

Engineering Portfolio

Attila Haas – 14.11.2023

Imperial College
London

Battery Pack for hybrid Go-Kart

Third year design make and test project – [Click here to see full report](#)
Teammates: Aditya Vencatesan Basu, Edward Lee, Edward Wang

Result: A

My
Roles:



Project
Management



Mechanical
design



Engineering
Analysis



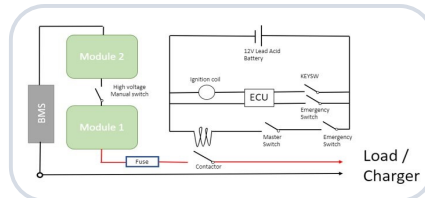
Reporting

Technical Requirements

1. **Life:** Battery needs to last 30 minutes on the track.
2. **Safety:** Withstand debris and vibration; fully insulated
3. **Battery Voltages:** In range 50-120V for motor controller and 12V battery for engine control
4. **Temperature:** To operate within 10°C to 60°C
5. **Cost:** £1000 allocated

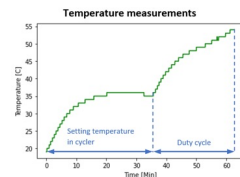
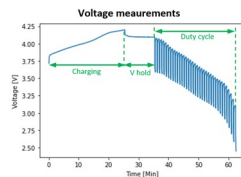
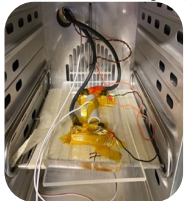
Molice! INR-21700-P42A

- Nominal voltage: 3.6V
- Capacity: 4200mAh
- Maximum current: 45A



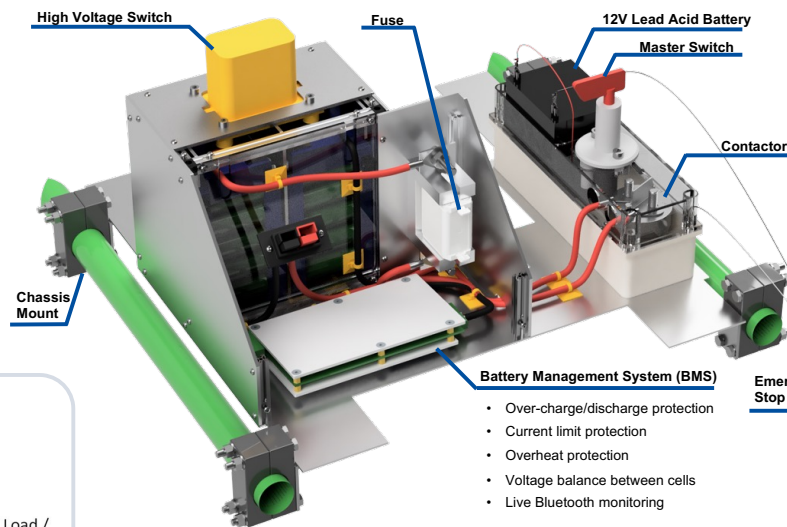
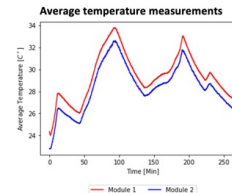
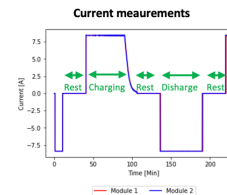
Single Cell test

Single-cell discharge tests were conducted to verify the cell's ability to meet the power delivery requirements and to gain insights into the cell's thermal behaviour during high current draw contingency situations.

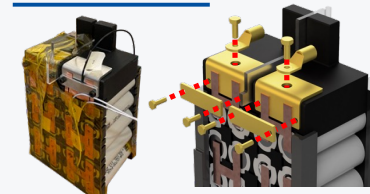


Module test

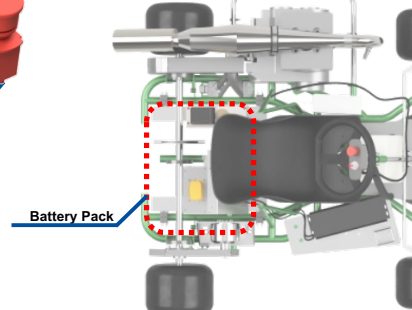
Module discharge tests were conducted on both modules separately to examine their manufacturing quality and thermal characteristics.



Module Assembly



- 2 identical modules of 10S 2P connected using bolted aluminium clamps.
- TIG Welded H shaped Copper busbars were used for cell connections to minimise contact resistances.
- 3D Printed insulation coupled with acrylic sheets for constraining and additional safety.

