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Neural Network-based Predictive Control (NN-MPC) System for Energy Optimization in Sports Facilities: A Case Study

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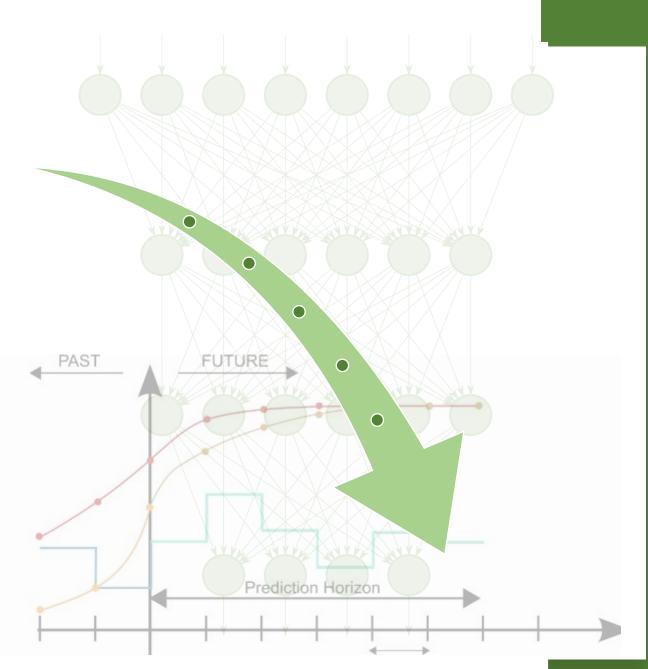
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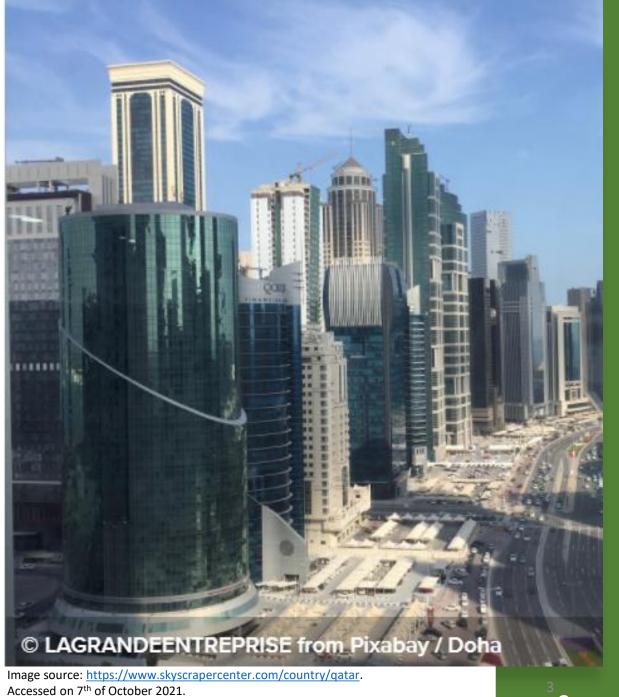
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# Introduction

- The buildings sector accounts for<sup>1</sup>:
  - Over one-third of global final energy consumption
  - Nearly 40% of total direct and indirect CO2 emissions
- Energy demand from buildings continues to rise, driven by:
  - Improved access to energy in developing countries
  - Greater ownership and use of energyconsuming devices
  - Rapid growth in global buildings floor area



# Introduction

- The great dependence on fossil fuels is a key factor
- Mitigating the climate change is a key challenge of the 21st century<sup>2</sup>
- The world urgently needs to use energy efficiently while embracing clean energy sources



