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MARINE
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Hydraulic Power Take-Off for Wave Energy Converters

Design, Validation and Control

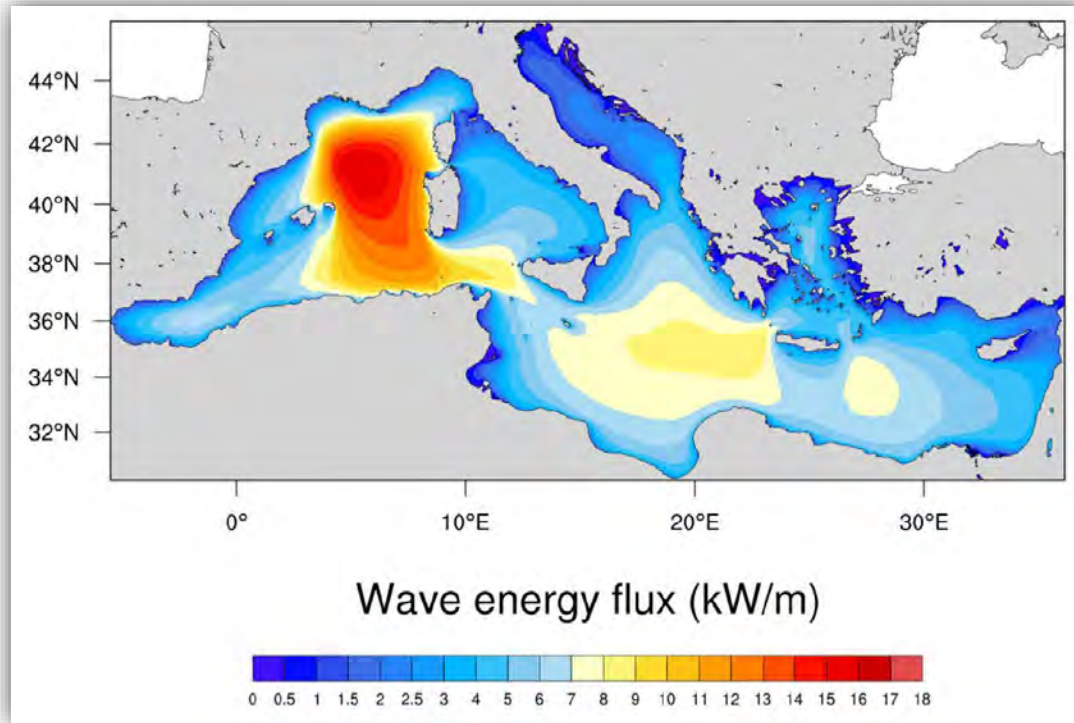
Torino, 23/10/2020

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- ❑ The Hydraulic Power Take-Off
- ❑ Preliminary studies
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Introduction – Wave Energy Context



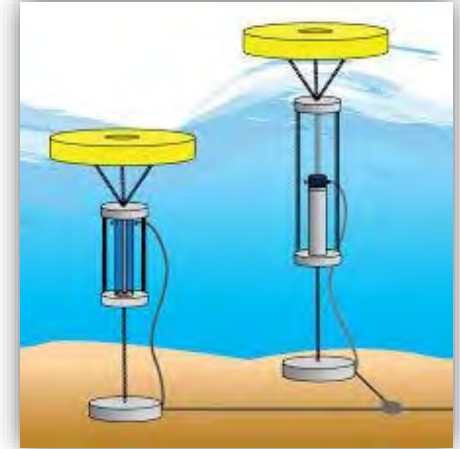
The global ocean wave power potential is approximated to be up to 10 TW, and the annual ocean wave energy is approximated to be up to 93,000 TWh [1]

Harness wave energy damping directly the motion of a floating or submerged buoy

Relative motion of mechanical and structural parts constantly in contact with the harsh marine environment

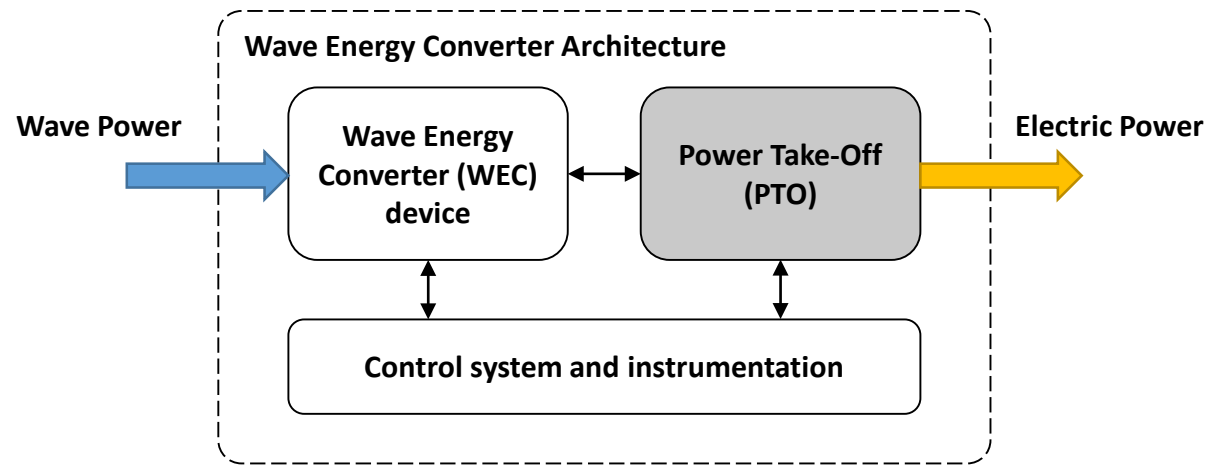
Ocean waves convey power through motions that feature high torques and low speeds

The torques become too high to be handled by electro-mechanical components, up to a level that commercial units are no longer available



[1] Melikoglu, M. Current status and future of ocean energy sources: A global review.

Introduction – Motivation



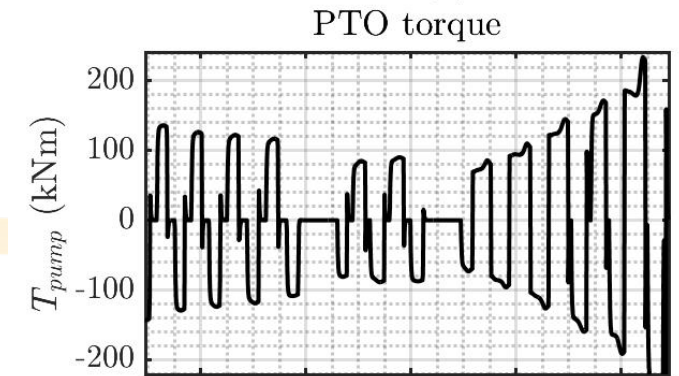
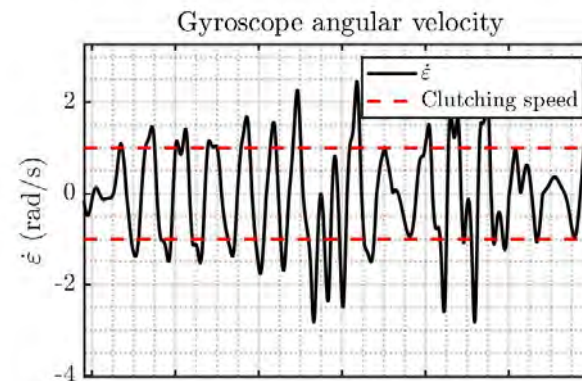
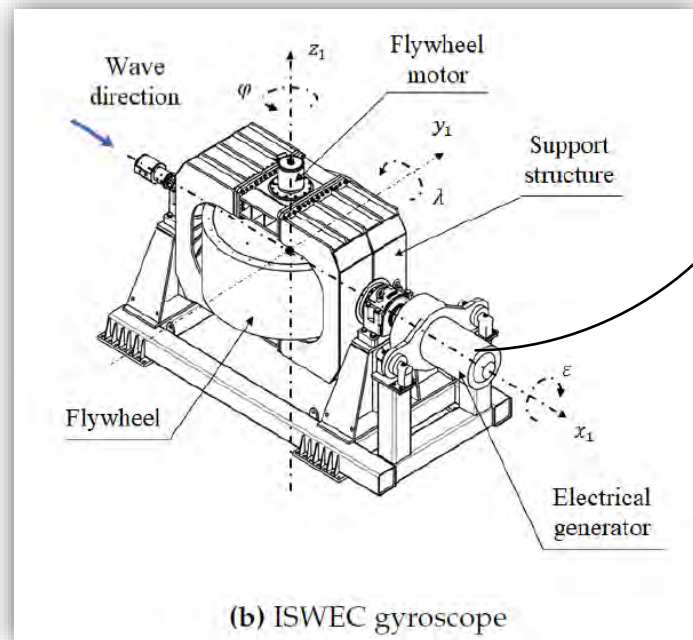
Ocean waves convey power through motions that feature high torques and low speeds

Bidirectional motion and wide range of operational conditions

The torques become too high to be handled by electro-mechanical components, up to a level that commercial units are no longer available

Handle power peaks, irregular power generation and Isolated grid management

Typical PTO Torque and Speed



Solution?

The Hydraulic Power Take-Off - Advantages

“Hydraulic PTO components are the choice of the vast majority of developers since they offer unmatched force density at low velocities, high controllability and relatively easy rectification (valves) and smothering (accumulators) solutions” [1].

Advantages

High torque densities at low velocities with standard components

High Controllability and relatively easy rectification

High scalable employing the same technology

Smothering and storing a large quantity of energy

Elettro-mechanical PTO comparison

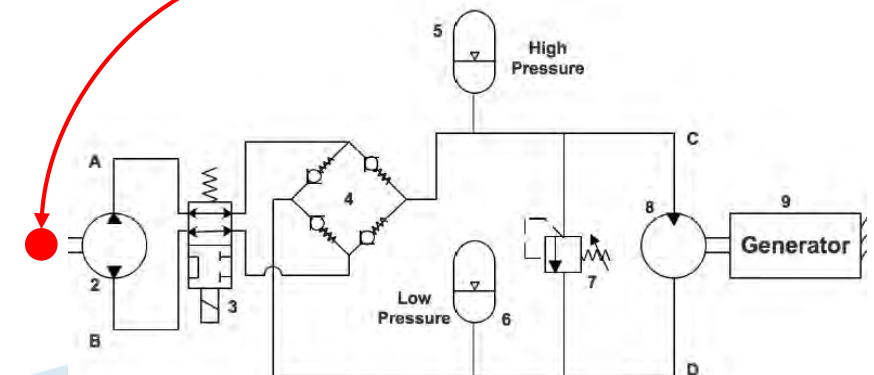
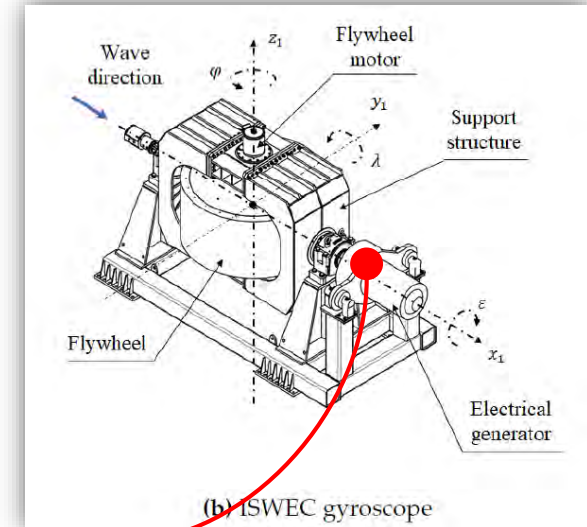
No Gear box and No supercapacitors required

Downsize of the power electronics and electrical generator

To be assessed

Efficiency – expected 70-80%

PTO costs: 25 % of the total device cost



The Hydraulic Power Take-Off – PhD Timeline

1 st Year	2 nd Year	3 rd Year
Hydraulic PTO preliminary studies HIL test rig design	Hydraulic PTO advanced non-linear MPC control HIL test rig construction	Hydraulic PTO design-tool HIL test rig experimental tests and validation



