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Model Predictive Torque Control of Synchronous Machines Without a Current or Stator Flux Reference Generator

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School of Mechanical
Engineering

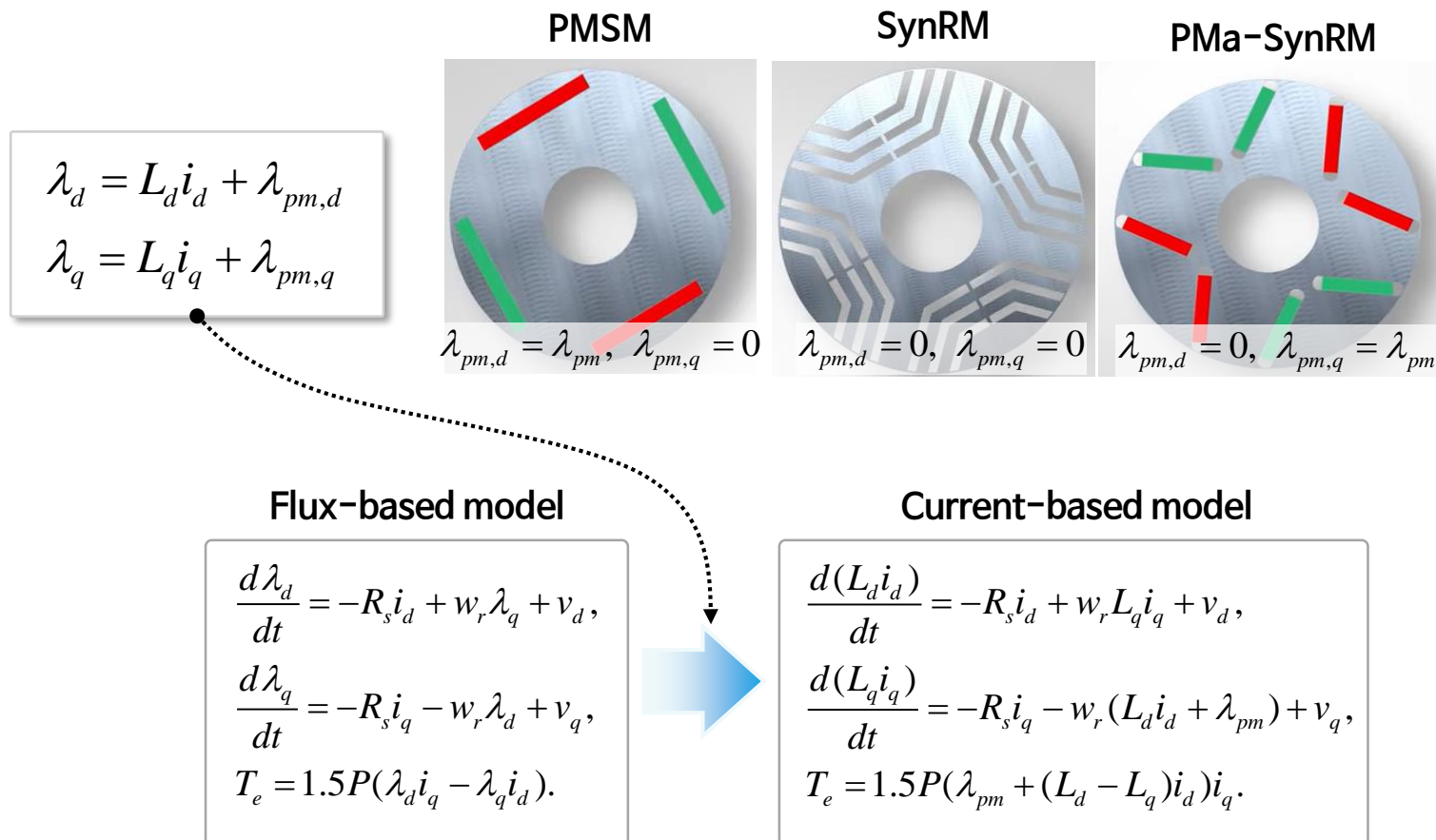
MIC LAB

1. Introduction
2. Proposed MPTC
3. Validation
4. Conclusion & Further Work

1. Introduction

Synchronous Machines (SMs)

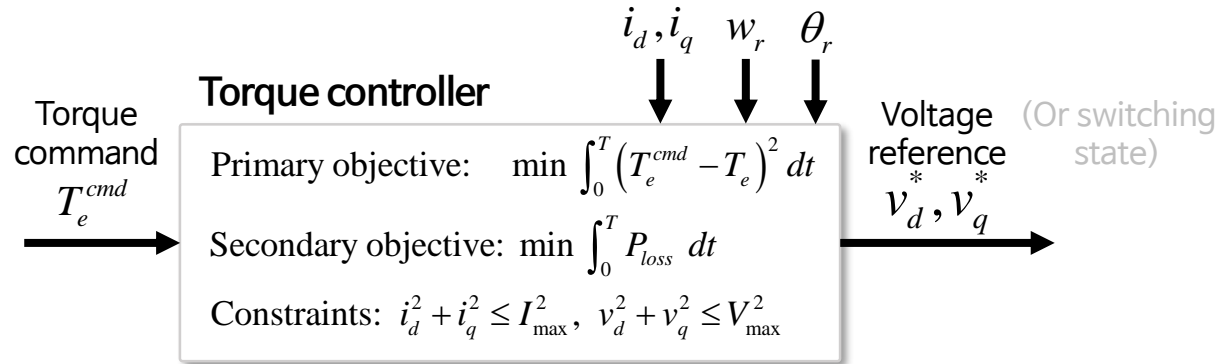
Modeling



- $L_{d(q)}$: $d(q)$ -axis inductance, λ_{pm} : $d(q)$ -axis magnetic flux linkage, R_s : stator resistance, P : number of pole pairs, J : rotational inertia, B : damping coefficient, T_l : load torque.

Synchronous Machines (SMs)

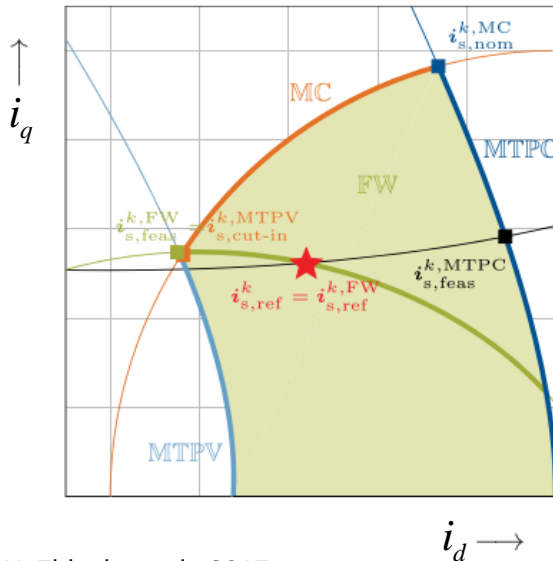
Torque control



Note DOF in the torque equation: $T_e = 1.5P(\lambda_{pm} + (L_d - L_q)i_d)i_q$

➤ Need to optimize operating points [1]

➤ How to determine the operating point?



1) Solve the optimization analytically [1]/numerically [2]

Case 1. $T_e^* \leq T_e^{max}$

$$\min i_d^2 + i_q^2$$

$$\text{subject to } T_e^* = (k_1 + k_2 i_d) i_q,$$

$$v_d^2 + v_q^2 \leq V_{max}^2.$$

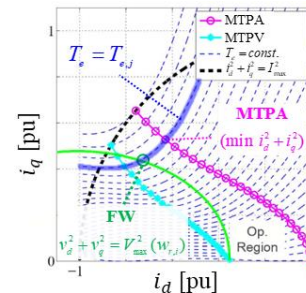
Case 2. $T_e^* > T_e^{max}$

$$\max \text{sgn}(T_e) T_e$$

$$\text{subject to } v_d^2 + v_q^2 \leq V_{max}^2,$$

$$i_d^2 + i_q^2 \leq I_{max}^2.$$

2) Identify experimentally [3]



Stored in LUTs

Speed	Torque	i_d^*	i_q^*
$w_{r,1}$	$T_{e,1}$	$i_{d,11}$	$i_{q,11}$
$w_{r,1}$	$T_{e,2}$	$i_{d,12}$	$i_{q,12}$
...
$w_{r,j}$	$T_{e,j-1}$	$i_{d,j(j-1)}$	$i_{q,j(j-1)}$
$w_{r,j}$	$T_{e,j}$	$i_{d,jj}$	$i_{q,jj}$
...
$w_{r,m}$	$T_{e,m-1}$	$i_{d,m(m-1)}$	$i_{q,m(m-1)}$
$w_{r,m}$	$T_{e,m}$	$i_{d,mm}$	$i_{q,mm}$

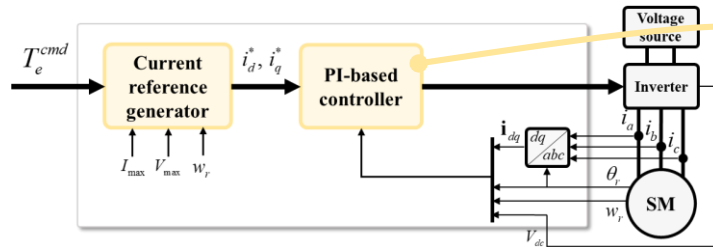
[1] H. Eldeeb, et al., 2017.

[2] K. Choi, et al., 2020.

[3] B. Gallert, et al., 2017

Existing schemes for Torque control of SMs

Reference generator + PI-based controller [4]



PI control

$$\begin{aligned} v_{d,fb}^* &= K_{pd}e_d + K_{id} \int e_d \\ v_{q,fb}^* &= K_{pq}e_q + K_{iq} \int e_q \end{aligned}$$

Anti-windup

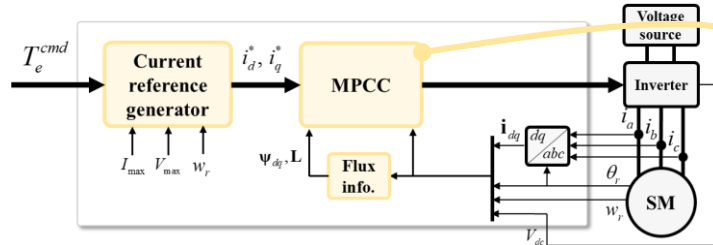
$$\begin{aligned} e_d &\leftarrow e_d - K_{ad}(v_d^* - v_d) \\ e_q &\leftarrow e_q - K_{aq}(v_q^* - v_q) \end{aligned}$$

FF comp.

$$\begin{aligned} v_{d,ff}^* &= -w_r \hat{L}_q i_q \\ v_{q,ff}^* &= w_r (\hat{L}_d i_d + \hat{\lambda}_{pm}) \end{aligned}$$

Reference generator + MPCC [5]

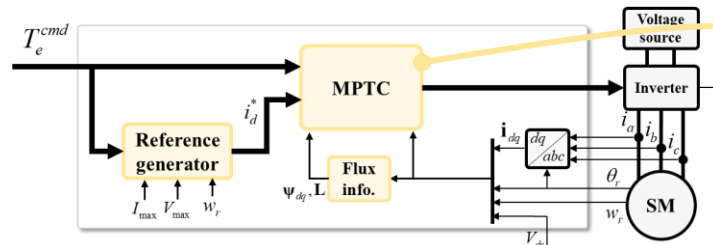
MPCC: Model Predictive Current Control



$$\begin{aligned} \min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \quad & \sum_{j=1}^{N_p} (i_d^* - i_{d,k+j})^2 + (i_q^* - i_{q,k+j})^2 + \alpha J_{p,k+j} \\ \text{s.t.} \quad & i_{d,k+j}^2 + i_{q,k+j}^2 \leq I_{\max}^2, \quad v_{d,k+j-1}^2 + v_{q,k+j-1}^2 \leq V_{\max}^2 \end{aligned}$$

Reference generator + MPTC [6]

MPTC: Model Predictive Torque Control



$$\begin{aligned} \min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \quad & \sum_{j=1}^{N_p} (T_e^{\text{cmd}} - T_{e,k+j})^2 + (i_d^* - i_{d,k+j})^2 + \alpha J_{p,k+j} \\ \text{s.t.} \quad & i_{d,k+j}^2 + i_{q,k+j}^2 \leq I_{\max}^2, \quad v_{d,k+j-1}^2 + v_{q,k+j-1}^2 \leq V_{\max}^2 \end{aligned}$$

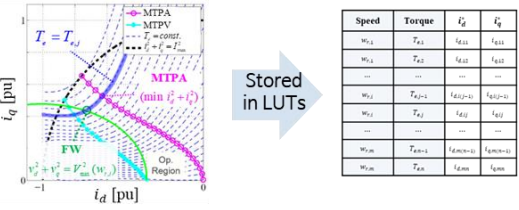
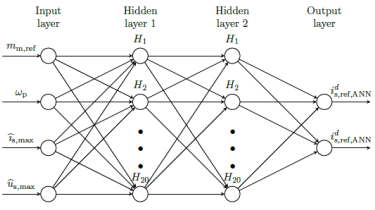
[4] S.-K. Sul, Control of electric machine drive systems. *John Wiley & Sons*, 2011, vol. 88.