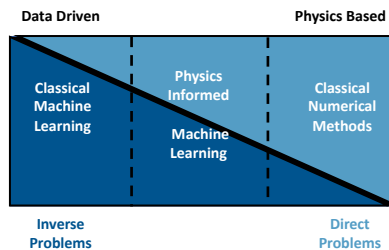


Physics Informed Machine Learning

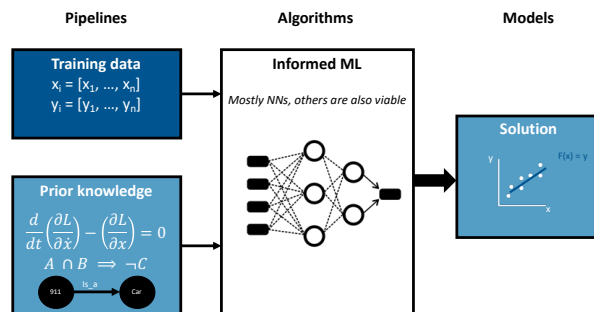
Third year literature review – [Click here to see full report](#)

What is PIML?

Physics Informed Machine Learning (PIML) is the middle ground between classic numerical methods and "black box" ML methods.



Rueden et al. (2021) define Informed Machine Learning as an ML algorithm that learns from multiple, independent pipelines consisting of both training data and prior knowledge.



von Rueden, L., Mayer, S., Bech, K., Georgiev, B., Giesselbach, S., Heese, R., Kirsch, B., Pfrommer, J., Pick, A., Ramamurthy, R., Walczak, M., Garcke, J., Bauckhage, C. & Schuecker, J. (2021).

'Informed Machine Learning – A Taxonomy and Survey of Integrating Knowledge into Learning Systems', IEEE Transactions on Knowledge and Data Engineering pp. 1–1.

How can Physics be incorporated?

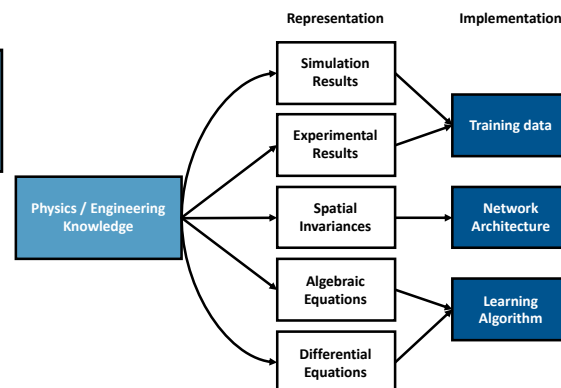
Firstly, the most straightforward method is to convey information in the training data itself. This quintessentially does not differ from a traditional ML approach, however with physics for example, multiple pipelines of data can be considered. One could use a hybrid of both experimental and simulation results as training data, thus relying on both previously discovered physical rules and measurements. This not only allows for significant cost reductions in data acquisition, but also to reduce noise.

Secondly, the ANNs architecture can be altered to help it better optimise for certain solutions. Only changes that have real world meaning are considered a second pipeline. Mapping certain neurons that are known to represent parts of a logic statement to particular neurons is a prime example. CNNs used to add translation invariance is another.

Thirdly, the learning algorithm can be tweaked. The cost function of a NN can usually be altered

in a way for the results to conform to known algebraic or differential equations. As one would

expect this is the most widely used method in PIML, as most systems can be described by known relationships.



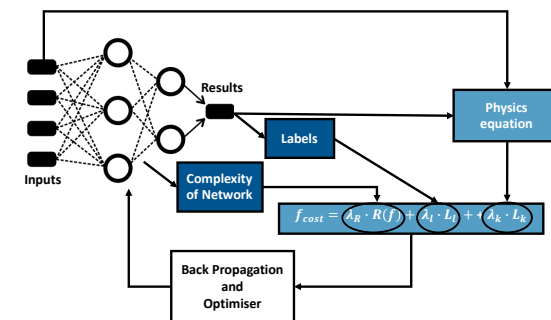
The most common method for NNs

The full report goes into many sub-variants of PIML implementations, see the link. For the sake of conciseness, here only the most common NN solution will be showcased

As common with NNs there is a set of training data to which the network will be trained with a cost function.

In examples involving physics, a results accuracy can be judged on existing laws. For example, if the energy isn't conserved, during a process, the output is known to be false. To guide the NN towards solutions that obey the known physical laws, the cost function is expanded with a term which penalises deviations from the known relationships. This allows the NN to not only confirm to measured data, which is subject to noise, but also follow laws of nature.