→ Configurations analysis

→ Controllability

→ Performance evaluation

→ MPC control framework

→ Efficiency aware optimal control

→ Performance evaluation

→ Genetic algorithm

→ Tecno-economic optimization

→ Multi-objective performance evaluation

Existing numerical model supported the test rig design

Advanced control techniques call for a performing hardware for the test rig



→ Circuit configuration

→ Hydraulic components

→ Sensors and acquisition system

→ Test rig construction

→ Hardware design

→ Preliminary tests

→ Efficiency evaluation

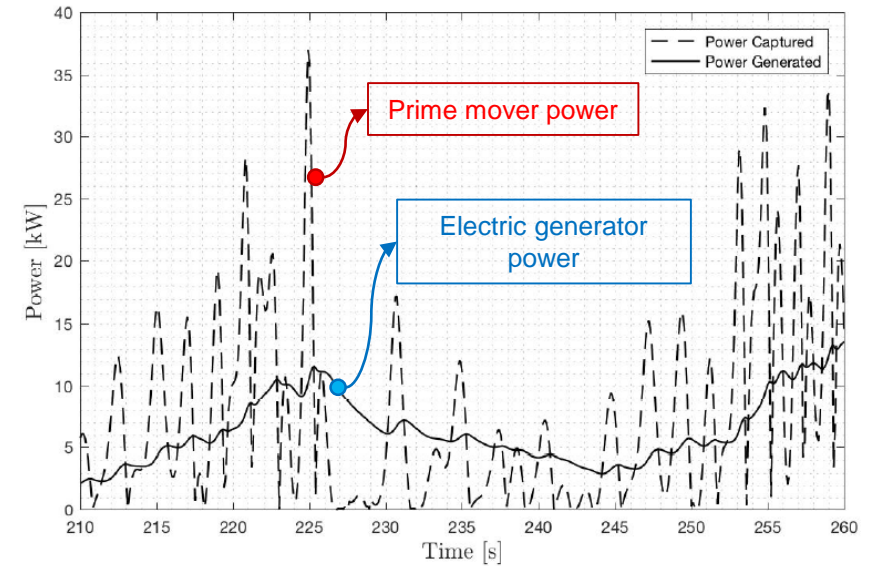
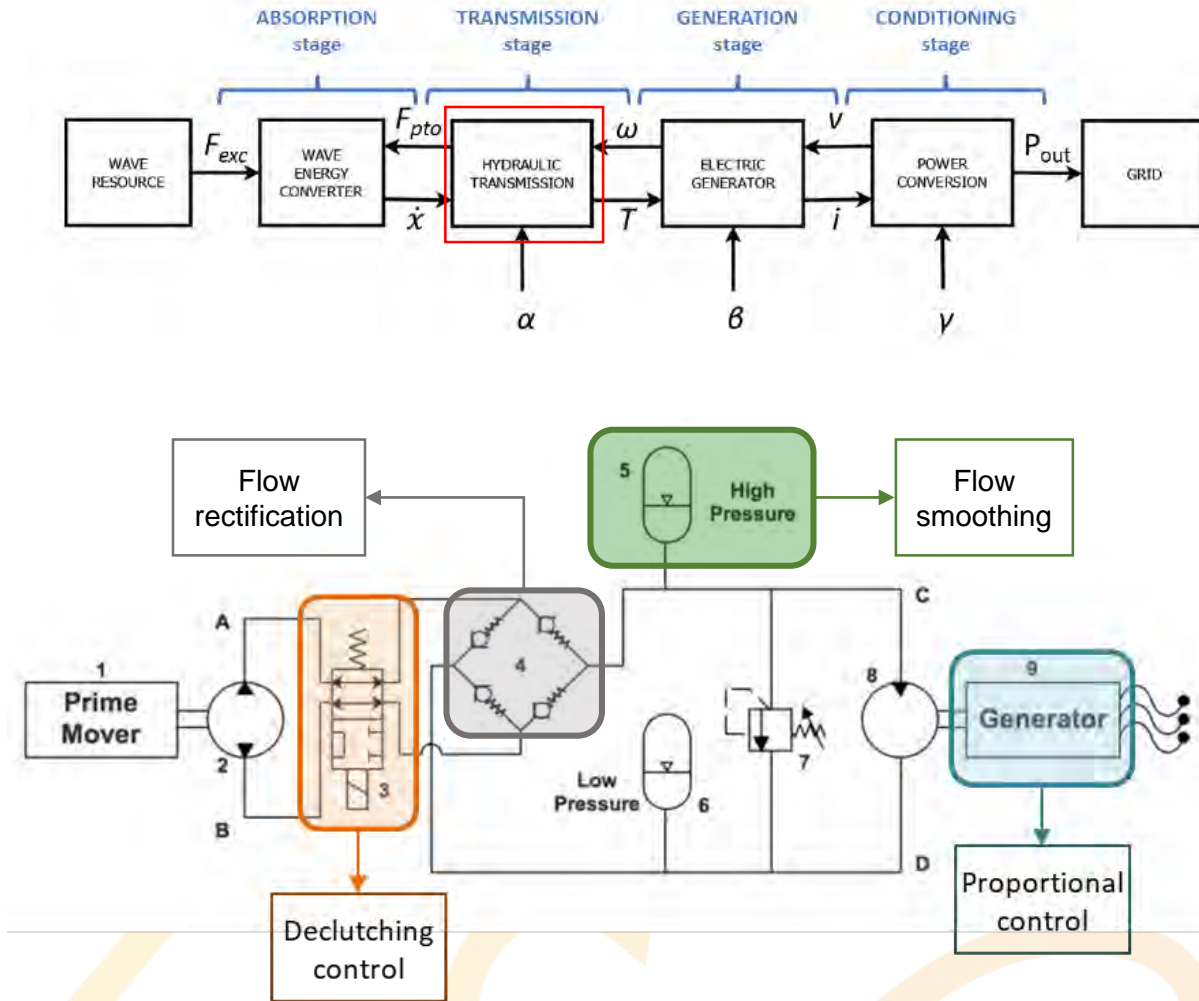
→ Numerical model validation

→ Test rig performances

Hydraulic PTO model validation supported the numerical design-tool

Ongoing activities

Preliminary studies - Architecture

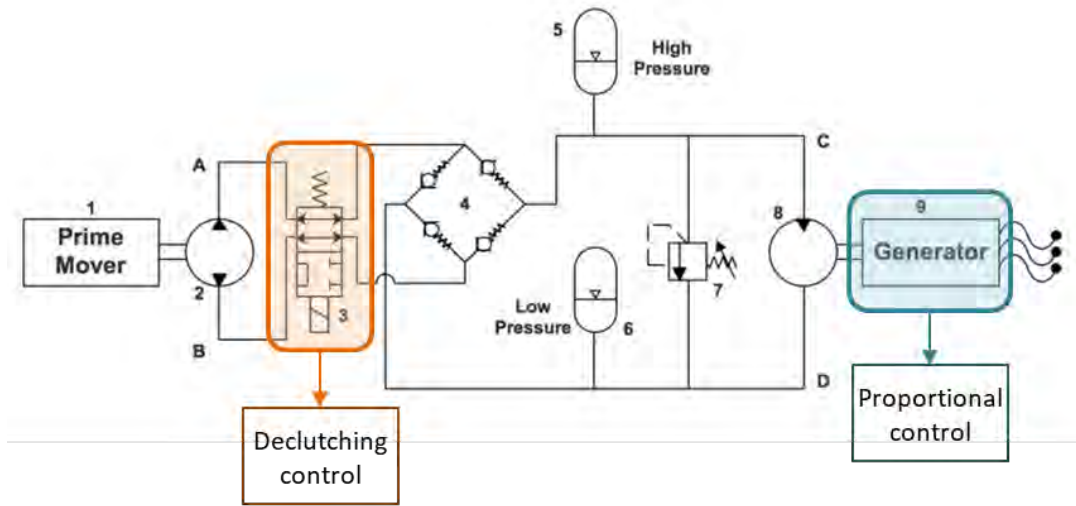


Relatively simple configuration – low components

Possibility to handle more than one gyroscopic unit with one electrical generator

Smothering and control properties to improve the extracted power

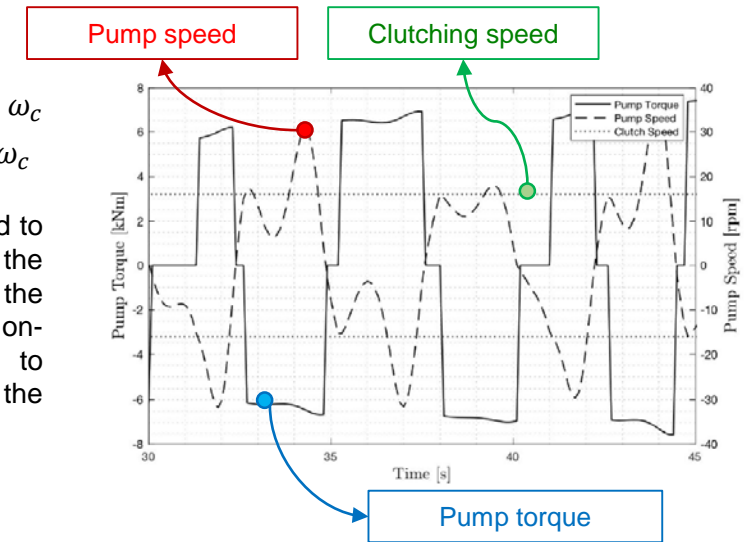
Preliminary studies - Controllability



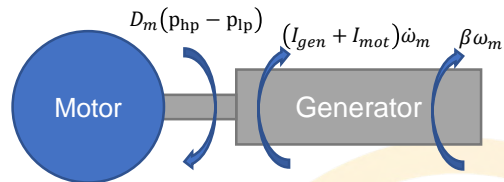
Declutching control

$$\begin{cases} \Delta p_{pump} = \Delta p & \text{if } \omega_p > \omega_c \\ \Delta p_{pump} = 0 & \text{if } \omega_p < \omega_c \end{cases}$$

Declutching control is used to decouple the pump (and the prime mover) from the accumulator action. This on-off modulation allows to 'regulate' the torque on the prime mover.



Proportional control

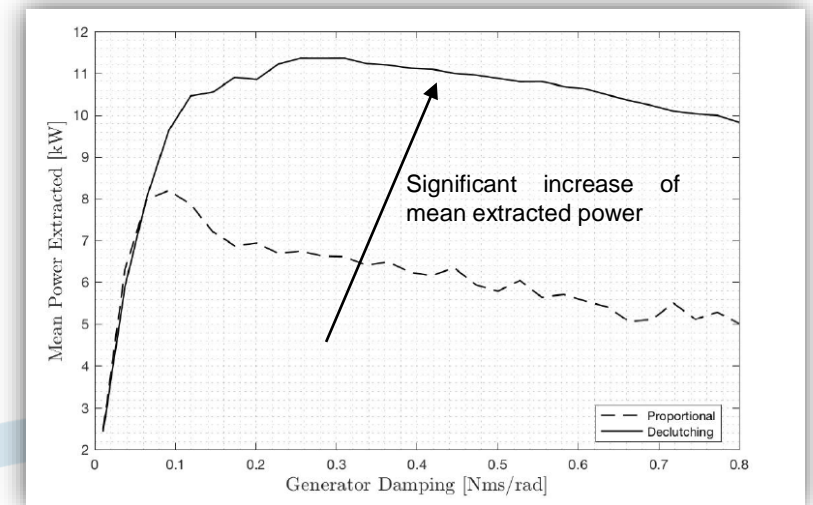
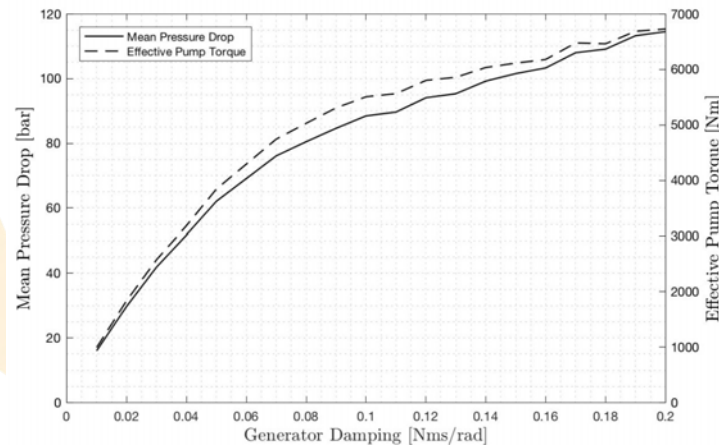