[5] J. Rodriguez, et al., "Predictive current control of a voltage source inverter," *IEEE TIE*, vol. 54, no. 1, pp. 495 - 503, 2007.

[6] T. Englert and K. Graichen, "Nonlinear model predictive torque control of PMSMs for high performance applications," *CEP Eng. Prac.*, vol. 81, pp. 43 - 54, 2018.

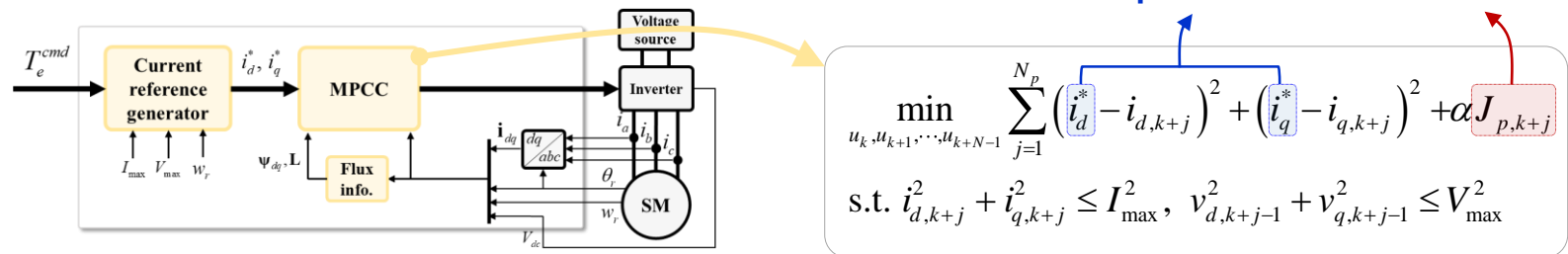
Existing schemes for Torque control of SMs

- Most schemes rely on a reference generator.
 - However, implementing a reference generator **requires a lot of resources**.

Experimental [3]	Analytical[1]/numerical[2]	ANN-based [7]
 <p>Stored in LUTs</p> <p>➤ Experiment time (e.g., 3 months)</p>	<p>Case 1. $T_e^* \leq T_e^{\max}$</p> $\min i_d^2 + i_q^2$ <p>subject to $T_e^* = (k_1 + k_2 i_d) i_q$,</p> $v_d^2 + v_q^2 \leq V_{\max}^2$ <p>➤ Computation time [7] (e.g., ~400 us)</p> <p>Case 2. $T_e^* > T_e^{\max}$</p> $\max \text{sgn}(T_e) T_e$ <p>subject to $v_d^2 + v_q^2 \leq V_{\max}^2$,</p> $i_d^2 + i_q^2 \leq I_{\max}^2$	 <p>➤ Training [7]</p>

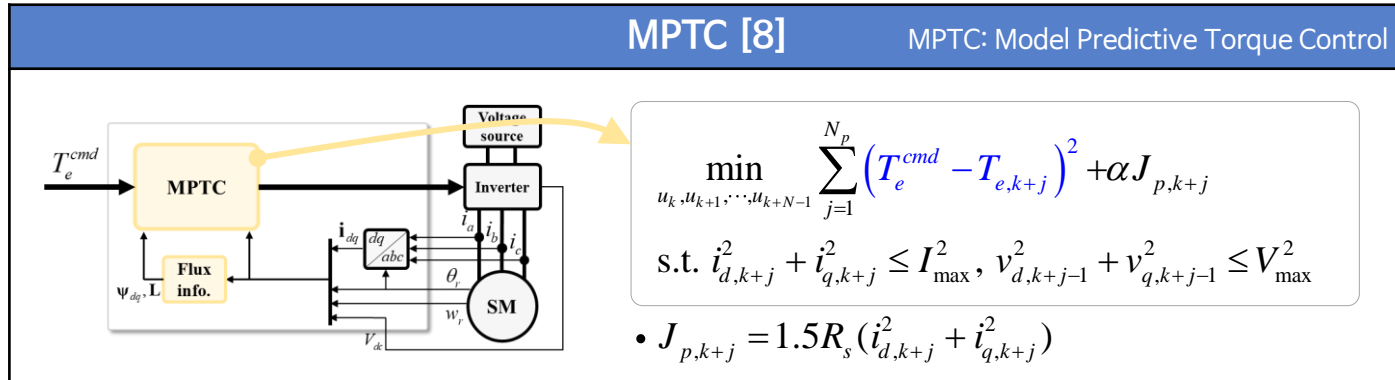
- In addition, using a reference generator **restricts MPC** from fully utilizing the DOF.

– E.g.,

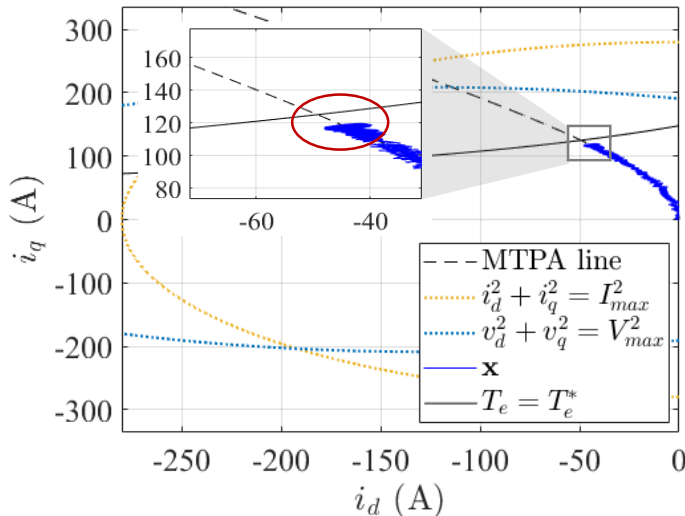


Existing schemes for Torque control of SMs

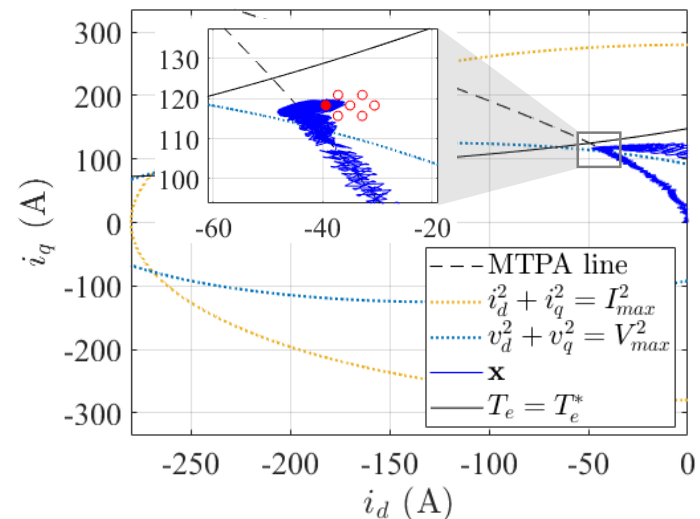
- One of the few attempts not to use a reference generator.



Limitation 1. Trade-off b/w the objectives



Limitation 2. Instability in the FW region



FW: Flux-weakening

Research Objectives

- Develop a MPTC scheme for SMs that
 - Does not rely on a reference generator
 - But guarantees optimal operation under all operating regions
 - Can be implemented based on both the FCS and CCS
 - FCS: finite control set, CCS: continuous control set
 - Can be implemented with various performance indices (J_p)
- } Primary objectives
- } Secondary objectives

[5] J. Rodriguez, et al., "Predictive current control of a voltage source inverter," *IEEE TIE*, vol. 54, no. 1, pp. 495 - 503, 2007.

[6] T. Englert and K. Graichen, "Nonlinear model predictive torque control of PMSMs for high performance applications," *Control Eng. Prac.*, vol. 81, pp. 43 - 54, 2018.

[7] L. Samaranayake and S. Longo, "Degradation control for electric vehicle machines using nonlinear model predictive control," *IEEE TCST*, vol. 26, no. 1, pp. 89 - 101, 2017.

2. Proposed MPTC

Problem Formulation

- Overcome two limitations of the existing MPTC [8] by modifying it.
 - **Limitation 1.** Trade-off in the objective function
 - Resolved by moving the torque error term to the equality constraint.
 - **Limitation 2.** Instability in the FW region
 - Resolved by modifying the voltage constraint.

Existing MPTC [8]

$$\min (T_{e,k+1}^{cmd} - T_{e,k+1})^2 + \alpha J_{p,k+1}$$

s.t. system model,

$$I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0.$$

Proposed MPTC

$$\min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) = J_{p,k+1}$$

$$\text{s.t. } c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0,$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0.$$

Problem Formulation

▪ Overcome two limitations of the existing MPTC [8] by modifying it.

• Limitation 1. Trade-off in the objective function

- Resolved by moving the torque error term to the equality constraint.
- *Is it possible to consider the tracking error as an equality constraint?*

Yes, with a solver that allows tolerance during the optimization process.

Existing MPTC [8]

$$\min \left(T_{e,k+1}^{cmd} - T_{e,k+1} \right)^2 + \alpha J_{p,k+1}$$

s.t. system model,

$$I_{\max}^2 - \left\| \mathbf{i}_{dq,k+1} \right\|^2 \geq 0,$$

$$V_{\max}^2 - \left\| \mathbf{v}_{dq,k} \right\|^2 \geq 0.$$

Proposed MPTC

$$\min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) = J_{p,k+1}$$

$$\text{s.t. } c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0,$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \left\| \mathbf{i}_{dq,k+1} \right\|^2 \geq 0,$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \left\| \mathbf{v}_{dq,k+1} \right\|^2 \geq 0.$$

Problem Formulation

- Overcome two limitations of the existing MPTC [8] by modifying it.
 - **Limitation 2.** Instability in the FW region
 - Resolved by modifying the voltage constraint.

System model

$$\begin{aligned} d\lambda_d / dt &= -R_s i_d + w_r \lambda_q + v_d, \\ d\lambda_q / dt &= -R_s i_q - w_r \lambda_d + v_q. \end{aligned}$$

Existing MPTC [8]

$$\min (T_{e,k+1}^{cmd} - T_{e,k+1})^2 + \alpha J_{p,k+1}$$

s.t. system model,

$$I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$V_{\max}^2 - \|\mathbf{v}_{dq,k}\|^2 \geq 0.$$

System model in steady-states

$$0 = -R_s i_d + w_r \lambda_q + \tilde{v}_d,$$

$$0 = -R_s i_q - w_r \lambda_d + \tilde{v}_q.$$

Proposed MPTC

$$\min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) = J_{p,k+1}$$

$$\text{s.t. } c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0,$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0.$$

Problem Formulation

- Overcome two limitations of the existing MPTC [8] by modifying it.
 - Limitation 2. Instability in the FW region
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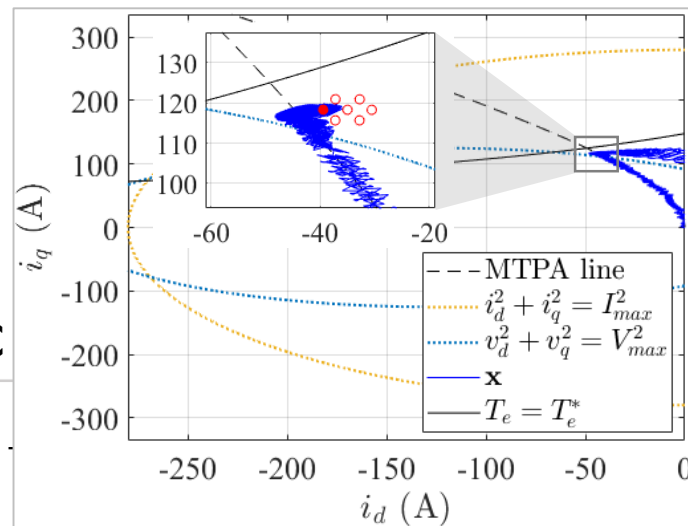
Existing MPTC

$$\min (T_{e,k+1}^{cmd} - T_{e,k+1})^2$$

s.t. system model,

$$I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$V_{\max}^2 - \|\mathbf{v}_{dq,k}\|^2 \geq 0.$$



Proposed MPTC

$$J_{p,k+1}$$

$$c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0,$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0.$$

Solver

- The **augmented Lagrangian method** is adopted to solve the proposed MPTC.
 - Concept