(Diligenti et al. 2017)



Diligenti, M., Roychowdhury, S. & Gori, M. (2017), Integrating Prior Knowledge into Deep Learning, in '2017 16th IEEE International Conference on Machine Learning and Applications (ICMLA)', pp. 920–923.



Research



Reporting

Sailing Vessel Energy Harvester

Second year design project – Spring

Teammates: Alexander Christopherson, Jansen Papworth, Kayman Krishnamohan

Result: A

My
Roles:



Mechanical
design



Engineering
Analysis



Reporting

Aim

Develop a sailing vessel energy harvester which powers onboard electrical systems including navigation, lights and a laptop while also charging a battery to allow continued use of electronics while stationary.

Target Market & Expected Use

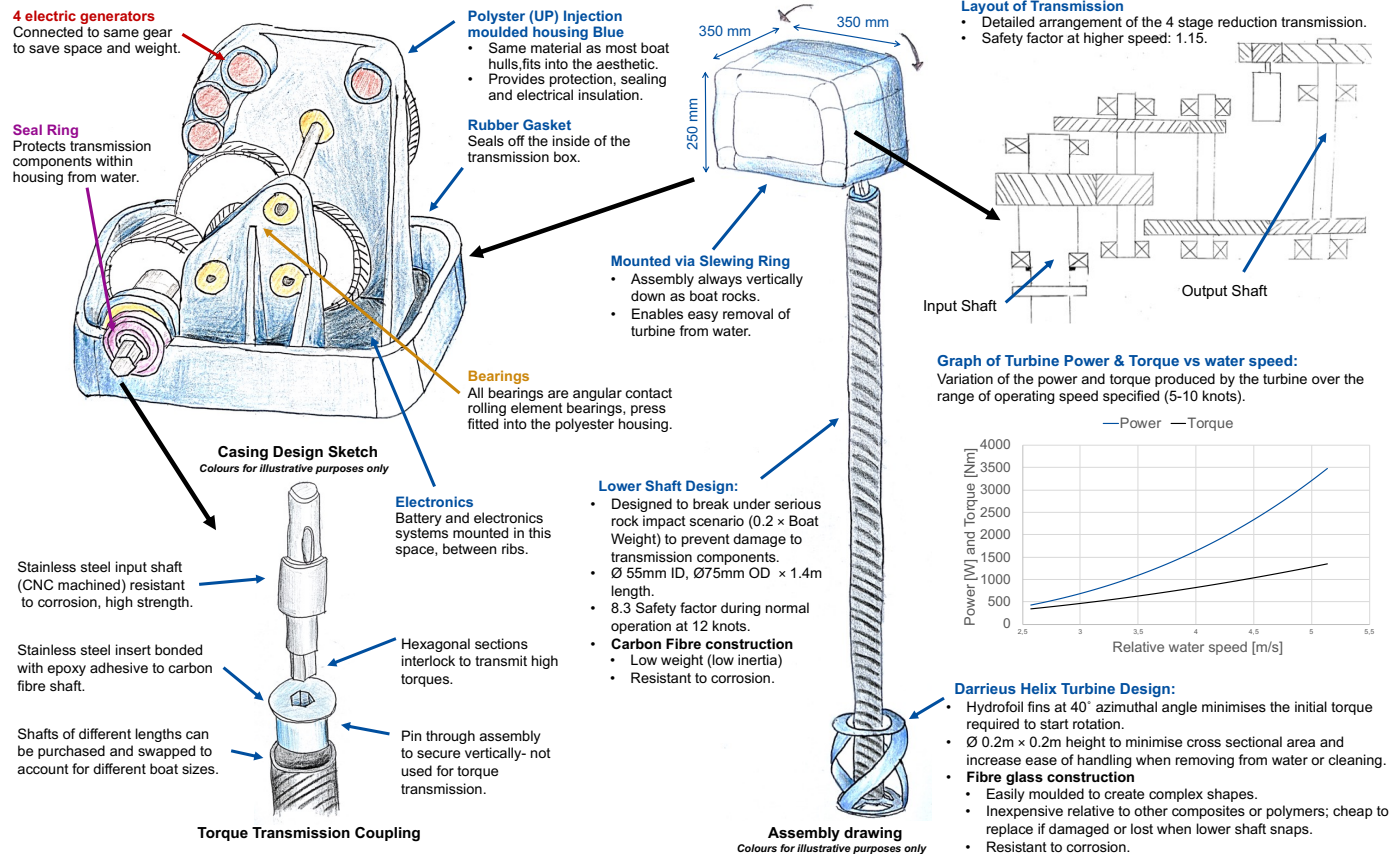
Initially designed for 30+ ft sailboats (14% of US sailboat market). Boats this size are often used for day trips with pauses at tourist locations or scenic spots.