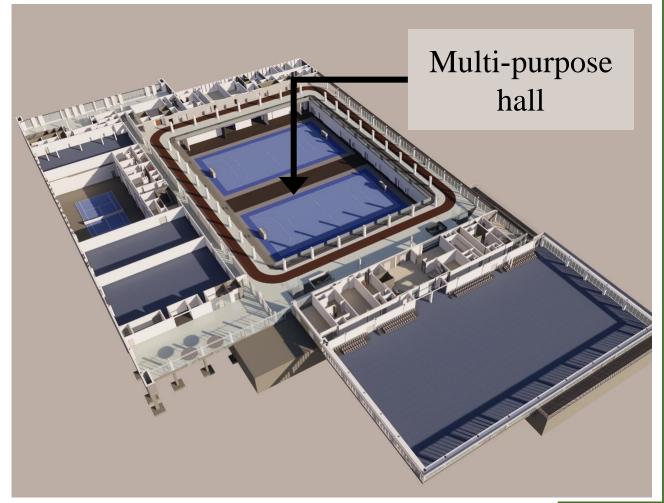
Image source: <u>https://rmi.org/financing-the-transition-from-coal-to-clean-energy/</u> Accessed on 7<sup>th</sup> of October 2021

# Objective

- The theory of model predictive control (MPC) is combined with neural networks (NNs) for temperature setpoint selection to achieve energy and performance optimization of sports facilities
- The proposed approach represents a temperature setpoint optimization system that accounts for the current and future system transitions in the decision-making process

#### Description of the Case Study

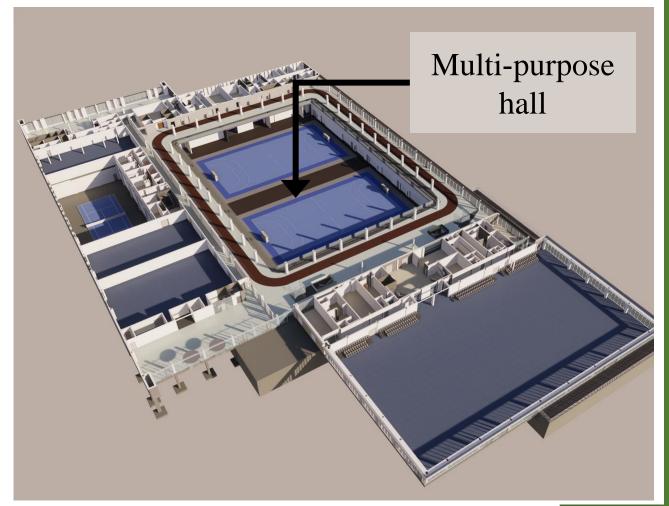
- The sports and events complex of Qatar University operates from 8 am to 3 pm
- The case study in this work is demonstrated on the multipurpose hall, the largest conditioned space in the complex
- The hall extends from the ground floor to the roof with a total floor area of about 7,500 m<sup>2</sup>



The image is a courtesy of the Capital Project Department of Qatar University.