$$D_m(p_{hp} - p_{lp}) = (I_{gen} + I_{mot})\dot{\omega}_m + \beta\omega_m$$

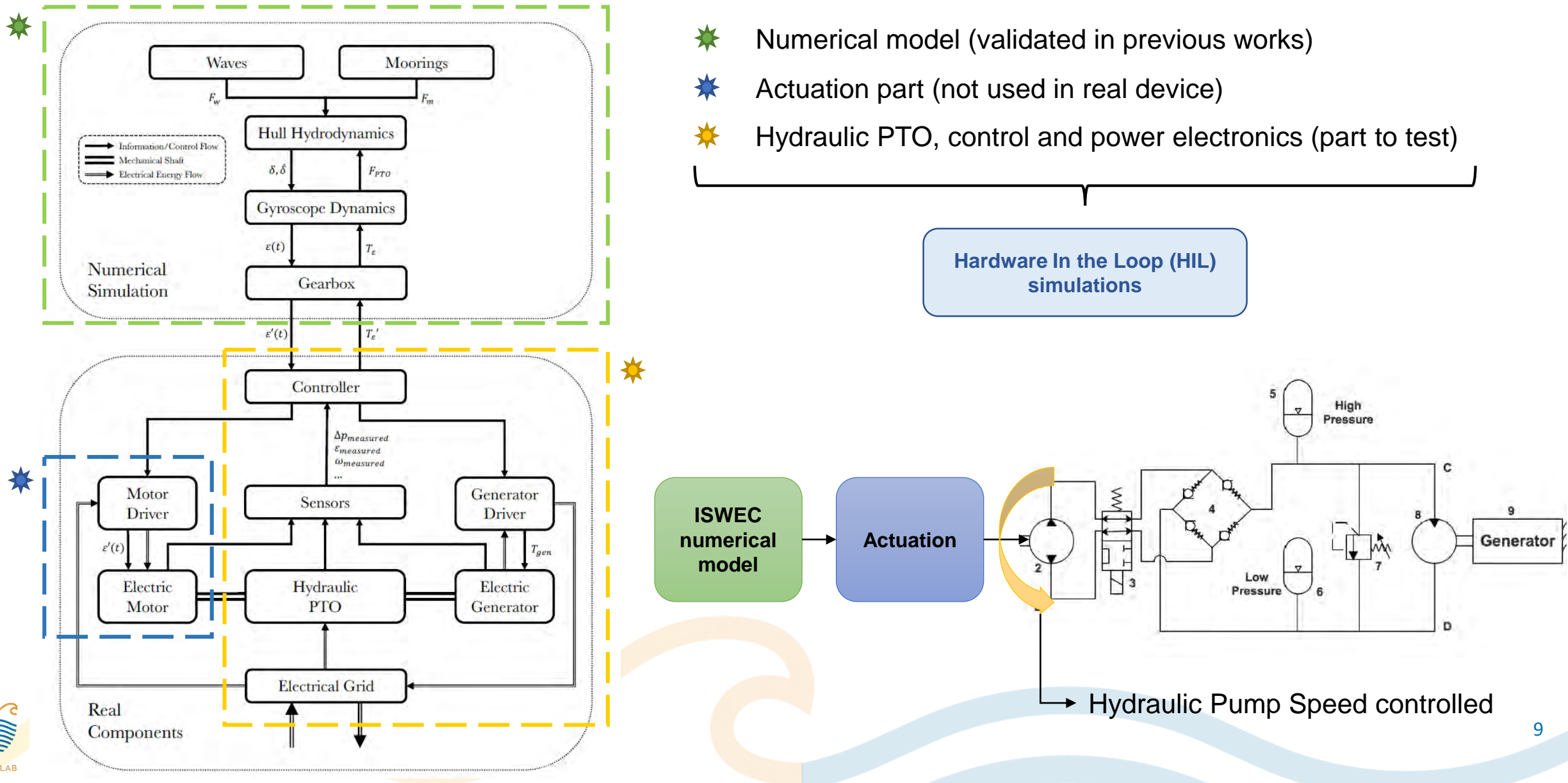
$$D_m(p_{hp} - p_{lp}) \cong \beta\omega_m \rightarrow \frac{D_m^2}{\beta}(p_{hp} - p_{lp}) \cong Q_m$$

$$\text{Equilibrium state: } \bar{\Delta}_p \cong \frac{\beta}{D_m^2} \bar{Q}_m$$

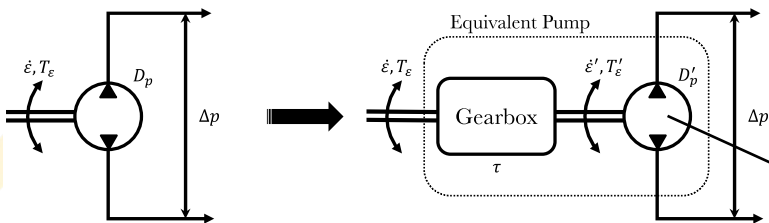
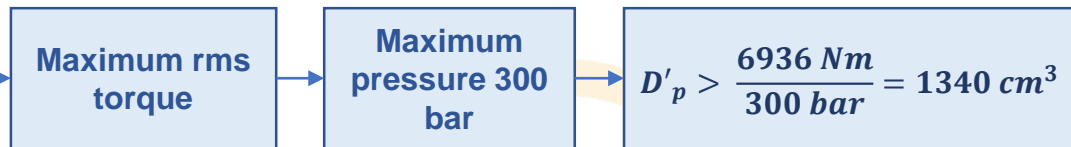
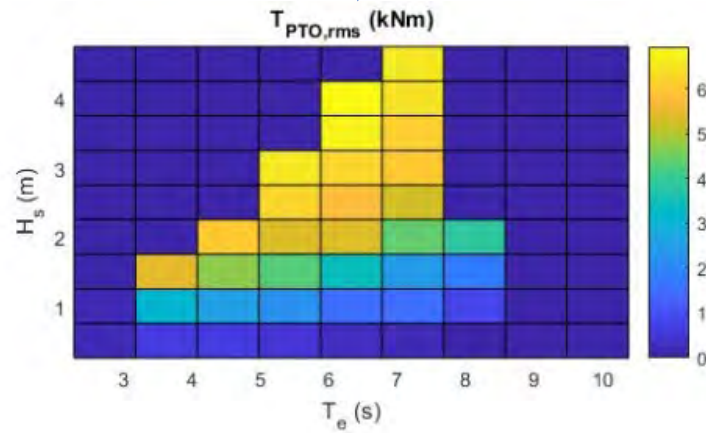
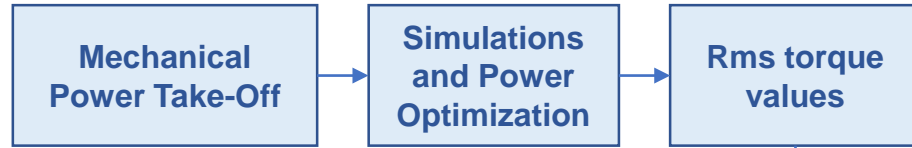
Proportional control ends up being a control on the pressure values, such in a way that their difference is kept around the mean value



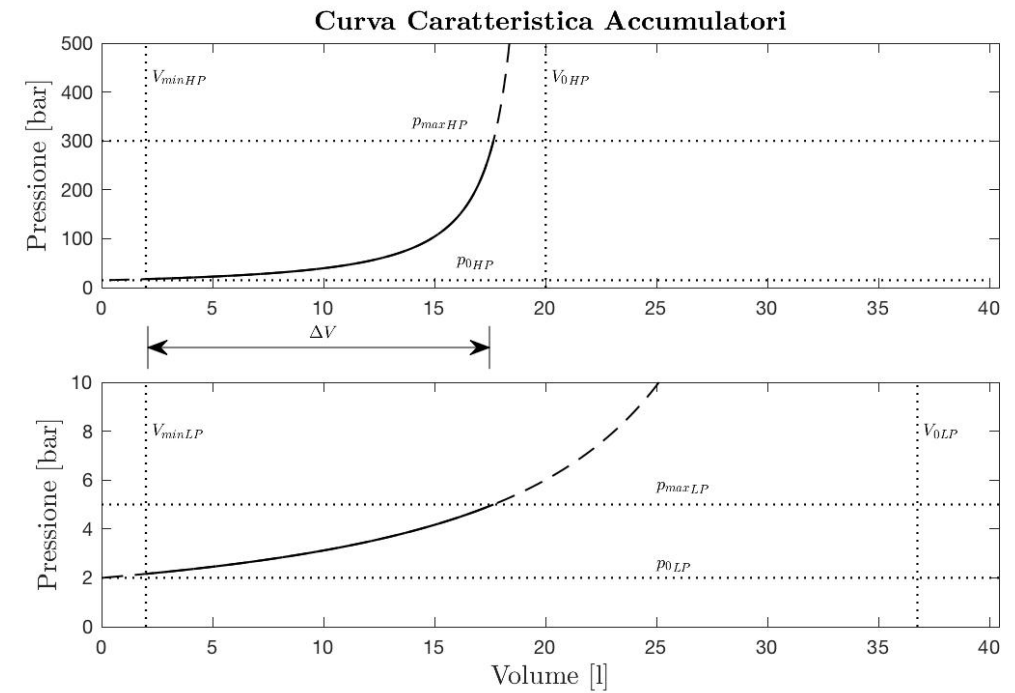
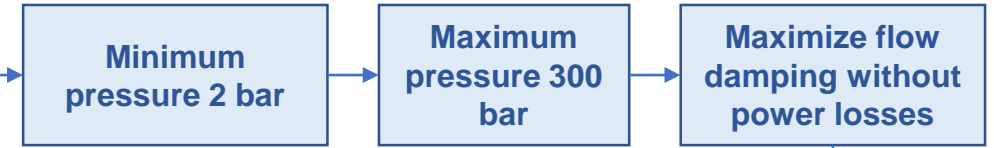
Preliminary studies – HIL test rig



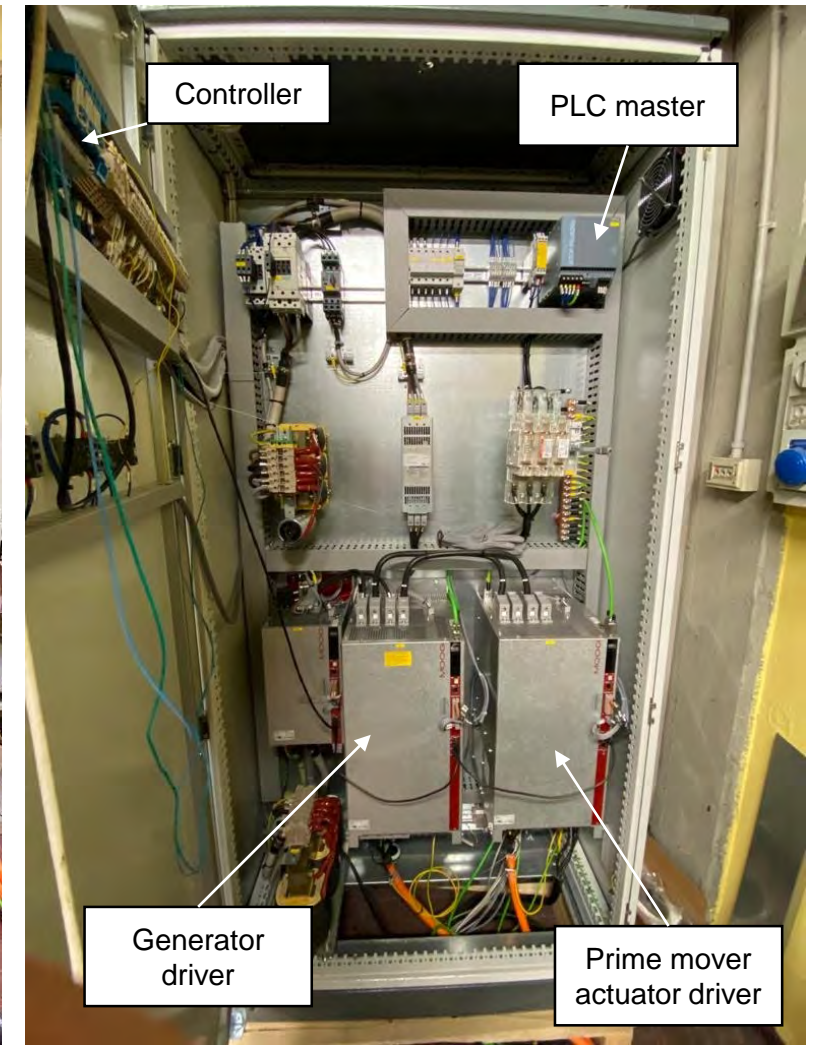
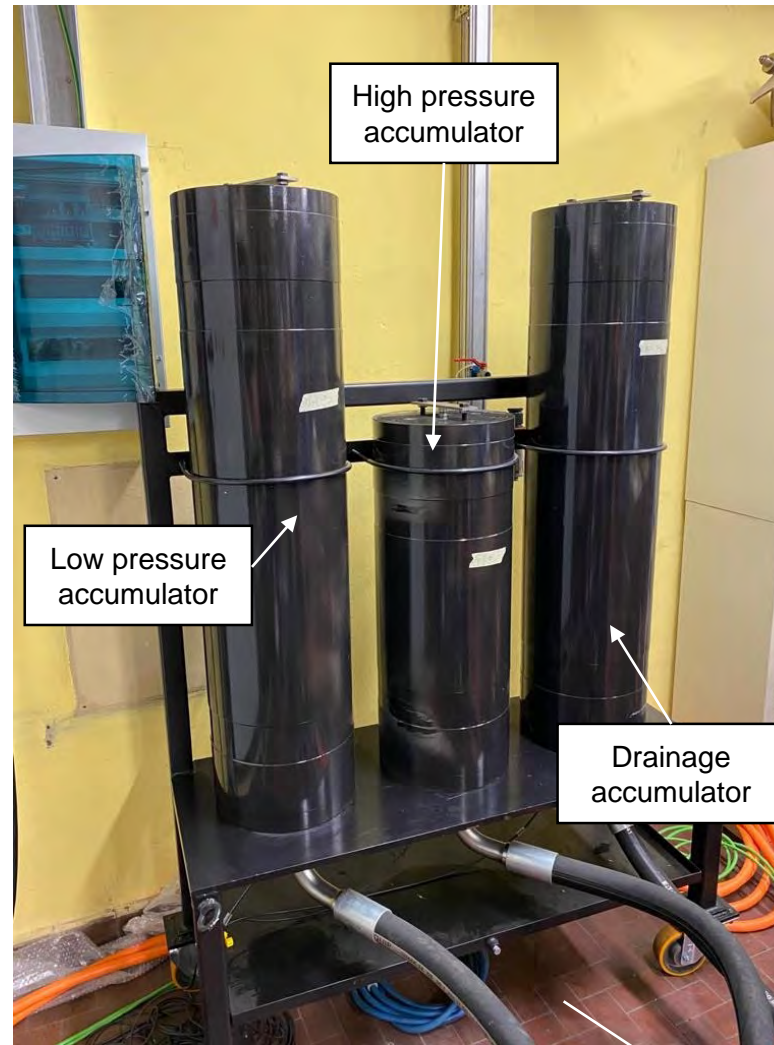
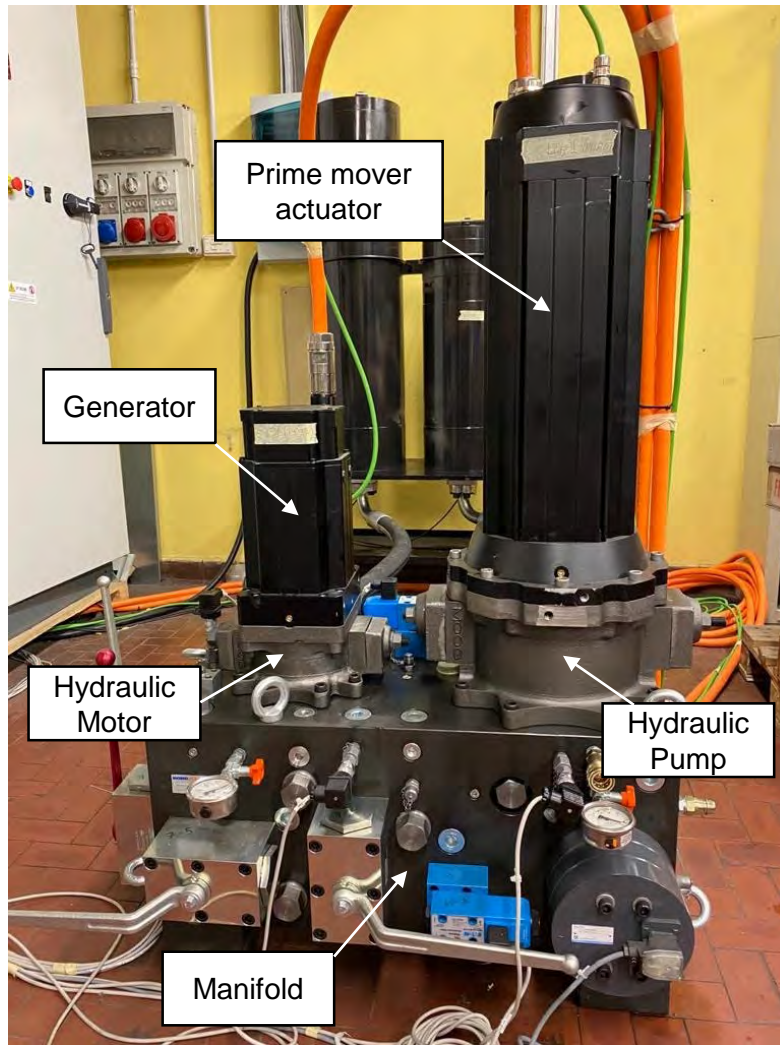
Preliminary studies - Test rig design