Nonlinear Programming (NLP)

$$\begin{aligned} \min_x & J_p(x) \\ \text{s.t. } & c_i(x) = 0, i \in E, \\ & c_i(x) \geq 0, i \in I. \end{aligned}$$



Minimize the augmented Lagrangian function instead

$$\begin{aligned} \min_x & L_A(x, \lambda_k; \mu_k) \\ & = J_p(x) - \sum_{i \in E} \lambda_i c_i(x) + \frac{1}{2\mu} \sum_{i \in E} \lambda_i c_i^2(x) + \sum_{i \in I} \psi(c_i(x), \lambda_i; \mu) \end{aligned}$$

Multiplier update: $\lambda_{i,k+1} = \lambda_{i,k} - c_i(x_k) / \mu_k, i \in E,$
 $\lambda_{i,k+1} = \max(\lambda_{i,k} - c_i(x_k) / \mu_k, 0), i \in I.$

- Properties

- 1. Can handle NLP → Allow J_p to be any function (e.g., copper loss or inverter loss)
- 2. Allow tolerances to constraints
 - Allow the tracking error to be the equality constraint
 - Allow smooth transitions b/w operating regions (i.e., MTPA, FW, MC, MTPV)

$$\bullet \psi(c_i(x), \lambda_i; \mu) = \begin{cases} -\lambda_i c_i(x) + c_i^2(x) / (2\mu), & \text{if } c_i(x) - \mu \lambda_i \leq 0, \\ -\mu \lambda_i^2 / 2, & \text{otherwise.} \end{cases}$$

Solver

Implementation

Proposed MPTC based on CCS

$$\mathbf{V}_c = \{ \mathbf{v}_{dq,k} \in \mathbb{R}^2 \mid \mathbf{v}_{dq,k} \text{ is within the limit hexagon} \}$$

Algorithm 1: CCS-based GMPTC solver

```

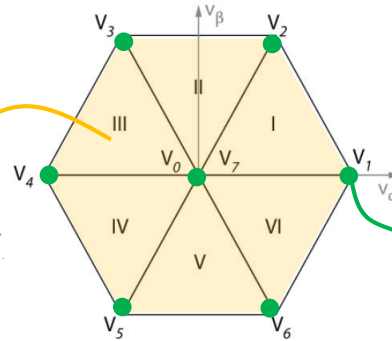
Determine the expression of  $J_p$ ;
Choose positive parameters  $\mu_t, \mu_{\bar{v}}, \mu_i$ ;
Set  $\lambda_{t,1} \leftarrow 0, \lambda_{i,1} \leftarrow 0, \lambda_{\bar{v},1} \leftarrow 0, \mathbf{v}_{dq,0}^* \leftarrow 0$ ;
for  $k = 1, 2, 3, \dots$  do
    Input:  $T_{e,k+1}^{cmd}$ 
    Set  $\mathbf{v}_{dq,k}^* \leftarrow \mathbf{v}_{dq,k-1}^*$ ;
    repeat
         $\mathbf{v}_{dq,k}^* \leftarrow \mathbf{v}_{dq,k}^* - \nabla^2 L_{A1}^{-1}(\mathbf{v}_{dq,k}^*, \lambda_k, \mu) \nabla L_{A1}(\mathbf{v}_{dq,k}^*, \lambda_k, \mu)$ ;
    until  $\|\nabla L_{A1}(\mathbf{v}_{dq,k}^*, \lambda_k, \mu)\| \leq \tau$ ;
     $\lambda_{t,k+1} = \lambda_{t,k} - c_t(\mathbf{v}_{dq,k}^*)/\mu_t$ ;
     $\lambda_{i,k+1} = \max(\lambda_{i,k} - c_i(\mathbf{v}_{dq,k}^*)/\mu_i, 0)$ ;
     $\lambda_{\bar{v},k+1} = \max(\lambda_{\bar{v},k} - c_{\bar{v}}(\mathbf{v}_{dq,k}^*)/\mu_{\bar{v}}, 0)$ ;
    Limit  $\mathbf{v}_{dq,k}^*$  within the CCS;
    Output:  $\mathbf{v}_{dq,k}^*$ 

```

Tuning
parameters

Any combinations of

- Copper loss
- Iron loss
- Inverter loss
- Temperature
- ...



Proposed MPTC based on FCS

$$\mathbf{V}_f = \{ \mathbf{v}_{dq,k}^{(0)}, \mathbf{v}_{dq,k}^{(1)}, \mathbf{v}_{dq,k}^{(2)}, \mathbf{v}_{dq,k}^{(3)}, \mathbf{v}_{dq,k}^{(4)}, \mathbf{v}_{dq,k}^{(5)}, \mathbf{v}_{dq,k}^{(6)}, \mathbf{v}_{dq,k}^{(7)} \}$$

Algorithm 2: FCS-based GMPTC solver

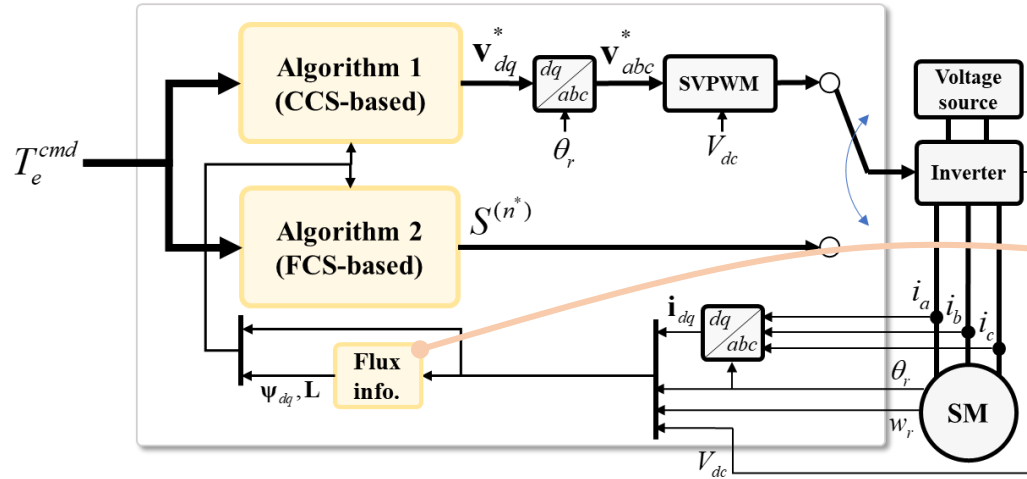
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Determine the expression of  $J_p$ ;
Choose positive parameters  $\mu_t, \mu_{\bar{v}}, \mu_i$ ;
Set  $\lambda_{t,1} \leftarrow 0, \lambda_{i,1} \leftarrow 0, \lambda_{\bar{v},1} \leftarrow 0, \mathbf{v}_{dq,0}^* \leftarrow 0$ ;
for  $k = 1, 2, 3, \dots$  do
    Input:  $T_{e,k+1}^{cmd}$ 
    Compute  $L_{A1}(\mathbf{v}_{dq,k}^{(n)}, \lambda_k, \mu)$  for  $n = 0, \dots, 7$ ;
     $n^* = \arg \min_n L_{A1}(\mathbf{v}_{dq,k}^{(n)}, \lambda_k, \mu)$ ;
     $\lambda_{t,k+1} = \lambda_{t,k} - c_t(\mathbf{v}_{dq,k}^{(n^*)})/\mu_t$ ;
     $\lambda_{i,k+1} = \max(\lambda_{i,k} - c_i(\mathbf{v}_{dq,k}^{(n^*)})/\mu_i, 0)$ ;
     $\lambda_{\bar{v},k+1} = \max(\lambda_{\bar{v},k} - c_{\bar{v}}(\mathbf{v}_{dq,k}^{(n^*)})/\mu_{\bar{v}}, 0)$ ;
    Set  $\mathbf{v}_{dq,k}^* \leftarrow \mathbf{v}_{dq,k}^{(n^*)}$ ;
    Output:  $S^{(n^*)}$ 

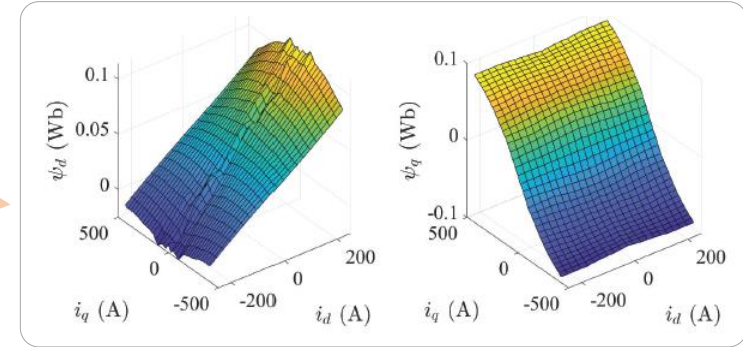
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Schematic Diagrams

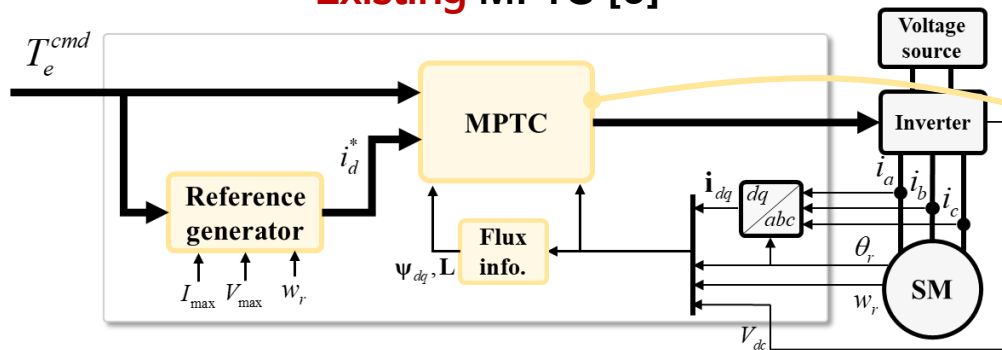
Proposed MPTC



Required information



Existing MPTC [6]



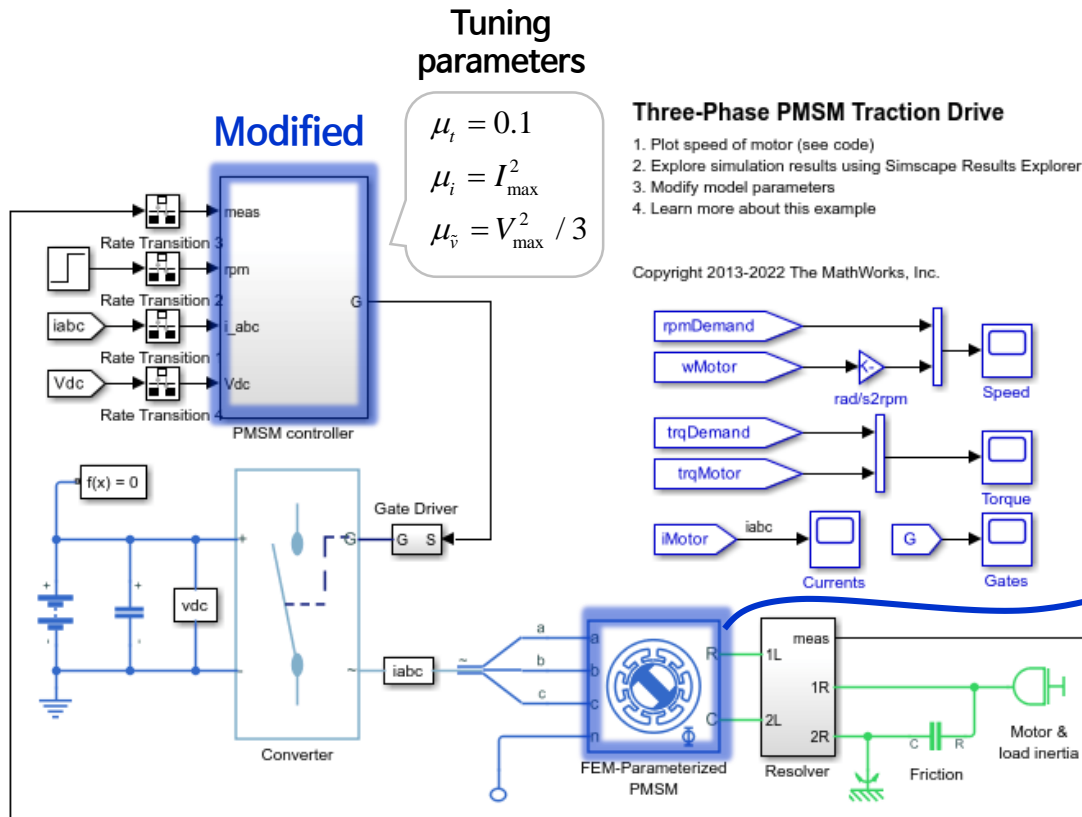
$$\min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \sum_{j=1}^{N_p} (T_e^{cmd} - T_{e,k+j})^2 + (i_d^* - i_{d,k+j})^2 + \alpha J_{p,k+j}$$

$$\text{s.t. } i_{d,k+j}^2 + i_{q,k+j}^2 \leq I_{\max}^2, v_{d,k+j-1}^2 + v_{q,k+j-1}^2 \leq V_{\max}^2$$

3. Validation

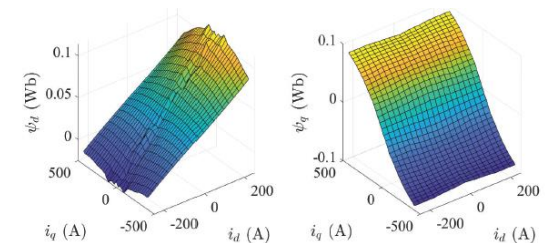
Simulation Setup

- Use MATLAB/SIMULINK example and modify the control part.