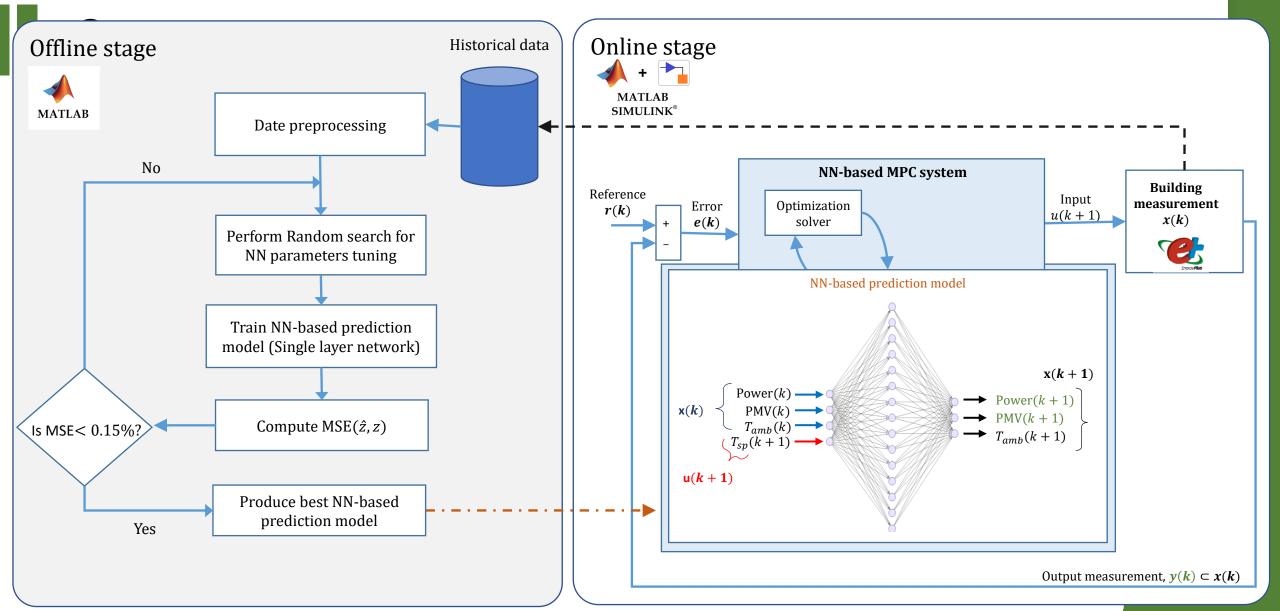
#### Description of the Case Study

- It is controlled at a temperature of 22°C during occupancy period
- The sports mode of the multipurpose hall accommodates about 1200 people
- The building information model of the complex was used to develop the Energy Plus simulation model using Design Builder software



The image is a courtesy of the Capital Project Department of Qatar University.

# The Proposed NN-based MPC Approach



# The Proposed NN-based MPC Approach

• Theory of Model-Predictive Control (MPC):

 The MPC system consists of an **optimizer** and a **prediction model** of the building operation to decide the temperature setpoint given a cost function, J:

$$J = \sum_{j=1}^{n_y} \sum_{i=1}^{n_p} \frac{w_j}{s_j} e_j^2(k+i|k) + \sum_{i=0}^{n_{p-1}} w_{\Delta u} \Delta u^2(k+i|k)$$

$$\mathbf{y}(\mathbf{k}) \begin{cases} Power(k) \\ PMV(k) \\ T_{sp}(k+1) \\ \mathbf{u}(\mathbf{k}+1) \end{cases}$$

States at time k	$\ensuremath{\textbf{Outputs}}$ at time $k$	$\label{eq:response} \textbf{Reference} \text{ at time } k$	Input at time k	<b>Error</b> at time $\mathbf{k}$
x(k)	<i>y</i> ( <i>k</i> )	r(k)	<i>u</i> ( <i>k</i> )	e(k) = y(k) - r(k)

 $\odot$  The main parameters of the MPC are:

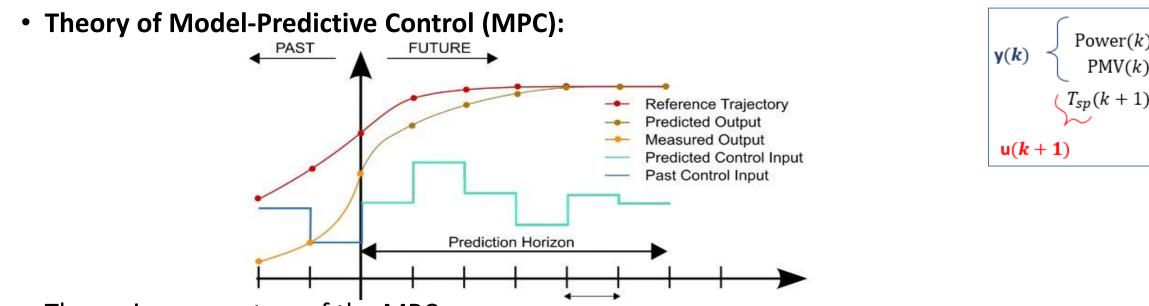
• <u>prediction horizon</u>,  $n_p$  determines the extent the controller investigates the future when optimizing u(k)

• <u>control horizon</u>,  $n_c \in [1, n_p]$ , represents the number of control actions u(k) to be optimized at every step

• <u>output weights,</u> w determine the relative importance of the variables to the optimization objective

• <u>Scale parameters</u>, *s* to normalize the error signals to avoid optimization failure or sub-optimality due to output variables' diverse magnitudes