Real Pump → 80 cm³

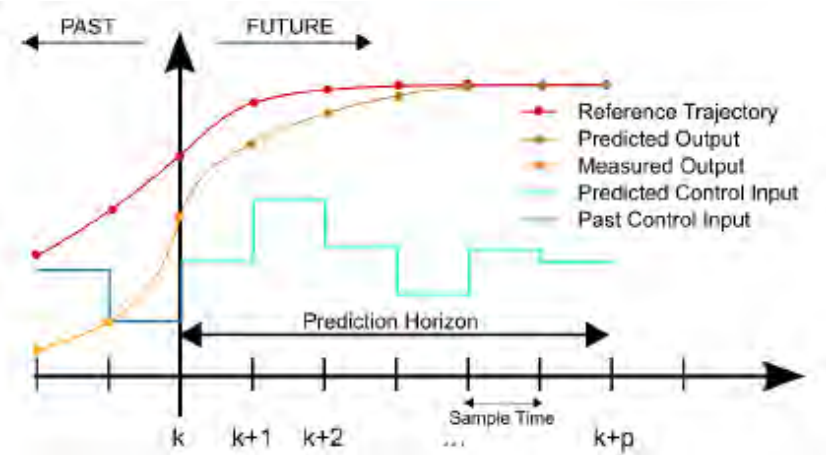


Preliminary studies – Test rig Construction



Optimal control – Control framework

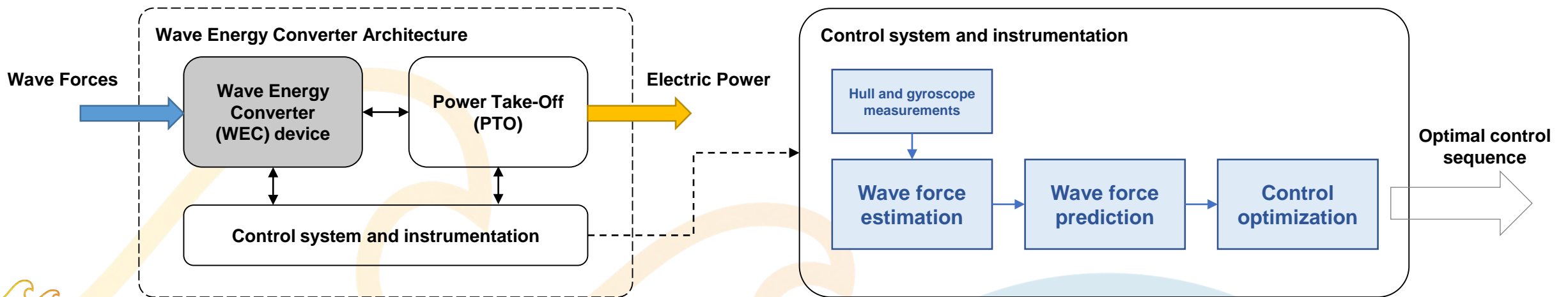
Model Predictive Control (MPC) logic:



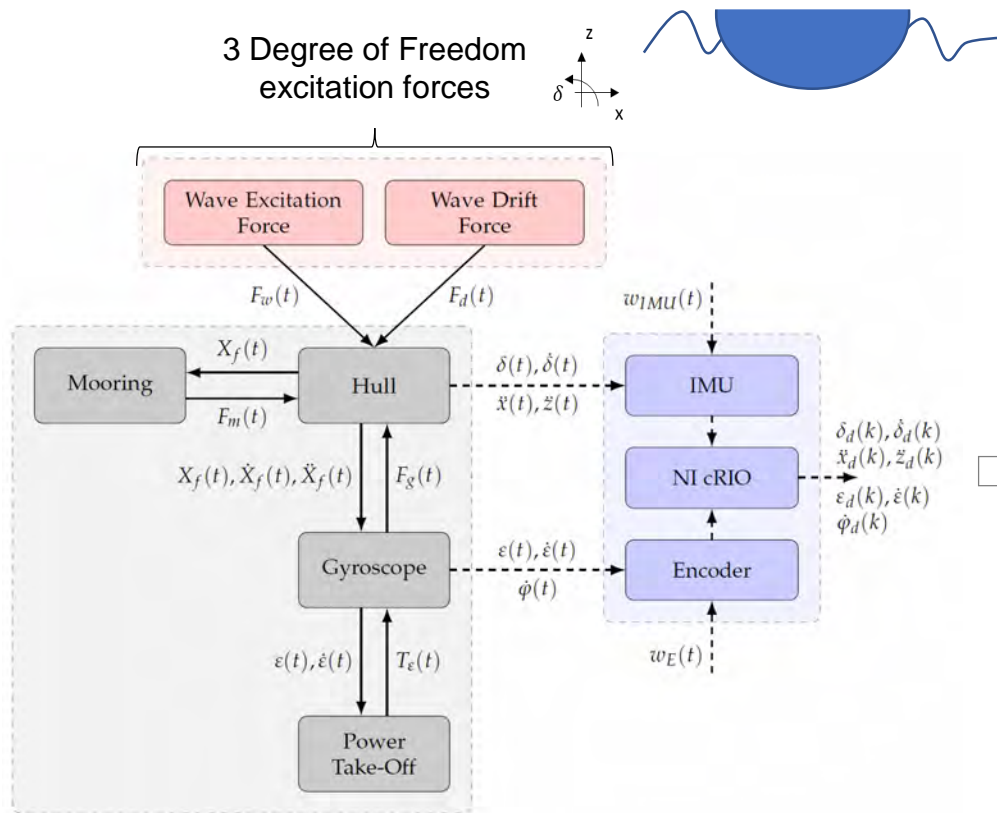
Model based control:

- The underlying model is used to build the cost function to optimize and to forecast the system behaviour
- The control action is computed step-by-step basing on an on-line optimization of a system performance. The system behaviour is predicted and the control action is computed each instant k .
- No gain scheduling required, the control action is computed by the controller

$$\text{Control Objective: } E(t) = - \int_0^t \langle u(t) | y(t) \rangle dt \rightarrow \text{Extracted power}$$



Optimal control – Wave Forces estimation problem



Measurement frameworks available. Four cases are considered: Full Measurement (FM), Motion Reference Unit (MRU), Inertial Measurement Unit with Differential GPS (IMU+DGPS) and Inertial Measurement Unit (IMU)

Measures	FM	MRU	IMU+DGPS	IMU
	Data	Data	Data	Data
$x(m)$	•	○	○	—
$\dot{x}(m/s)$	•	○	—	—
$\ddot{x}(m/s^2)$	•	○	○	○
$z(m)$	•	○	○	—
$\dot{z}(m/s)$	•	○	—	—
$\ddot{z}(m/s^2)$	•	○	○	○
$\delta(rad)$	•	○	○	○
$\dot{\delta}(rad/s)$	•	○	○	○
$\epsilon(rad)$	•	○	○	○
$\dot{\epsilon}(rad/s)$	•	○	○	○
$\dot{\phi}(rad/s)$	•	○	○	○

The estimation problem is addressed basing on the available measurements from **different sensor frameworks**

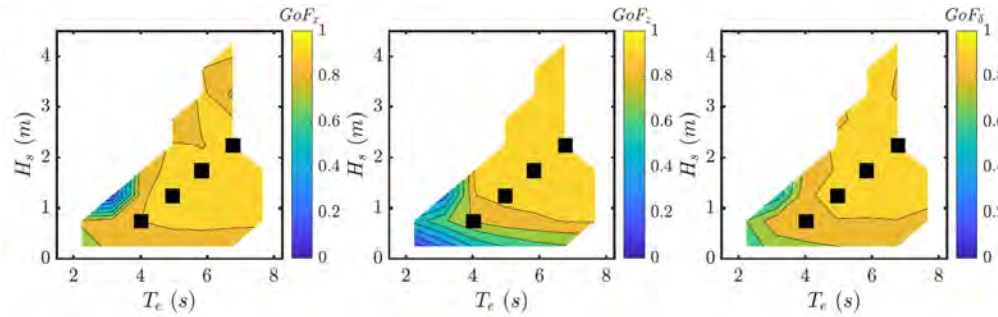
The estimation problem is addressed with two techniques: **Kalman Filter Observer** and **Neural Network Model**

Optimal control – Estimation results comparison

Kalman Filter

$$GoF_j = 1 - \frac{\sqrt{\sum_{k=1}^{T_s} (F_{wj}(k) - \hat{F}_{wj}(k))^2}}{\sqrt{\sum_{k=1}^{T_s} F_{wj}(k)^2}}$$