- In use during comfort sailing, i.e. 5-10 knots (2.57 - 5.14 m/s)
- Suitable for use in both fresh and saltwater conditions
- Used for 12 hours a day, 30 weeks of the year; 30ft+ sailboats often chartered for trips so in continuous use
- Must power navigational systems (110W, 12V), lighting (122W, 12V) and a laptop (60W, 19.5V)
- Can be mounted easily and safely by 2 people (18kg)
- Retail price approximately £3000

Technical Specifications

- 290 W generated at 5 knots, 440 W at 10 knots using 4 RS-655VA-28118 motors.
- Battery capacity of 81 Ah.
- Made up of 81 18650 cells, in a hexagonally close packed arrangement.
- 4 stage reduction gear box using helical gears
- Total reduction ratio of 1:525.
- Bearing L_{10} life 82000 hours, 33 years of expected use

Embodiment Design



Gravity light

Second year design project – Autumn

Teammates: Ore Pelumi, Ashay Dhingra, Diego Sanchez Loarte, Suheyb Adam

Result: A-

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Roles:



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Problem:

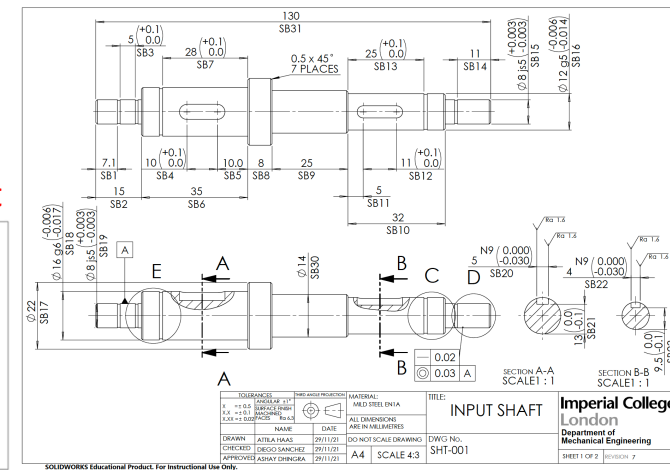
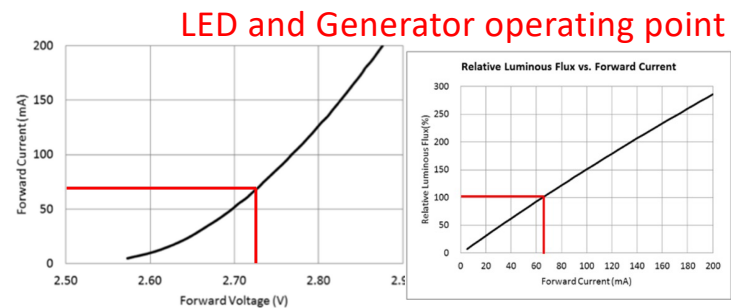
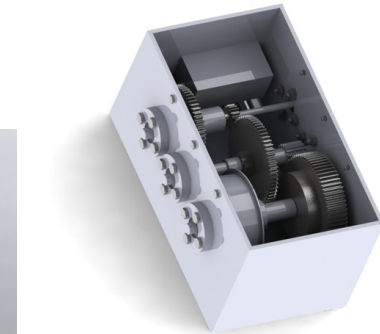
"Design, build, and test a so-called gravity-light: a reading light powered by a suspended mass that is slowly lowered, for use in locations where access to mains power and/or batteries is limited."

- Group 3-month project

Solution:

Sheet metal box houses the generator, transmission and pulley to release potential energy at an optimal pace:

- Provides about 6 minutes of light
- Simplistic design allows for cheap manufacturing methods
- A prototype has been built in the workshop



Brick Hoist

First year design project

Result: A

My
Roles:



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Presenting

Problem:

"Given a support structure and motor, design an actuation method and transmission to raise 50 bricks between single floors of a house!"

- Individual, 3-month project

Solution:

H-bridge inspired chain drive drives the platform

- Leaves the packing area open, and allows for fully lowering

Triple reduction gearbox provides the coupling between motor and the chain

- Compact, cheap design, easy to maintain