50 kW IPMSM model

Base speed	3200 RPM
Maximum torque	150 Nm
Rotor inertia	0.1234 kg·m ²
I_{max}	250 A
V_{dc}	325 V
R_s	10 mΩ
P	8
T_s	20 μs

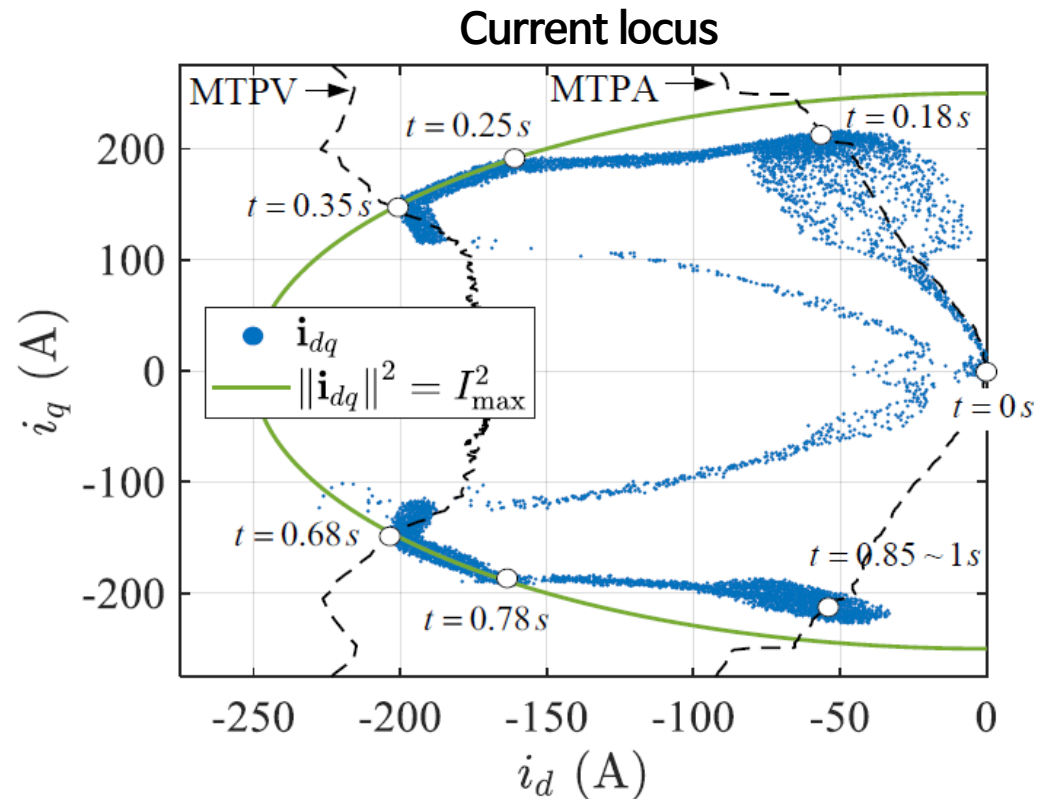
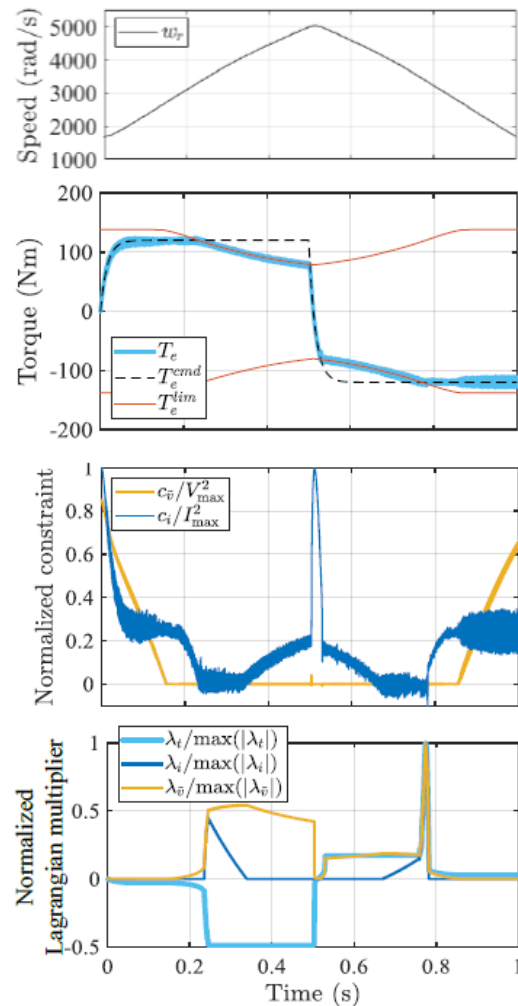


Simulation Results

Simulation 1. Validation of the Proposed MPTC

- Based on the CCS ($T_s = 50\mu s$)

$$J_p = P_{cu}$$



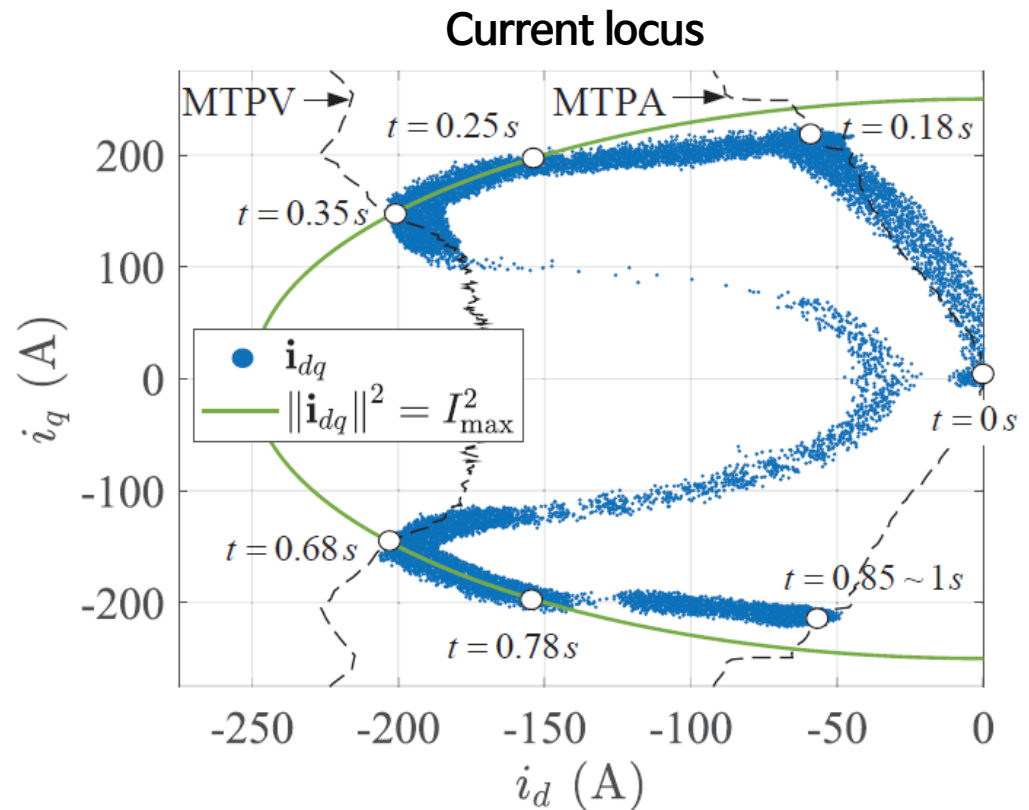
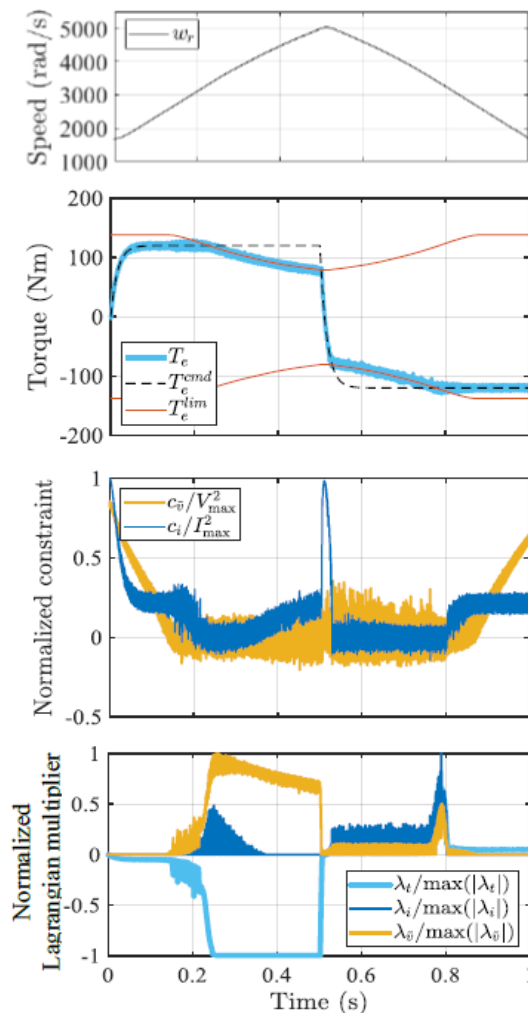
➤ Guarantee optimal operations in all regions.

Simulation Results

Simulation 1. Validation of the Proposed MPTC

- Based on the FCS ($T_s = 20\mu s$)

$$J_p = P_{cu}$$



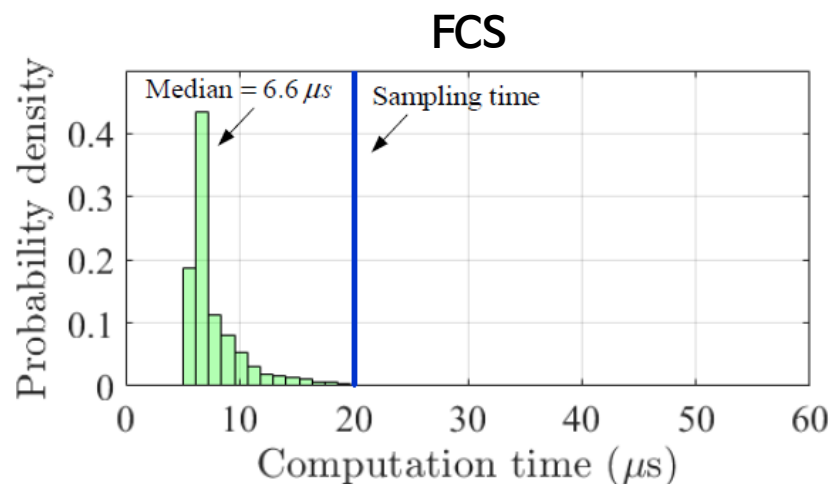
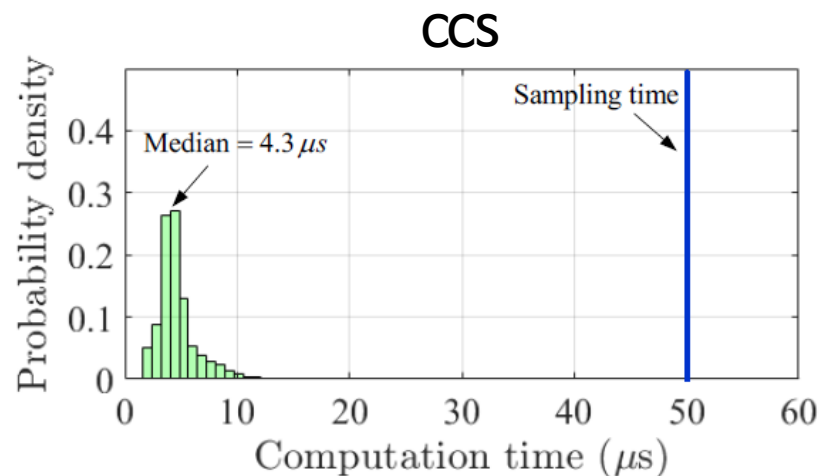
➤ Guarantee optimal operations in all regions.