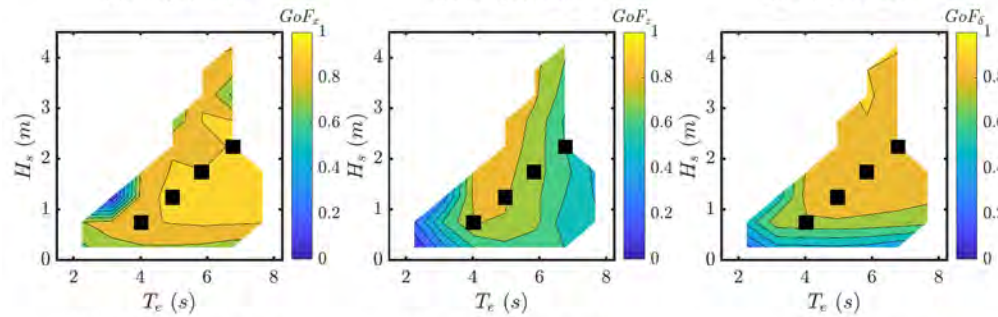
Neural Network



(a) \$GoF_x\$ of MRU

(b) \$GoF_z\$ of MRU

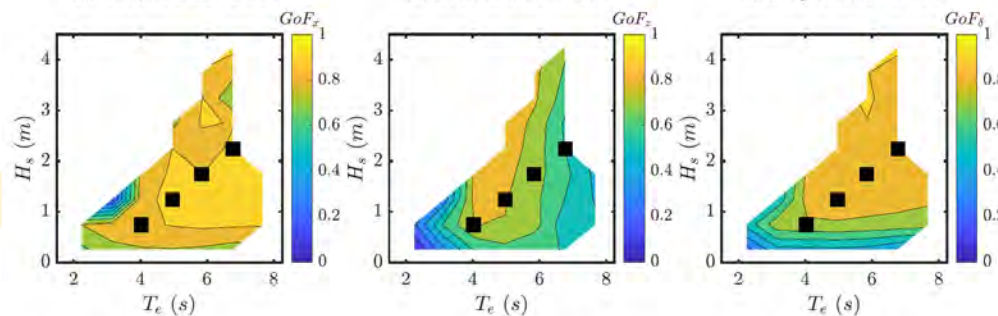
(c) \$GoF_{\delta}\$ of MRU



(d) \$GoF_x\$ of IMU+DGPS

(e) \$GoF_z\$ of IMU+DGPS

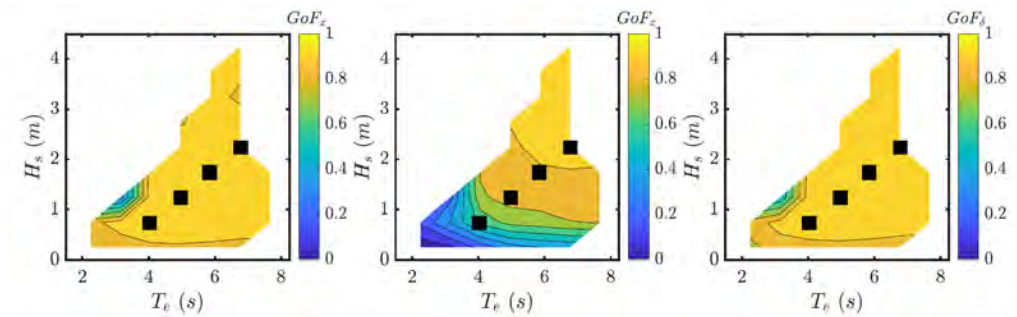
(f) \$GoF_{\delta}\$ of IMU+DGPS



(g) \$GoF_x\$ of IMU

(h) \$GoF_z\$ of IMU

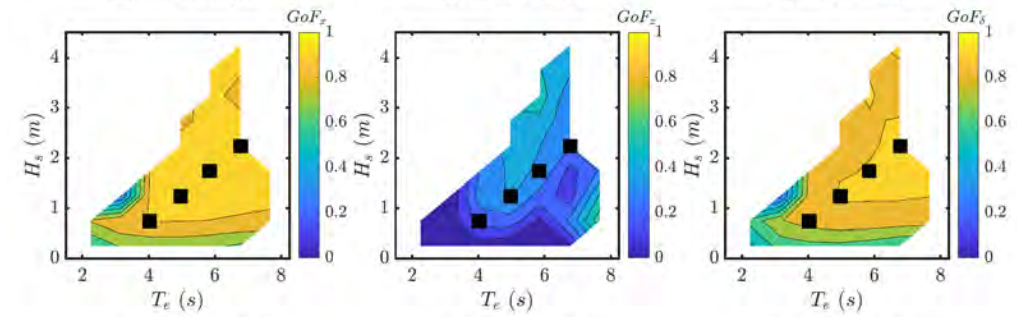
(i) \$GoF_{\delta}\$ of IMU



(a) \$GoF_x\$ of MRU

(b) \$GoF_z\$ of MRU

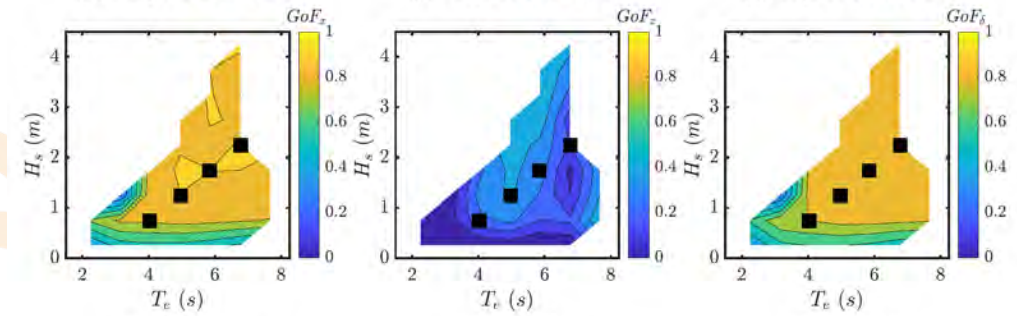
(c) \$GoF_{\delta}\$ of MRU



(d) \$GoF_x\$ of IMU+DGPS

(e) \$GoF_z\$ of IMU+DGPS

(f) \$GoF_{\delta}\$ of IMU+DGPS



(g) \$GoF_x\$ of IMU

(h) \$GoF_z\$ of IMU

(i) \$GoF_{\delta}\$ of IMU

Optimal control – Estimation results comparison

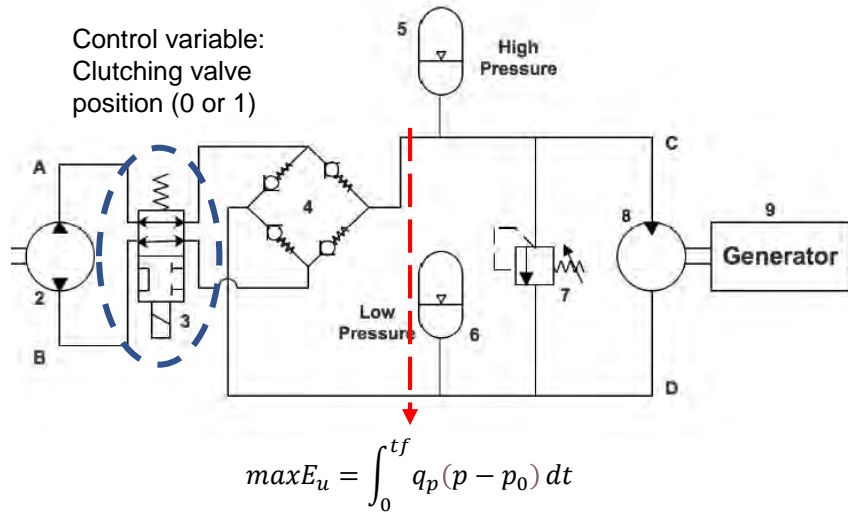
$$\overline{GoF}_j = \frac{\sum_{v=1}^V GoF_j(v) E(v)}{\sum_{v=1}^V E(v)}$$

$$\Delta \overline{GoF}_j = \frac{\overline{GoF}_j|_{NN} - \overline{GoF}_j|_{KF}}{\overline{GoF}_j|_{KF}} 100$$

Framework	Noise	Kalman Filter			Neural Network			$\Delta \overline{GoF}_x(\%)$	$\Delta \overline{GoF}_z(\%)$	$\Delta \overline{GoF}_\delta(\%)$
		\overline{GoF}_x	\overline{GoF}_z	\overline{GoF}_δ	\overline{GoF}_x	\overline{GoF}_z	\overline{GoF}_δ			
MRU*	×	0.910	0.971	0.901	0.951	0.941	0.961	4.50	-3.08	6.65
IMU+DGPS*	×	0.910	0.967	0.902	0.950	0.939	0.959	4.39	-2.89	6.31
IMU*	×	0.893	0.758	0.889	0.913	0.542	0.940	2.23	-28.5	5.73
MRU	✓	0.898	0.873	0.895	0.950	0.771	0.955	5.79	-11.6	6.70
IMU+DGPS	✓	0.875	0.744	0.824	0.912	0.327	0.863	4.22	-56.1	4.73
IMU	✓	0.878	0.744	0.819	0.854	0.322	0.839	-2.73	-56.7	1.82

The Neural Network framework has been chosen in the MPC application.

Optimal control – Optimization problem



$$X = [X_f \dot{X}_f \varepsilon \dot{\varepsilon} \omega V p \zeta_r]^t$$

$$\dot{X} = \begin{bmatrix} \dot{X}_f \\ M_f^{-1}(\tau_w + F_g - K_f X_f - C_r \zeta_r - D_r \dot{X}_f) \\ \dot{\varepsilon} \\ \frac{1}{I_g} \left(-J\dot{\phi}\delta \cos(\varepsilon) - \frac{D_p(\text{sign}\dot{\varepsilon})}{\eta_{mp}} \Delta P_p - mgd \sin(\varepsilon) \right) \\ \frac{1}{(I_{gen} + I_m + I_{added})} D_m \left(\Delta P_m \eta_{mm} \eta_{mgen} - \beta \omega \right) \\ -D_p \dot{\varepsilon} \eta_{vp} u + \frac{D_m \omega}{\eta_{vm}} + q_{rv} \\ -\frac{\gamma p}{V} \left(-D_p \dot{\varepsilon} \eta_{vp} u + \frac{D_m \omega}{\eta_{vm}} + q_{rv} \right) \\ A_r \zeta_r + B_r \dot{X}_f \end{bmatrix}$$

Hull

Gyroscope - Pump

Motor - Generator

Accumulator

Radiation states

Pontryagin's minimum principle

$$H = q_p(p - p_0) + \Lambda^t \dot{X} \rightarrow \text{Co-states trajectories}$$

Cost function

$$H(X, u, \Lambda, t) = \Phi u + \Theta$$

$$X(0) = X_0, \lambda(t_f) = 0$$

$$\frac{\partial \lambda}{\partial X} = \dot{\lambda}, \frac{\partial H}{\partial u} = \Phi, u = \begin{cases} u_{max} & \text{if } \Phi < 0 \\ u_{sing} & \text{if } \Phi = 0 \\ u_{min} & \text{if } \Phi > 0 \end{cases}$$

$$H = q_p(p - p_0) +$$

$$+ \bar{\lambda}_1^T \dot{X}_f + \bar{\lambda}_2^T M_f^{-1}(\tau_w + F_g - K_f X_f - C_r \zeta_r - D_r \dot{X}_f) + \lambda_3 \dot{\varepsilon} +$$

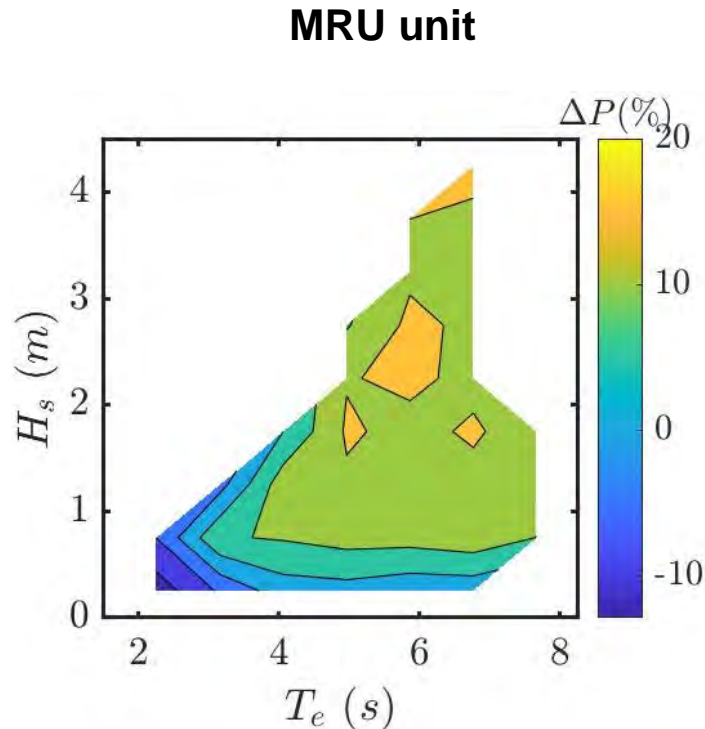
$$+ \lambda_4 \frac{1}{I_g} \left(-J\dot{\phi}\delta \cos(\varepsilon) - \frac{D_p(\text{sign}\dot{\varepsilon})}{\eta_{mp}} \Delta P_p - mgd \sin(\varepsilon) \right)$$

$$+ \lambda_5 \frac{1}{(I_{gen} + I_m + I_{added})} D_m \left((p - p_0) i_{svv} \eta_{mm} \eta_{mgen} - \beta \omega \right) + \lambda_6 \left(-D_p \dot{\varepsilon} \eta_{vp} u + \frac{D_m \omega}{\eta_{vm}} + q_{rv} \right) +$$