# The Proposed NN-based MPC Approach

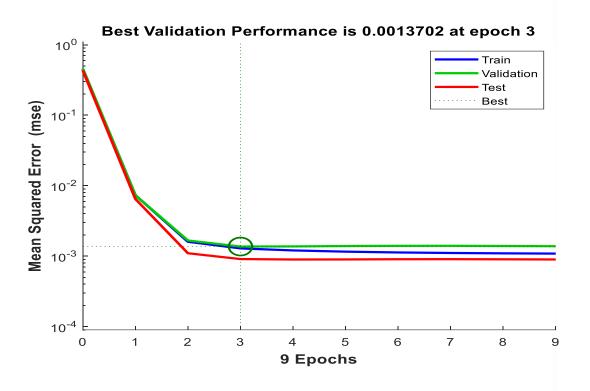


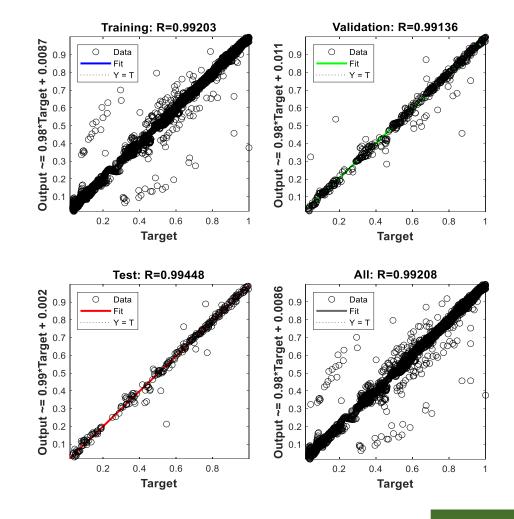
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#### Results and Discussion

• Training of the NN-based prediction model:





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• The performance of the NN-based MPC system:

Energy reduction	Average PMV	n <sub>p</sub>	n <sub>c</sub>	<i>w</i> <sub>1</sub>	<i>w</i> <sub>2</sub>	Computation time per simulation step (sec)
33.45%	1.26	2	1	6.05	12.05	0.19
9.94%	1.20	3	2	6.05	0.05	0.22
19.08%	1.24	4	3	6.05	8.05	0.28

Power(k)

PMV(k

 $T_{sp}(k+1)$ 

 $\mathbf{y}(\mathbf{k})$ 

u(k+1)

 $\circ n_p$  is the prediction horizon  $\circ n_c$  is the control horizon  $\circ w_1$  is for PMV variable  $\circ w_2$  is for Energy variable

Comparison: With MPC Without MPC PMV 1.5 10 11 12 13 9 14 15 3e5 Energy (J) ----- Without MPC With MPC 2e5 1e5 12 13 10 11 14 8 9 15 28 **Femperature** (Degree C) Temperature setpoint ——— Space air temperature 26 24 22 10 11 12 13 14 15 9 Time (hour of the day)

# Conclusion

- Effective utilization of a neural network-based MPC system for setpoint selection to achieve energy and performance optimization of sports facilities using simulation tools
- MPC systems allow integrated dynamic optimization that accounts for the future system behavior in the decision-making process
- A neural network was used for the system prediction element of the MPC system since it is unpractical and difficult to obtain explicit models for complex buildings such as sports facilities
- Neural networks are advantageous for their ability to represent complex interdependencies with high accuracy

# Conclusion

- The proposed approach was able to achieve a total energy savings of about 34%
  Considerations about the prediction model performance, tuning of the MPC settings, and optimization sub-optimality or failure are essential during both design and implementation phases
- The MPC system for setpoint optimization complements the existing management and automation system of the facility, thus can be easily integrated

# Future Work

We plan to work on improvements of the proposed system by:

- Including additional controlling variables such as occupancy rate, air ventilation rate, etc.
- Expanding the objective to include factors related to the safety and health of users
- \* We plan to validate the proposed framework using practical experiments

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# Thank you for your kind attention

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More information about the **SportE.3Q project** can be found in <u>https://www.sporte3q.com/</u>