Simulation Results

Simulation 1. Validation of the Proposed MPTC

- Computation time

$$J_p = P_{cu}$$

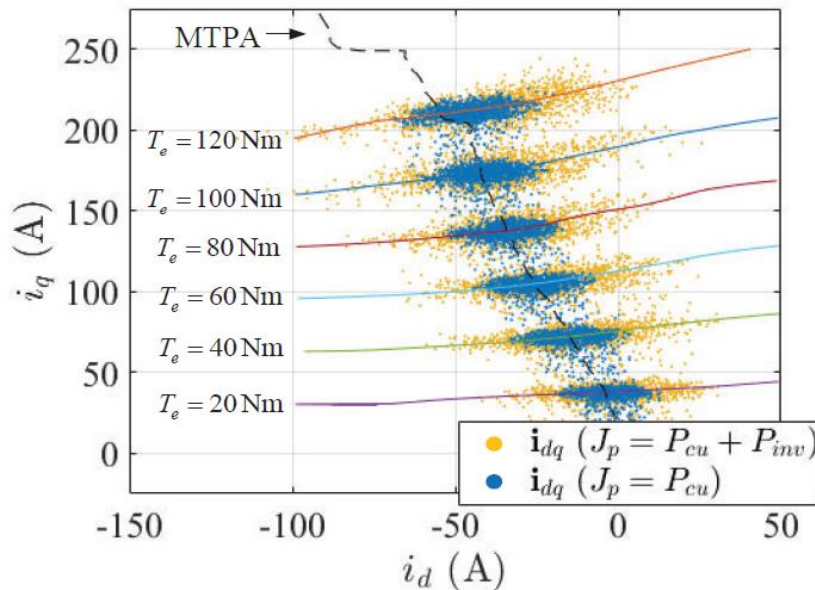


Simulation Results

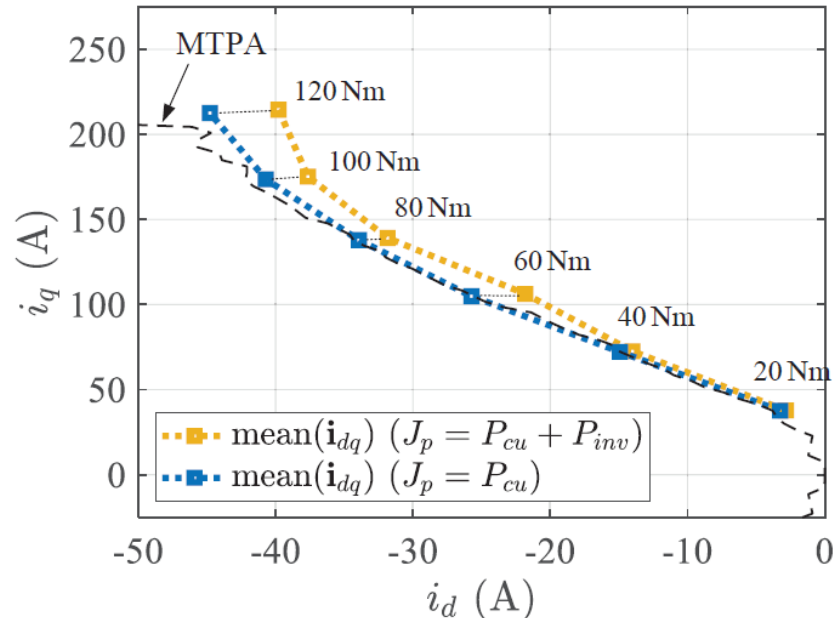
$$J_p = P_{cu} \quad \text{vs.} \quad J_p = P_{cu} + P_{inv}$$

- Simulation 2. Effects of Using Different Performance Indices
 - Based on the FCS ($T_s = 20\mu s$)

Current locus



Mean current locus

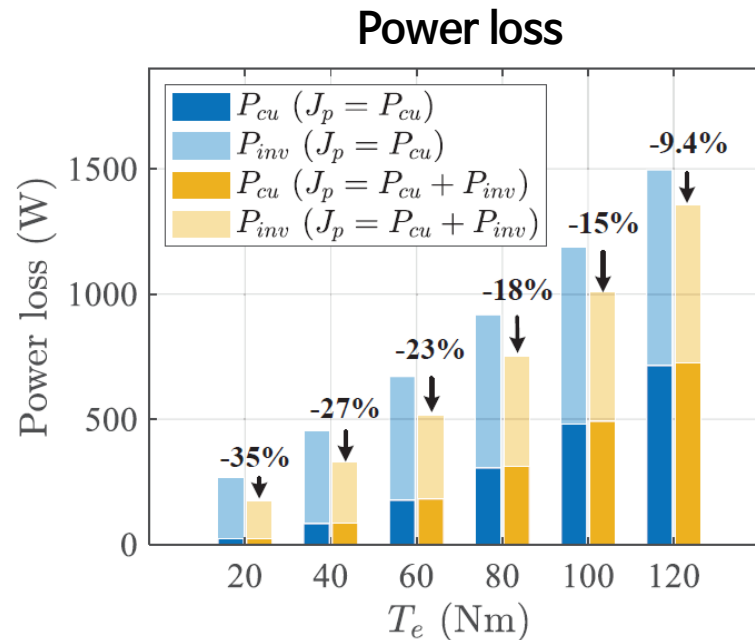


- Different performance indices result in different current loci.

Simulation Results

$$J_p = P_{cu} \quad \text{vs.} \quad J_p = P_{cu} + P_{inv}$$

- Simulation 2. Effects of Using Different Performance Indices
 - Based on the FCS ($T_s = 20\mu s$)



- Using a different performance index can improve performance significantly.

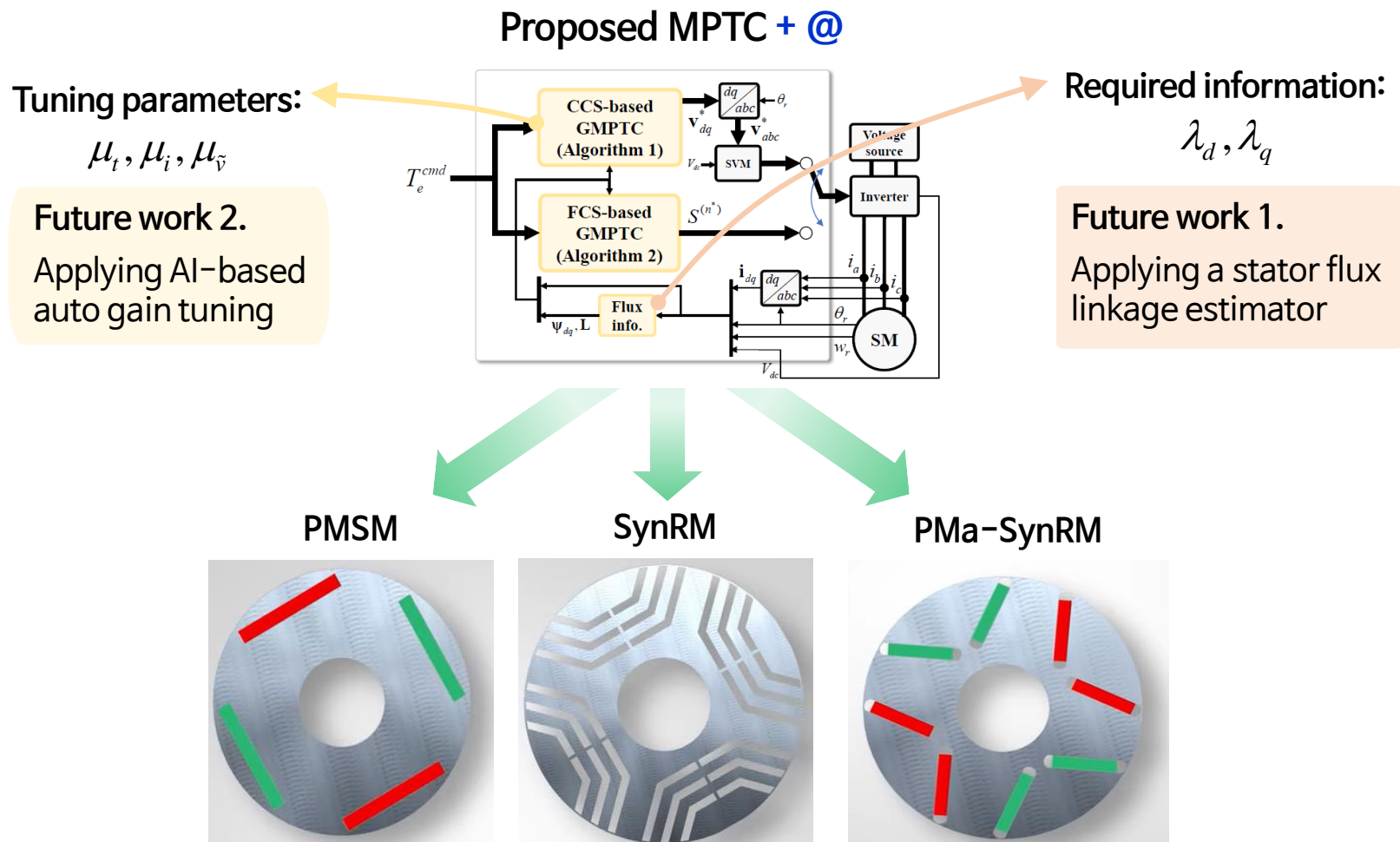
4. Conclusion & Further Work

Conclusion

- **A novel MPTC scheme** was presented that
 - Does not rely on a reference generator,
 - Guarantees optimal operation under all operating regions.
- **Two key ideas** were
 - Moving the torque error term to the equality constraint,
 - Redefining the voltage constraint.
- The proposed MPTC is **a general approach** in that
 - It can be implemented **without a reference generator**,
 - It can be implemented based on **both the FCS and CCS**,
 - **Various performance indices** can be used.

Further Work – Toward Intelligent SM controller

- How to control SMs without prior information and offline tuning?





Appendix

When the torque command (T_e^{cmd}) is not achievable

- Need to solve a different MPTC problem

When T_e^{cmd} is **achievable**

$$\min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) = J_{p,k+1} \quad (10a)$$

$$\text{s.t. } c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0, \quad (10b)$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0, \quad (10c)$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\tilde{\mathbf{v}}_{dq,k+1}\|^2 \geq 0. \quad (10d)$$

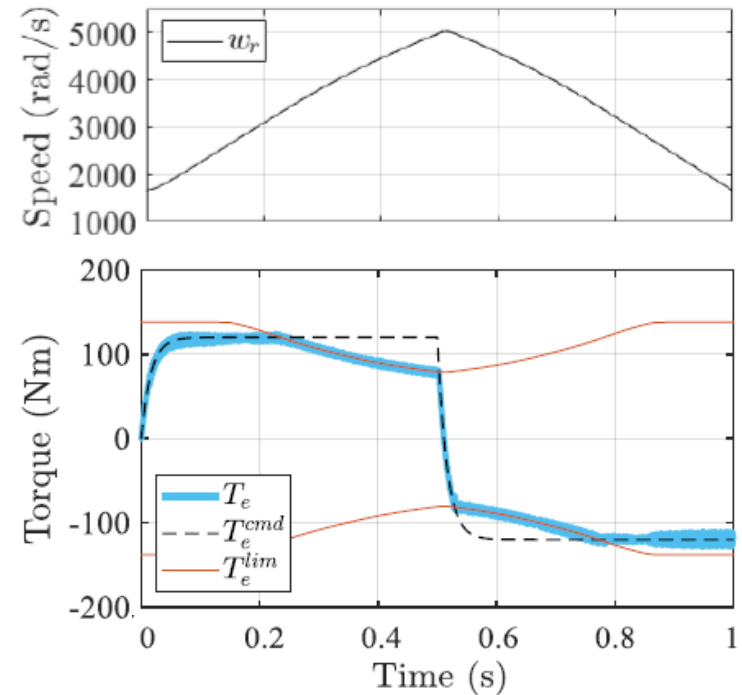
When T_e^{cmd} is **not achievable**

$$\min_{\mathbf{v}_{dq,k}} \text{sgn}(T_{e,k+1}^{cmd}) c_t(\mathbf{v}_{dq,k}) \quad (11a)$$

$$\text{s.t. } c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0, \quad (11b)$$

$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\tilde{\mathbf{v}}_{dq,k+1}\|^2 \geq 0. \quad (11c)$$