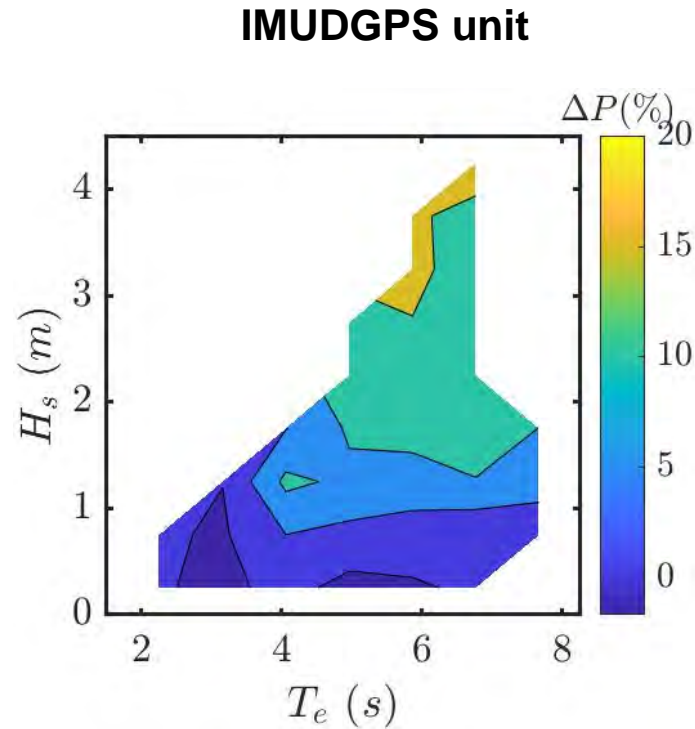
$$+ \lambda_7 \left(-\frac{\gamma p}{V} \left(-D_p \dot{\varepsilon} \eta_{vp} u + \frac{D_m \omega}{\eta_{vm}} + q_{rv} \right) \right) + \bar{\lambda}_8^T (A_r \zeta_r + B_r \dot{X}_f)$$

Optimal control – Results

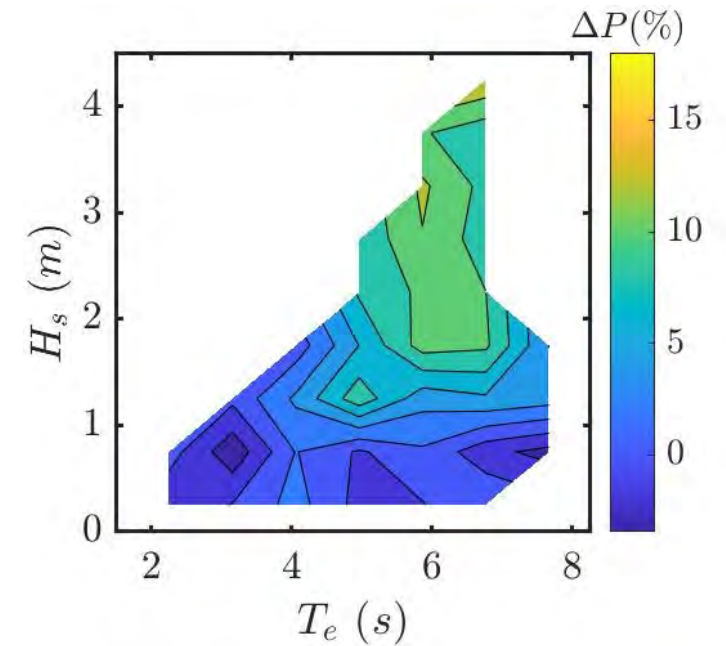
Mean energy increase / decrease:



$\Delta P_w = 12\%$



$\Delta P_w = 9\%$



$\Delta P_w = 6\%$

The MPC application results in a increase of the mean net annual power extracted equal to 12% for the MRU framework, 9% with the IMU+DGPS framework and 6% with the IMU framework. The Neural Network has been considered for the estimation process.

Design tool – Aim

1. Scope

Perform the second optimization considering all the hydraulic PTO free parameters and a small range of gyroscope and hull free parameters. The aim is to obtain the optimal device with the hydraulic PTO, the overall cost including all the hydraulic components and a more accurate performance evaluation.

2. Numerical model

Non linear time domain model with the hydraulic PTO. The performance evaluation is done in the time domain considering only one long realization for each input wave to avoid long computational time. The power losses considered are the flywheel losses, base loads, volumetric losses, pressure losses, valve losses, and generator losses.

3. Expected results

1. Device dimensions, gyroscope and hydraulic PTO components (pump, control logics, accumulator, valves, motor and generator)
2. A serie of ISWEC devices collected by power size
3. Optimization on Cost of Energy (CoE), Relative Capture Width (RCW) and Annual Productivity
4. Decision making for Phase 2

Device free parameters

Hull shape and dimensions (small range) – Gyroscope dimensions (small range)
Hydraulic components (Pump, valves, motor and accumulator)
Electrical generator – Control logics



Productivity analysis with non linear model

Net annual productivity considering the net electric power at the generator minus all the system losses (mooring system excluded)



Annual Productivity, CoE and RCW optimization

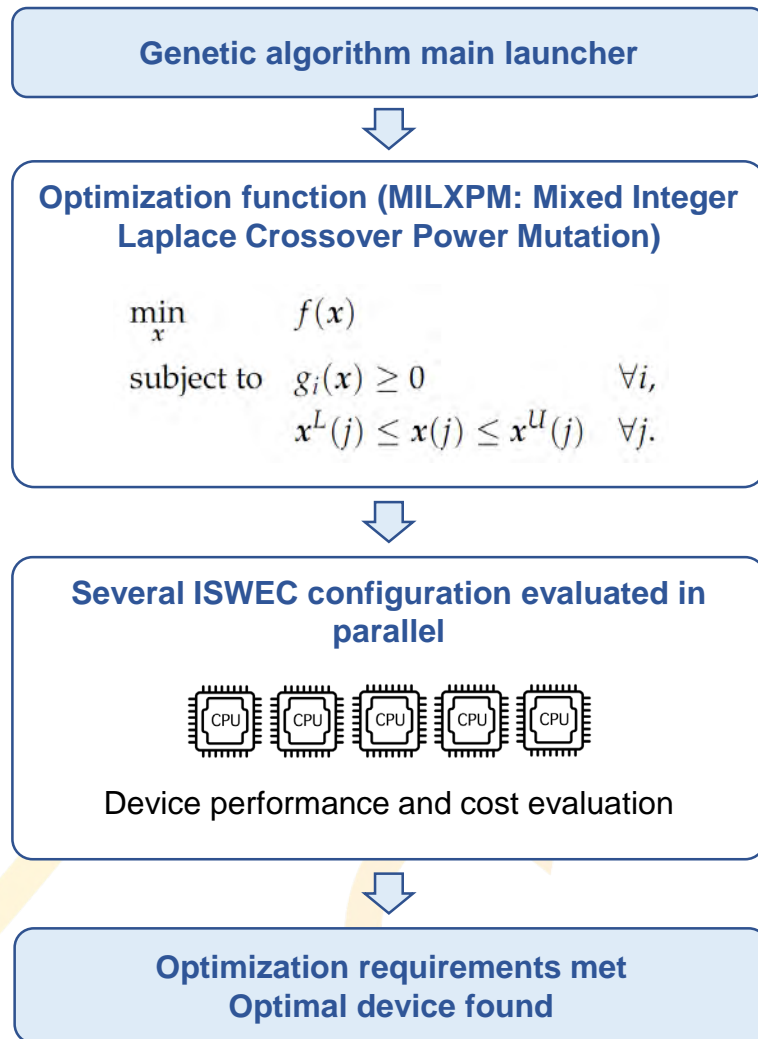
Pool of M optimal devices



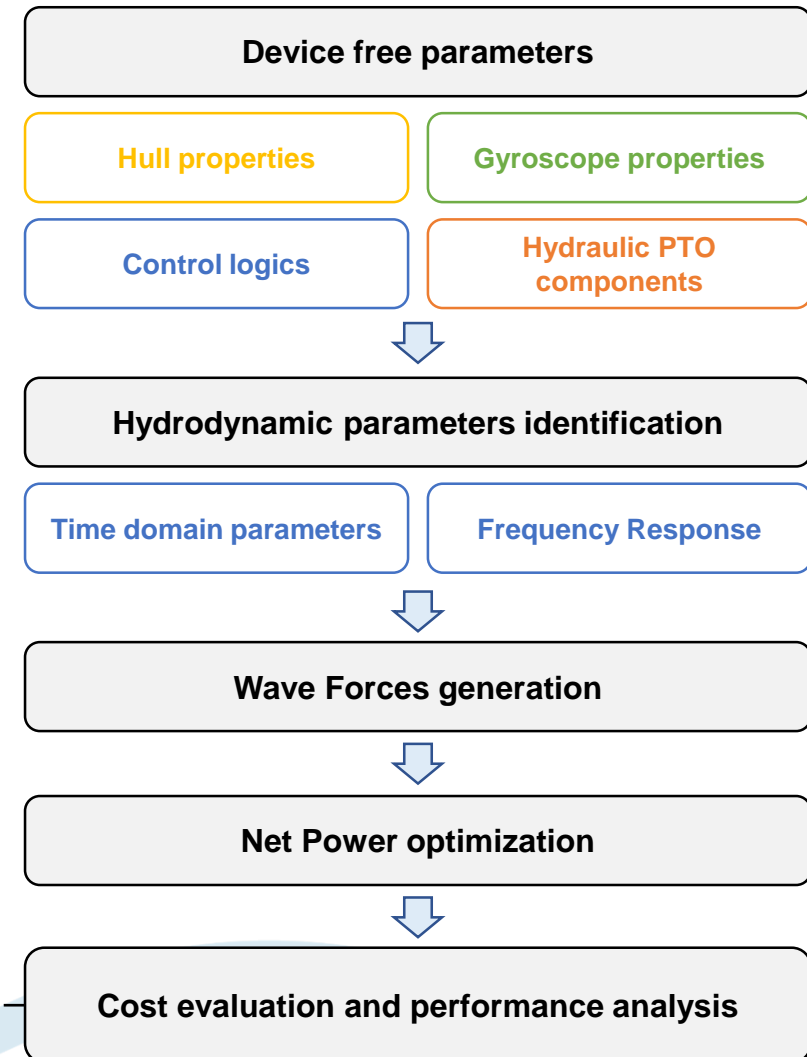
Decision making

Design tool – Architecture

Genetic Algorithm architecture



Generation and Optimization of a single individual



Design tool – Optimization parameters

1 - Pump id: The pump id represents an hydraulic pump model in the catalogue. 20/30 different solution and tandem configuration are collected together with their properties and costs.

2 – Pump control id: The pump control id identifies which control type is considered on the on the pump units. Four different logics are considered:

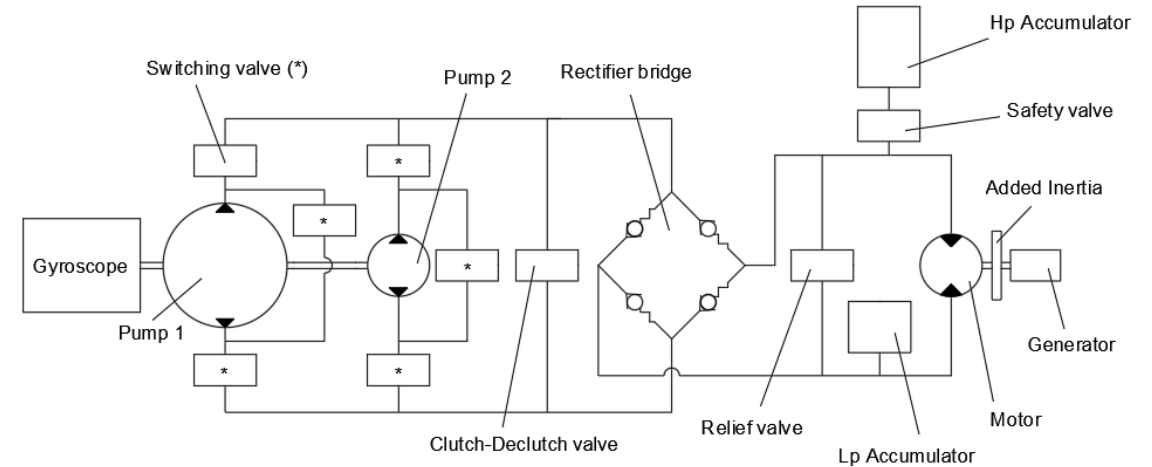
1. NO clutch-declutch valve, NO switching of the pumps
2. YES clutch-declutch valve, NO switching of the pumps
3. NO clutch-declutch valve, YES switching of the pumps
4. YES clutch-declutch valve, YES switching of the pumps

When a single pump is considered, logics 3 and 4 are not available

3 – Accumulator volume: The accumulator volume represents the capacity of the system to filter the flow peaks and store energy to continuous feed the electrical generator.

