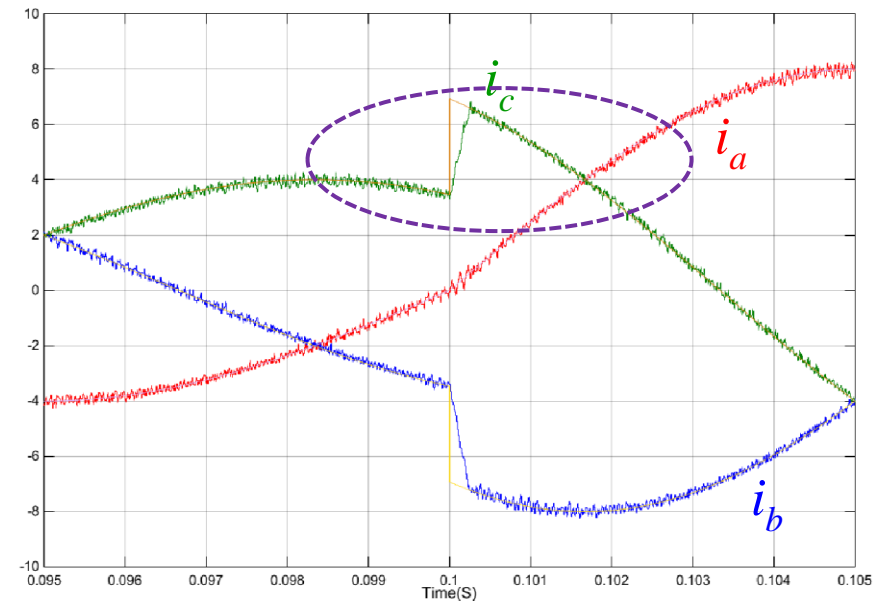
PARAMETERS	VALUE
RATED POWER $P_N$	2 kW
DC BUS VOLTAGE $U_{DC}$	300 V
AC BUS VOLTAGE	115 V (RMS)
FILTER INDUCTANCE $L$	1 mH
GRID FUNDAMENTAL FREQUENCY	50 Hz
LINE RESISTANCE $R$	0.01 $\Omega$

## Dynamic process of the active current

### Conventional FCS-MPC



### Improved CSF-MPC



Constant switching frequency

- Understand the basic concept of the FCS-MPC in power converters
- Know the tips of the FCS-MPC for a power inverter, i.e., model, cost function, select methods.
- Know the improved FCS-MPC method for a power inverter.

*Thank you!*

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