Simulation result



Modified voltage constraint

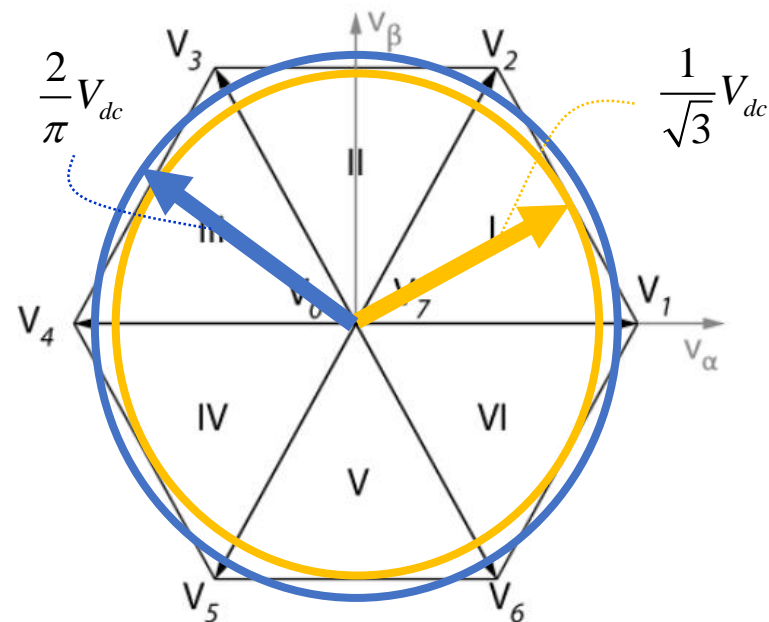
Proposed MPTC

$$\min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) = J_{p,k+1}$$

$$\text{s.t. } c_t(\mathbf{v}_{dq,k}) = T_{e,k+1}^{cmd} - T_{e,k+1} = 0,$$

$$c_i(\mathbf{v}_{dq,k}) = I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0,$$

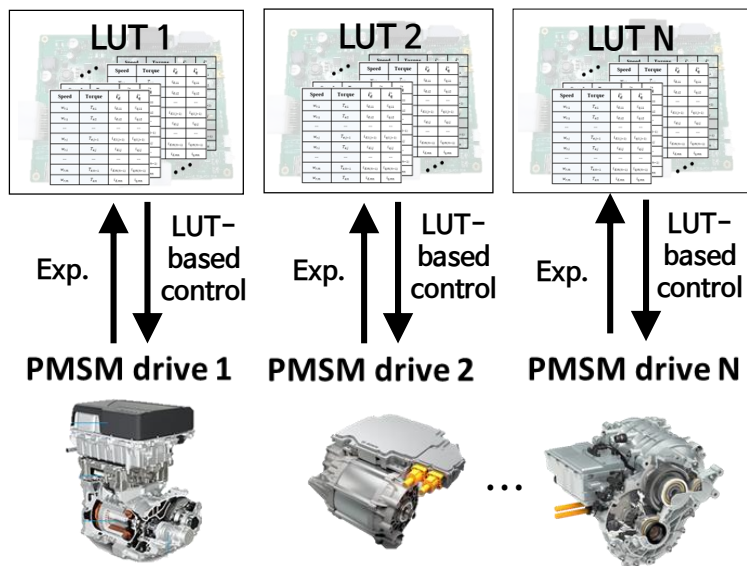
$$c_{\tilde{v}}(\mathbf{v}_{dq,k}) = V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0.$$



Intelligent SM controller

AS-IS

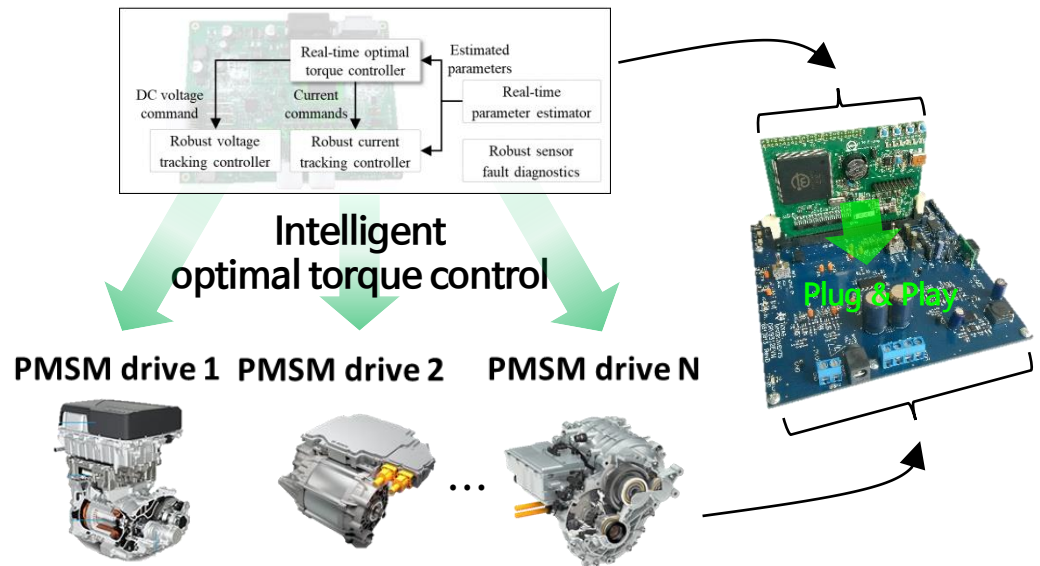
Experiment-based controllers



- Require extensive experiments (costs)
- Different controllers for different SMs

TO-BE

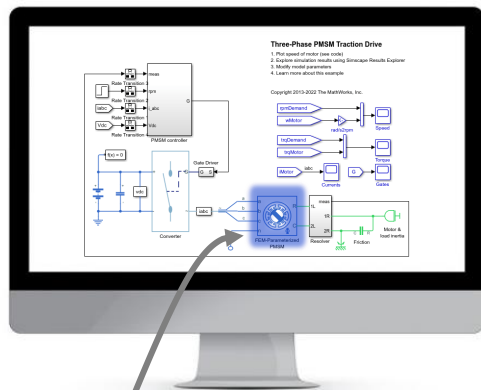
Intelligent SM controller



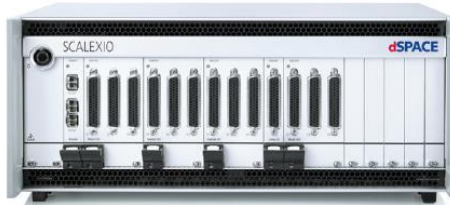
- Control any SM drives w/o prior experimental data
- Significant reduction in development costs

Experimental setup for AI-based automatic gain tuning

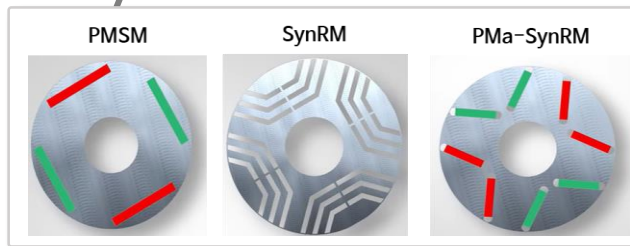
**MATLAB/SIMULINK
(SM drive model)**



**Real-time system
(SCALEXIO, dSPACE)**

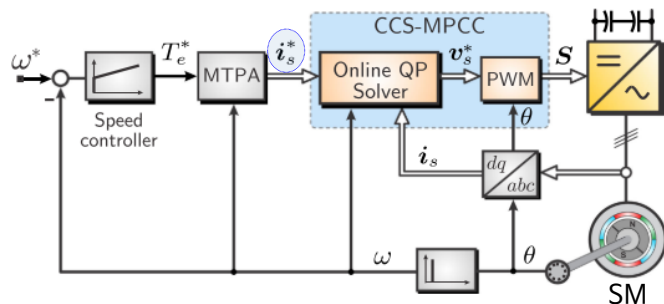


AI workstation



MPC application to SM torque control

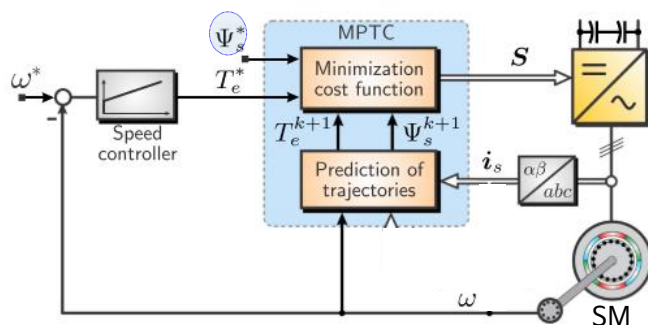
- Two control schemes [11]
 - 1) Model predictive current control (MPCC) w/ current reference generator



Primary objective

$$\min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \sum_{j=0}^{N_p-1} \left(i_d^* - i_{d,k+j+1} \right)^2 + \left(i_q^* - i_{q,k+j+1} \right)^2$$

- 2) Model predictive torque control (MPTC) w/ flux reference generator



Primary objective

$$\min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \sum_{j=0}^{N_p-1} \left(T_e^* - T_{e,k+j+1} \right)^2 + \rho \left(\left\| \lambda_{dq}^* \right\| - \left\| \lambda_{dq,k+j+1} \right\| \right)^2$$

SM: Synchronous Machine

[11] J. Rodriguez *et al.*, "Latest advances of model predictive control in electrical drives—Part II: Applications and benchmarking with classical control methods," *IEEE TPE*, 2021.