4 – Accumulator pre-charge pressure: The accumulator pre-charge represents the capacity of the system to filter the flow peaks and fix the lowest level of torque available on the motor and the pump. Remember that the torque acting on the pump – motor is computed from the pressure difference on them.

5 – Motor – Generator id: The motor-generator id represents an hydraulic motor coupled with an electrical generator in the catalogue. 10/12 different solution are collected together with their properties and costs.

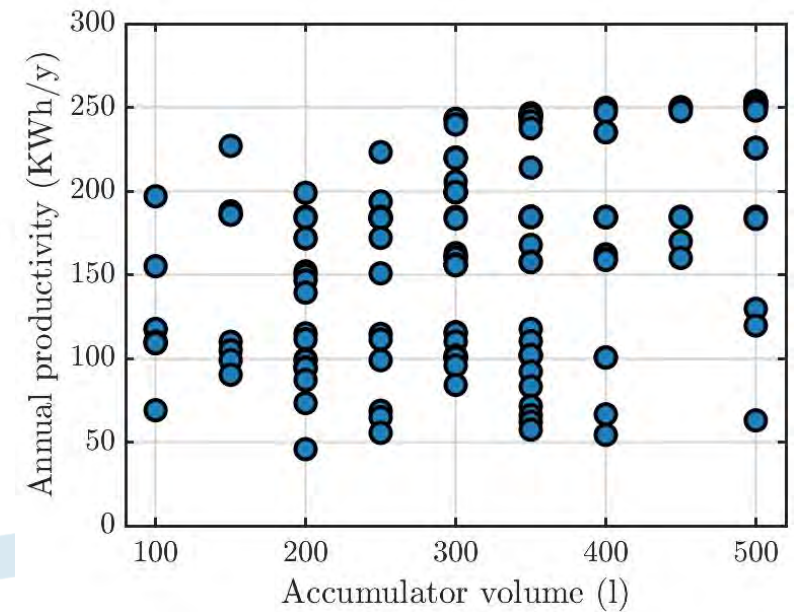
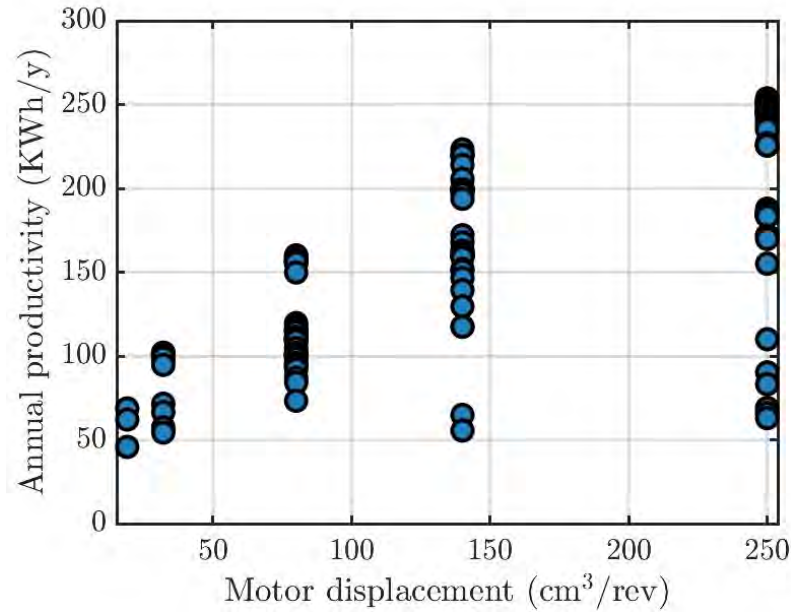
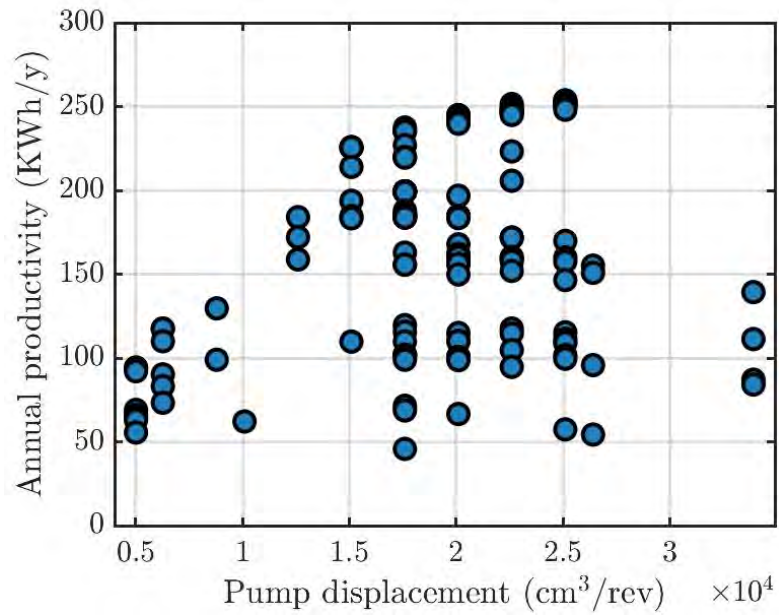


6 – Generator control id: The generator control id identifies which control type is considered on the on the generator. Two different logics are considered:

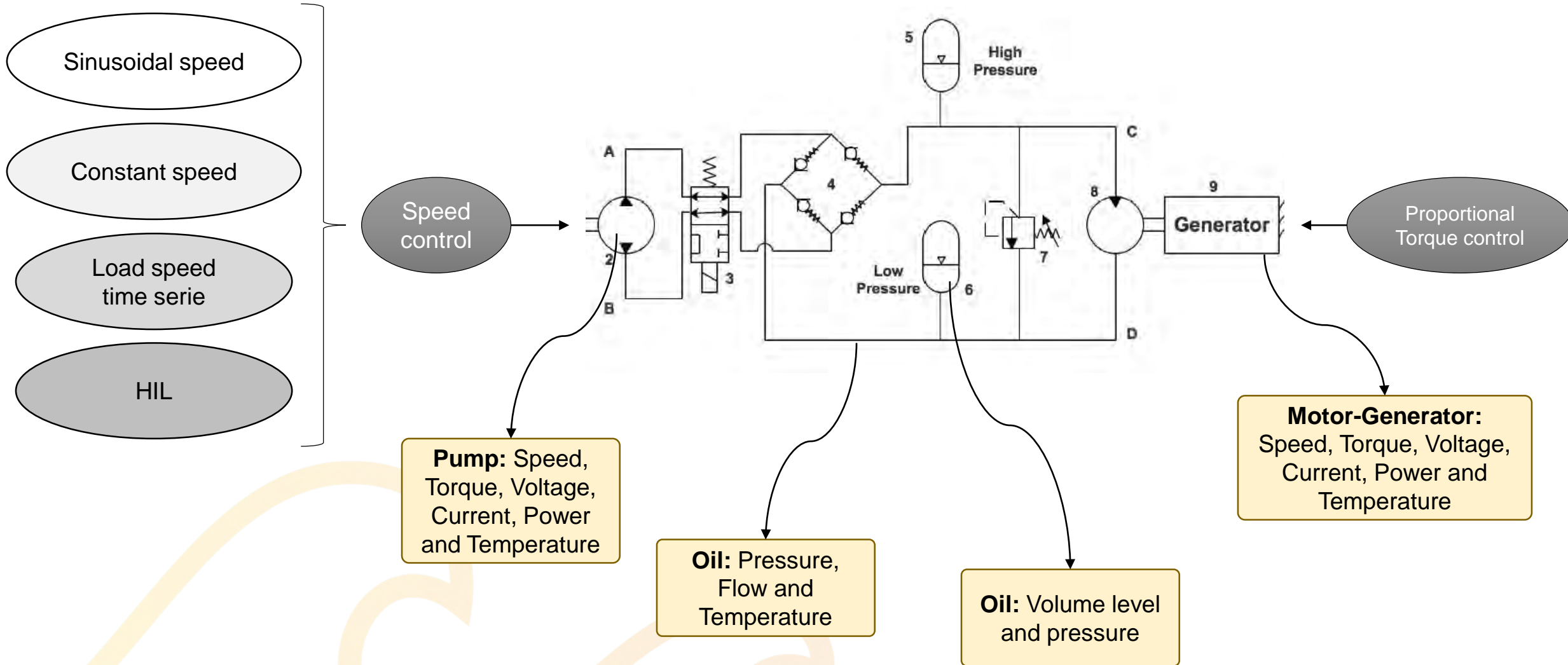
1. Torque control, with a proportional law in respect to the generator angular speed
2. Speed control, in respect to a fixed speed reference (PI)

7 – Added inertia: The low inertia of the motor-generator units is increased to avoid rapid accelerations when there are pressure peaks in the accumulator. Moreover, when the generator is speed controlled, the inertia helps to keep the generator speed around its reference value.

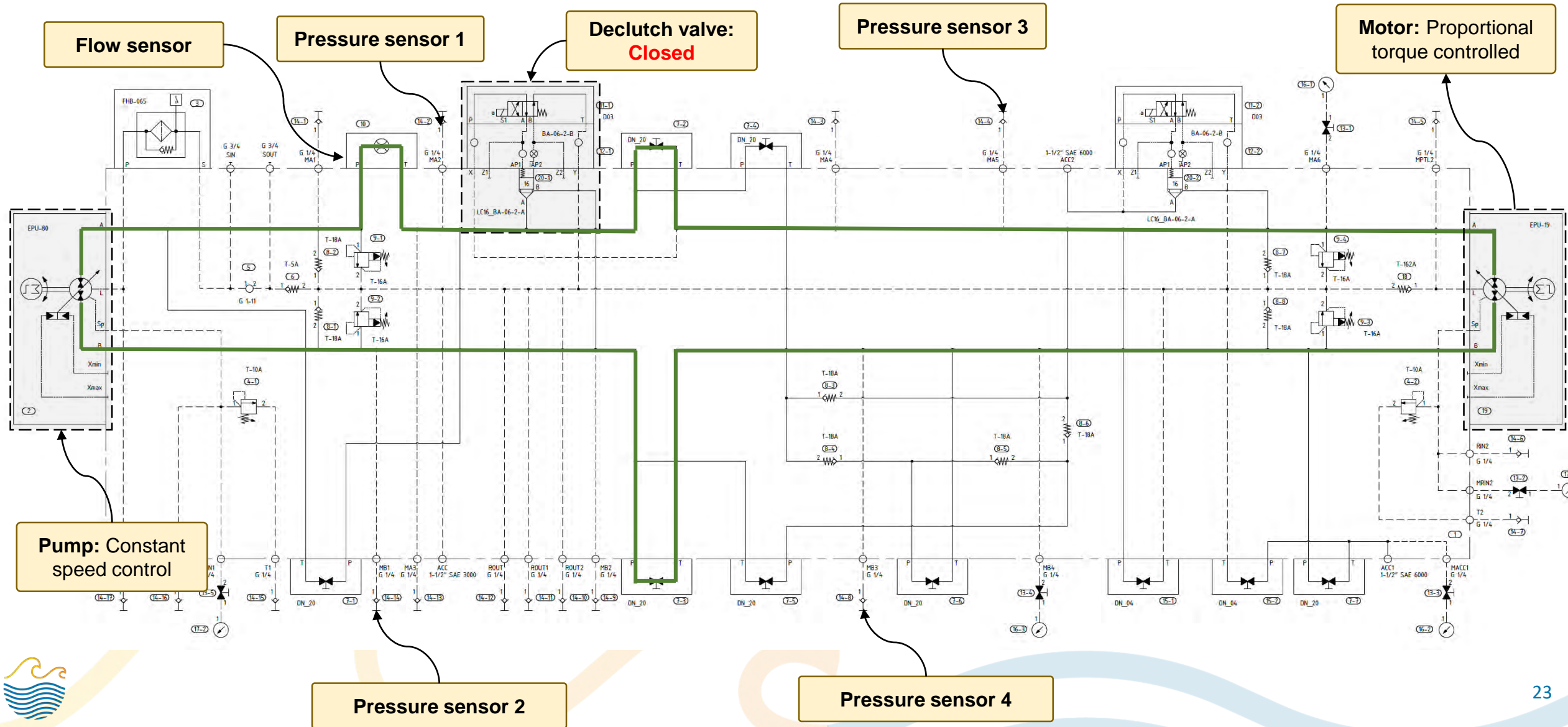
Design tool – Results



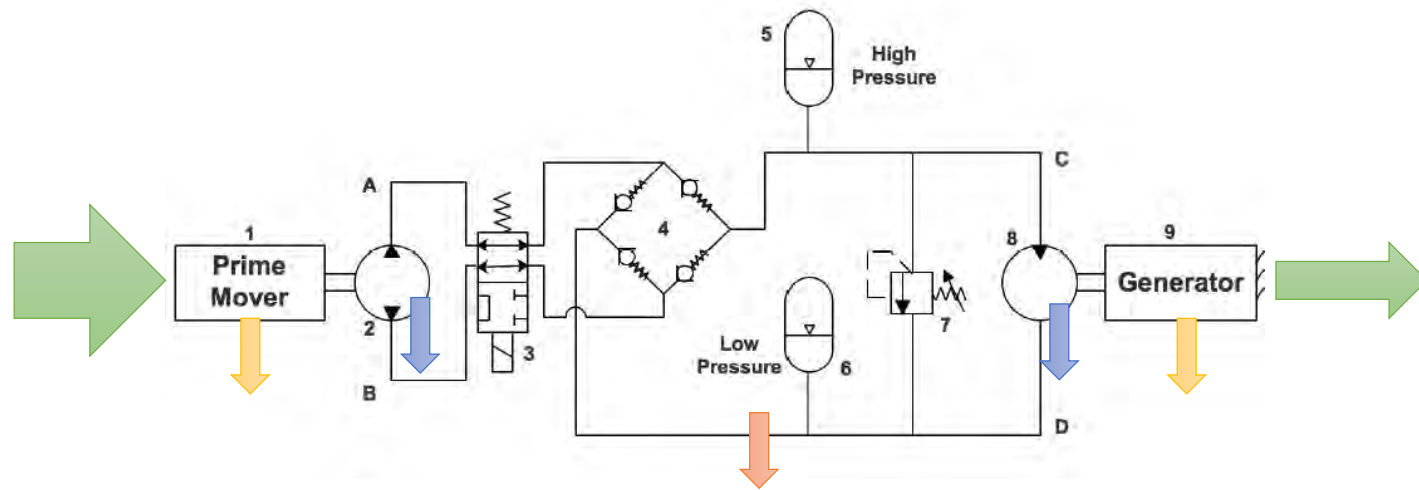
Test rig tests – Operating modes and sensing



Test rig tests – Efficiency tests, full circuit



Test rig tests – Efficiency tests, full circuit



Electric losses

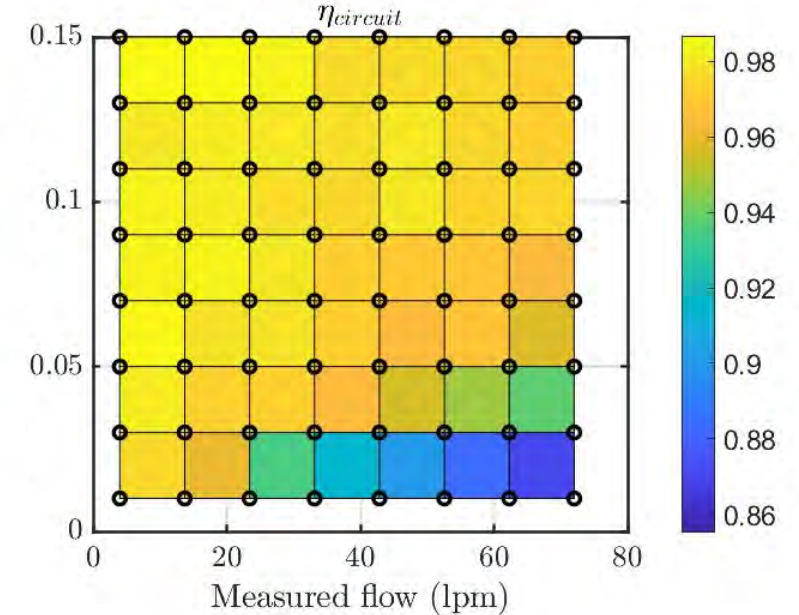
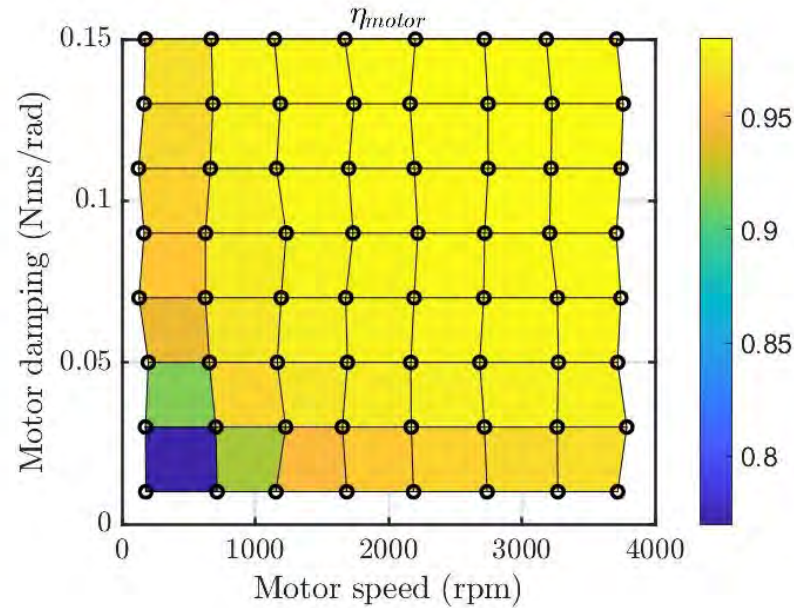
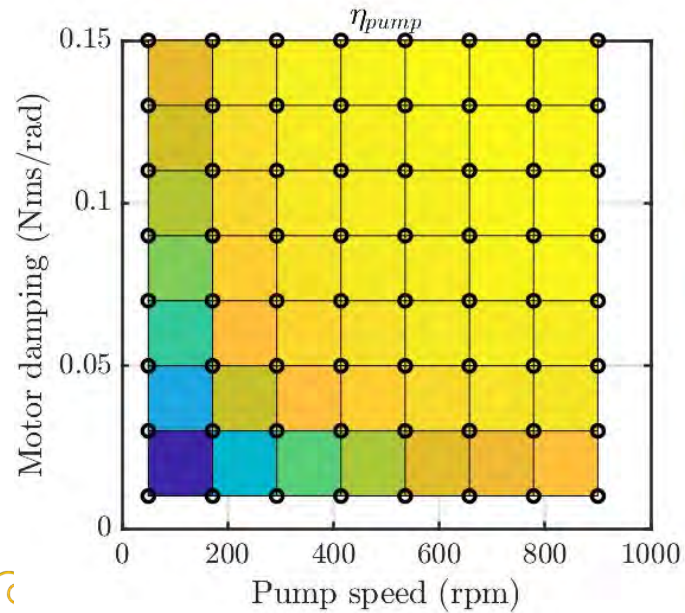
Pump and motor losses

Circuit pressure drop

$$\eta_{pump} = \frac{q_{pump} \cdot \Delta P_{pump}}{P_{in}}$$

$$\eta_{motor} = \frac{P_{out}}{q_{motor} \cdot \Delta P_{motor}}$$

$$\eta_{circuit} = \frac{q_{motor} \cdot \Delta P_{motor}}{q_{pump} \cdot \Delta P_{pump}}$$

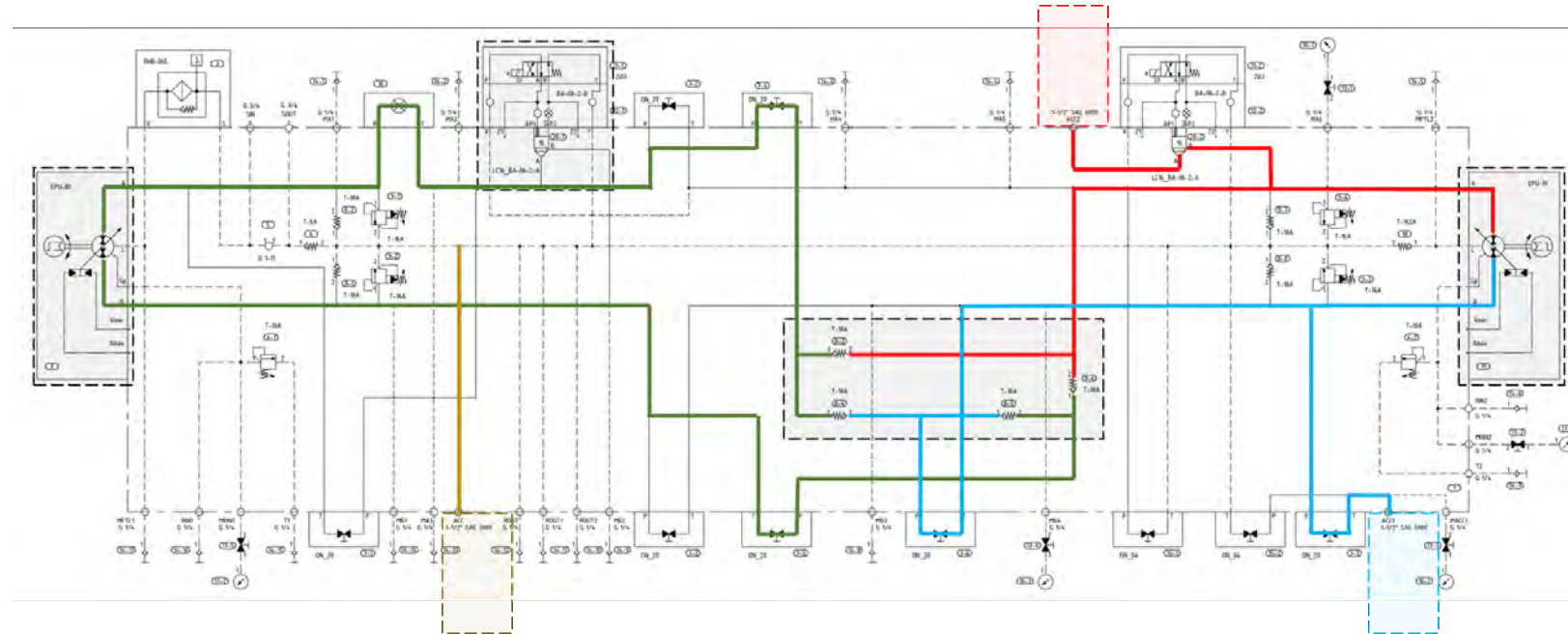


Experimental Validation - WIP



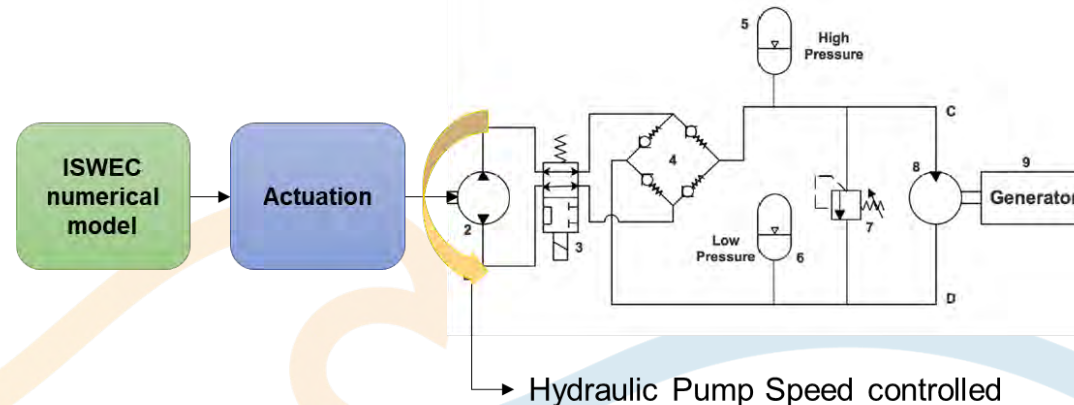
Performance of the Hydraulic PTO

- Controllability
- Extracted power
- Operating conditions efficiency



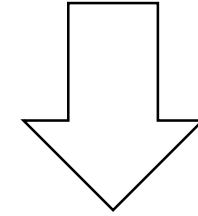
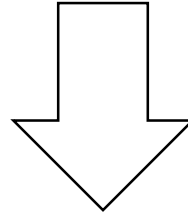
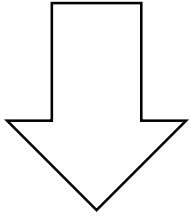
HIL mode

- HIL tests
- Numerical model validation



Conclusions

1 st Year	2 nd Year	3 rd Year
Hydraulic PTO preliminary studies HIL test rig design	Hydraulic PTO advanced non-linear MPC control HIL test rig construction	Hydraulic PTO design-tool HIL test rig experimental tests and validation



→ Design of a Full-scale HIL test rig

→ Controllability studies

→ Increase of ISWEC rated power

→ Advanced control algorithm

→ Relevant improvement of ISWEC
production

→ Numerical model validation

→ Design tool with hydraulic PTO

→ Full scale PTO tests



Grazie per
l'attenzione!

Q&A



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