Optimal current reference generation

Problem statement

Case 1. $T_e^* \leq T_e^{\max}$

$$\min i_d^2 + i_q^2$$

$$\text{subject to } T_e^* = (k_1 + k_2 i_d) i_q,$$

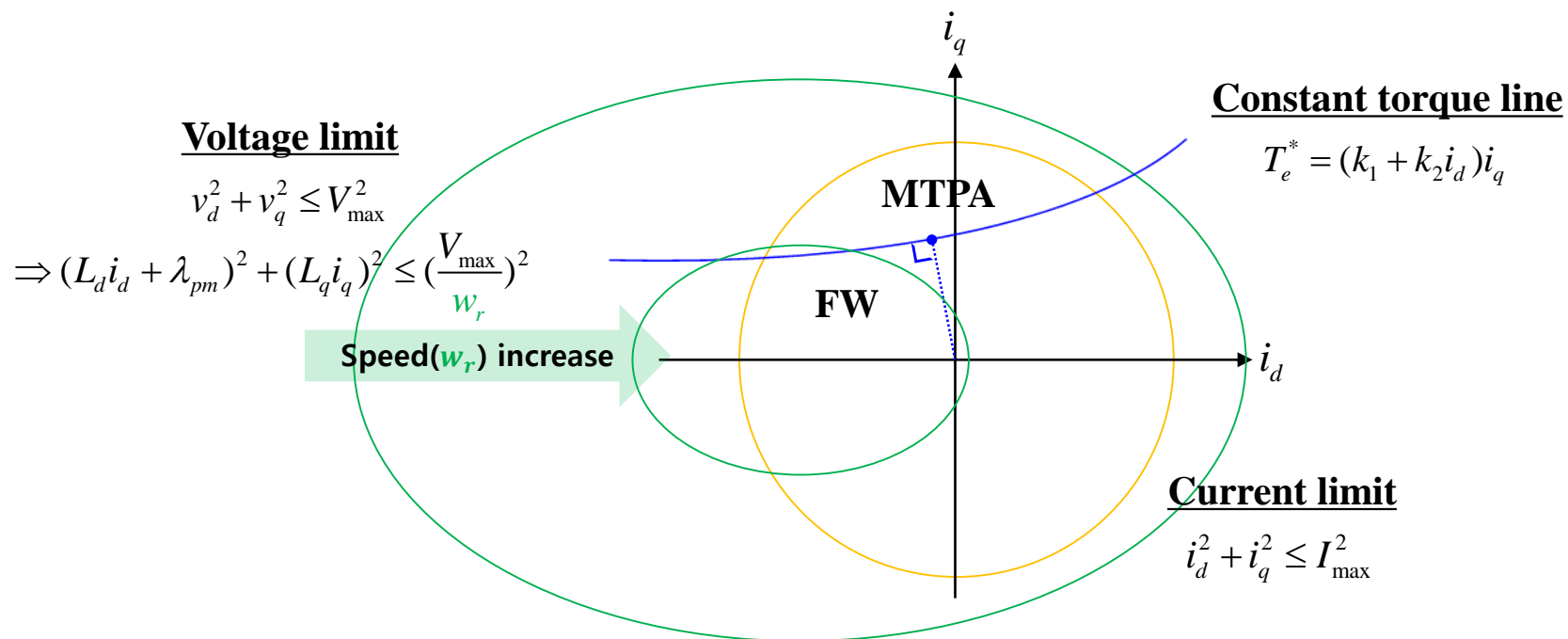
$$v_d^2 + v_q^2 \leq V_{\max}^2.$$

Case 2. $T_e^* > T_e^{\max}$

$$\max \pm T_e$$

$$\text{subject to } v_d^2 + v_q^2 \leq V_{\max}^2,$$

$$i_d^2 + i_q^2 \leq I_{\max}^2.$$



Optimal current reference generation

■ Problem statement

Case 1. $T_e^* \leq T_e^{\max}$

$$\min i_d^2 + i_q^2$$

$$\text{subject to } T_e^* = (k_1 + k_2 i_d) i_q,$$

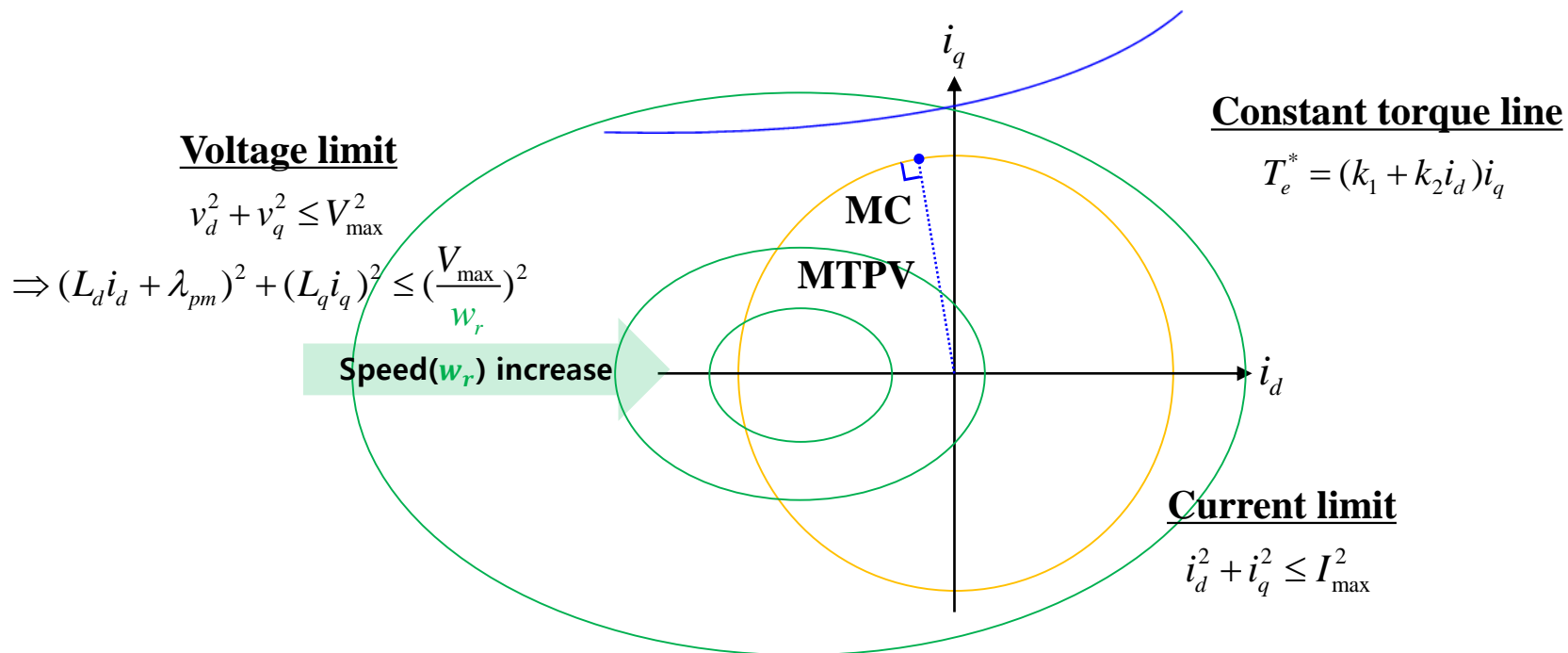
$$v_d^2 + v_q^2 \leq V_{\max}^2.$$

Case 2. $T_e^* > T_e^{\max}$

$$\max \pm_t T_e$$

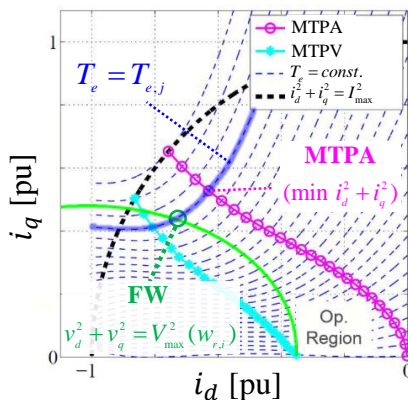
$$\text{subject to } v_d^2 + v_q^2 \leq V_{\max}^2,$$

$$i_d^2 + i_q^2 \leq I_{\max}^2.$$



Optimal current reference generation

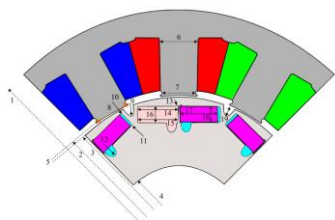
- Experimental solution [12]
 - Experimentally find solutions and store them in a LUT



stored
in a LUT

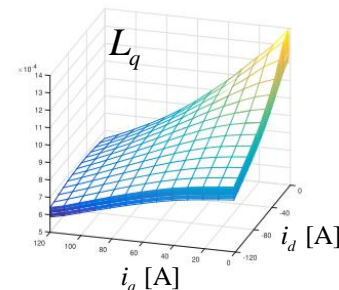
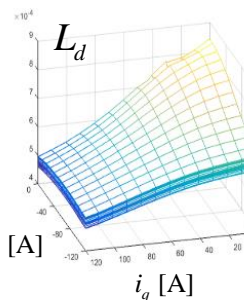
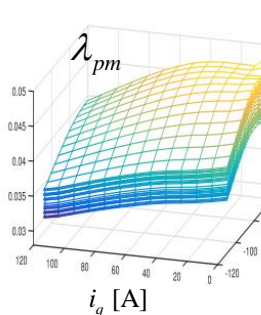
Speed	Torque	i_d^*	i_q^*
$w_{r,1}$	$T_{e,1}$	$i_{d,11}$	$i_{q,11}$
$w_{r,1}$	$T_{e,2}$	$i_{d,12}$	$i_{q,12}$
...
$w_{r,i}$	$T_{e,j-1}$	$i_{d,i(j-1)}$	$i_{q,i(j-1)}$
$w_{r,i}$	$T_{e,j}$	$i_{d,ij}$	$i_{q,ij}$
...
$w_{r,m}$	$T_{e,n-1}$	$i_{d,m(n-1)}$	$i_{q,m(n-1)}$
$w_{r,m}$	$T_{e,n}$	$i_{d,mn}$	$i_{q,mn}$

- Require extensive experiments
- A specific solution just for one product,
which cannot handle parameter deviations resulting from tolerances



design variables

manufacturing
tolerances [13]



[1] B. Gallert, et al., "Maximum efficiency control strategy of PM traction machine drives in GM hybrid and electric vehicles," in IEEE ECCE, 2017.

[13] H. Khreis, et al., "Sensitivity analysis on electrical parameters for permanent magnet synchronous machine manufacturing tolerances in EV and HEV," in ITEC, 2016.

Synchronous Machines in Electric Vehicles

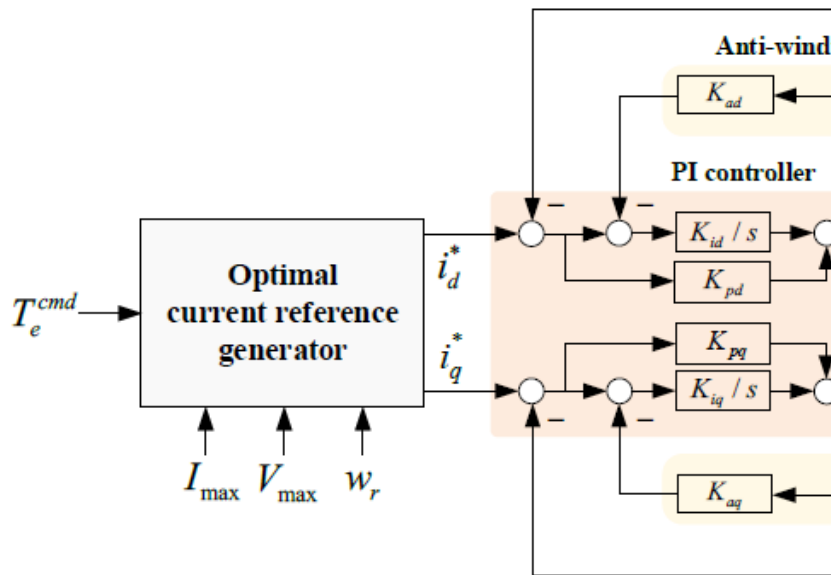
- Example of EVs on the market from 2010 to 2020 and their machines [14]

TABLE 1. Example of EVs on the market from 2010 to 2020, including their model, motor categories, and power.

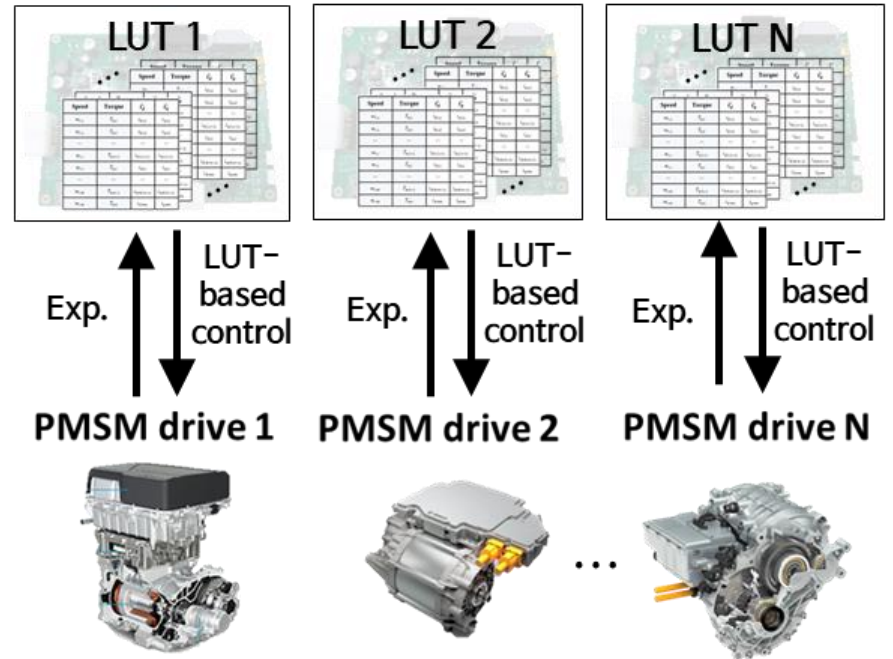
EV model	Power(kW)	Motor	Year
Mahindra e2o Plus	19-30	IM	2016
Renault Kangoo ZE	44	PMSM	2011
Mitsubishi i-MiEV	47	PM	2010
Volkswagen E-up	60	PMSM	2019
Renault Zoe	65	PMSM	2012
LandRover	70	SRM	2013
Renault Fluence Z.E.	70	PMSM	2012
Nissan Leaf	80	PMSM	2010
BJEV EC5	80	PMSM	2019
Hyundai Ioniq Electric	88	PMSM	2016
Hyundai Kona	88-150	PMSM	2018
BYD E6	90	PMSM	2014
BMW i3	125	PMSM	2013
Xpeng G3	139	PMSM	2018
Mercedes-Benz EQC	150*2	IM	2019
BJEV EU5	160	PMSM	2018
Tesla Model X	193-375	IM	2015
Tesla Model 3	211-340	PMSM	2020
Tesla Model S	235-568	IM	2012
NIO EC6	320	PMSM	2020
NIO ES6	320	PMSM	2020

Existing schemes for Torque control of SMs

- 1. Conventional scheme [4]
: Current reference generator + PI-



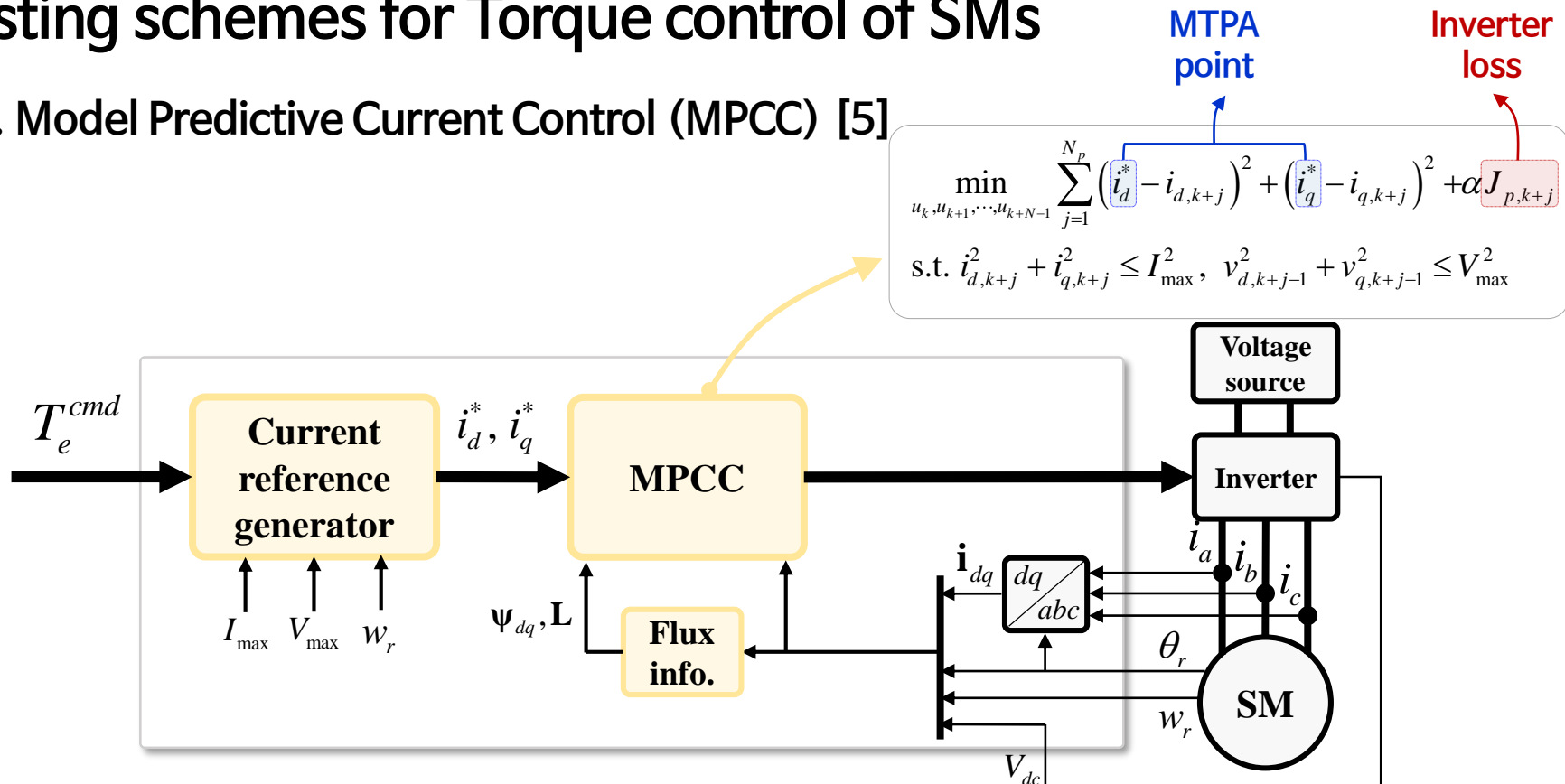
Experiment-based controllers



- Involve a number of components
- Require a lot of time (e.g., 3 months) to finalize this scheme

Existing schemes for Torque control of SMs

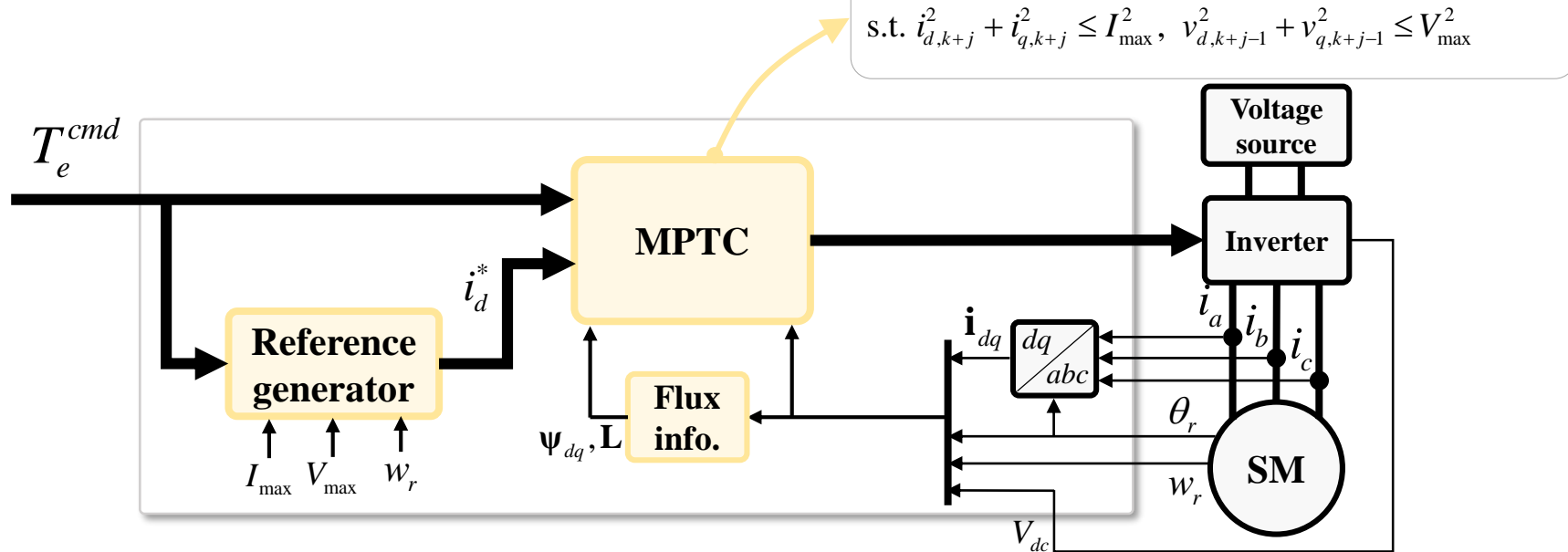
2. Model Predictive Current Control (MPCC) [5]



- Guarantee **improved current tracking performance** *But torque?*
- Still **rely on a current reference generator**
 - **Problem 1.** Need to solve **another optimization problem** other than MPC
 - **Problem 2.** Two optimization problems **handle the DOF separately**

Existing schemes for Torque control of SMs

- 3. Model Predictive Torque Control (MPTC)
 - 1) w/ reference generator [6]



- Guarantee improved torque tracking performance
- Still rely on a current reference generator

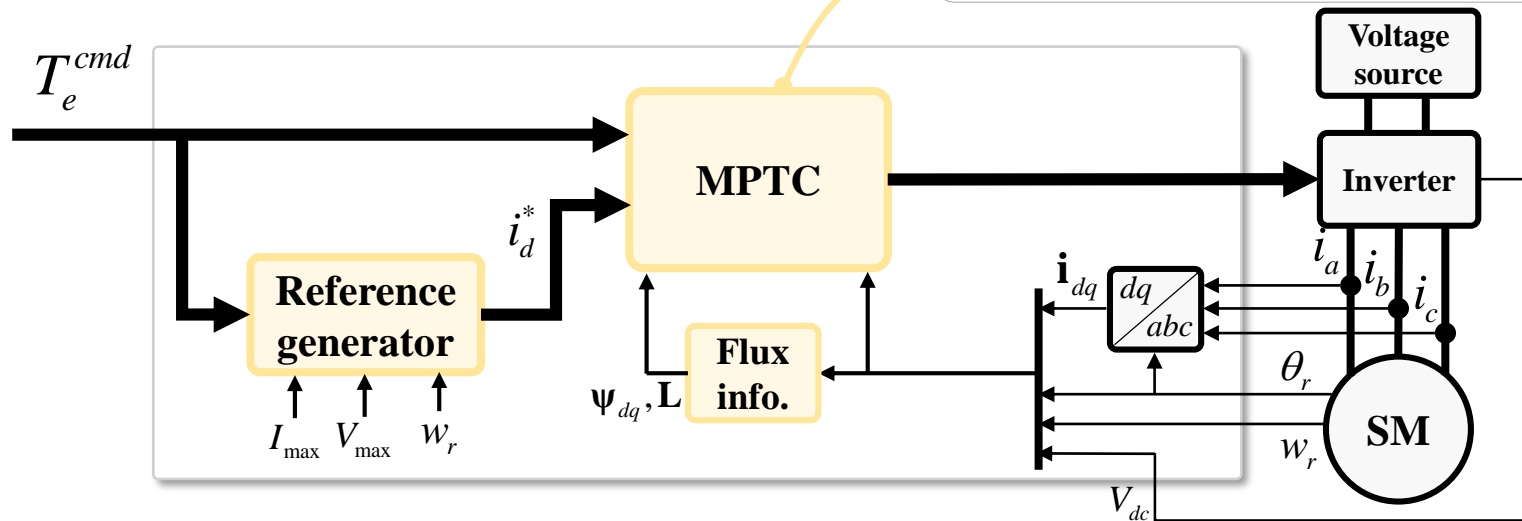
Existing schemes for Torque control of SMs

3. Model Predictive Torque Control (MPTC)

1) w/ reference generator [6]

$$\min_{u_k, u_{k+1}, \dots, u_{k+N-1}} \sum_{j=1}^{N_p} (T_e^{cmd} - T_{e,k+j})^2 + (i_d^* - i_{d,k+j})^2 + \alpha J_{p,k+j}$$

$$\text{s.t. } i_{d,k+j}^2 + i_{q,k+j}^2 \leq I_{\max}^2, \quad v_{d,k+j-1}^2 + v_{q,k+j-1}^2 \leq V_{\max}^2$$



- Guarantee **improved torque tracking performance**
- Still **rely on a current reference generator**
 - **Problem 1.** Need to solve **another optimization problem** other than MPC
 - **Problem 2.** Two optimization problems **handle the DOF separately**

Simulation Results

$$J_p = P_{cu} \quad \text{vs.} \quad J_p = P_{cu} + P_{inv}$$

- Simulation 2. Effects of Using Different Performance Indices
 - Discussion

Proposed MPTC

$$\begin{aligned} \min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) &= P_{cu,k+1} + P_{inv,k+1} \\ \text{s.t. } c_t(\mathbf{v}_{dq,k}) &= T_{e,k+1}^{cmd} - T_{e,k+1} = 0, \\ c_i(\mathbf{v}_{dq,k}) &= I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0, \\ c_{\tilde{v}}(\mathbf{v}_{dq,k}) &= V_{\max}^2 - \|\mathbf{v}_{dq,k+1}\|^2 \geq 0. \end{aligned}$$

➤ The MPTC fully utilizes the DOF.

Existing MPTC [5]

$$\begin{aligned} \min_{\mathbf{v}_{dq,k}} J_p(\mathbf{v}_{dq,k}) &= \|\mathbf{i}_{dq,k+1} - \mathbf{i}_{dq}^{MTPA}\|^2 + \alpha P_{inv,k+1} \\ \text{s.t. } c_i(\mathbf{v}_{dq,k}) &= I_{\max}^2 - \|\mathbf{i}_{dq,k+1}\|^2 \geq 0, \\ c_{\tilde{v}}(\mathbf{v}_{dq,k}) &= V_{\max}^2 - \|\mathbf{v}_{dq,k}\|^2 \geq 0. \end{aligned}$$

➤ The MPTC partially utilizes the DOF.

Research Objectives

Summary of Literature review

Concept	<div>Reference generator</div> <div>+</div> <div>MPCC [5] or MPTC [6]</div>	<div>MPTC [7]</div>
Property	<ul style="list-style-type: none"> ➤ Good tracking performance ➤ Two optimization problems 	<ul style="list-style-type: none"> ➤ Steady-state error, Instability ➤ One optimization problem

Research objectives

Develop a MPTC scheme for SMs that

- Does not rely on a reference generator
- But guarantees optimal operation under all operating regions
- Can be implemented based on both the FCS and CCS
- Can be implemented with various performance indices (J_p)
 - FCS: finite control set
 - CCS: continuous control set

} Primary objectives

} Secondary objectives

[5] J. Rodriguez, et al., "Predictive current control of a voltage source inverter," *IEEE TIE*, vol. 54, no. 1, pp. 495 - 503, 2007.

[6] T. Englert and K. Graichen, "Nonlinear model predictive torque control of PMSMs for high performance applications," *Control Eng. Prac.*, vol. 81, pp. 43 - 54, 2018.

[7] L. Samaranayake and S. Longo, "Degradation control for electric vehicle machines using nonlinear model predictive control," *IEEE TCST*, vol. 26, no. 1, pp. 89 - 101, 2017.