



Musically Embodied
Machine Learning

Introduction



Ffroeds

by Luuma



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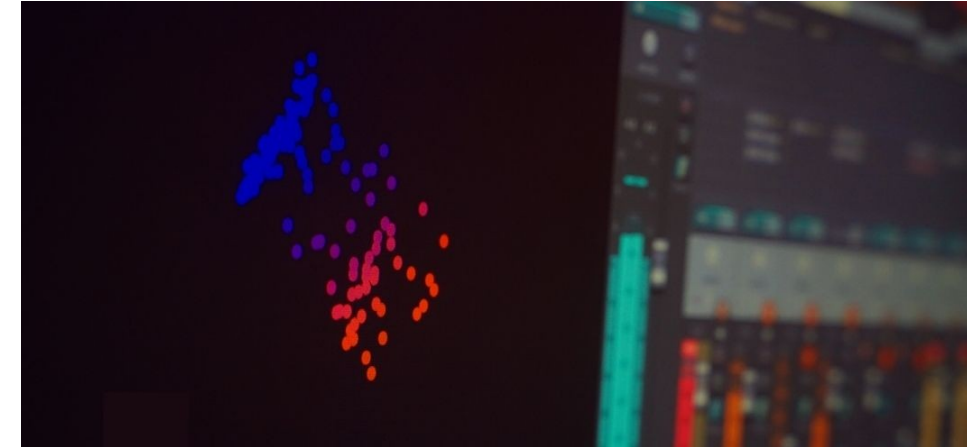
Musically Embodied Machine Learning

- AHRC Fellowship Project
- 18 months
- People
 - Chris Kiefer – PI (Music / Emute Lab)
 - Andrea Martelloni – Research Fellow (Music / Emute Lab)
 - Nic Seymour-Smith - Researcher Software Engineer (SHL)



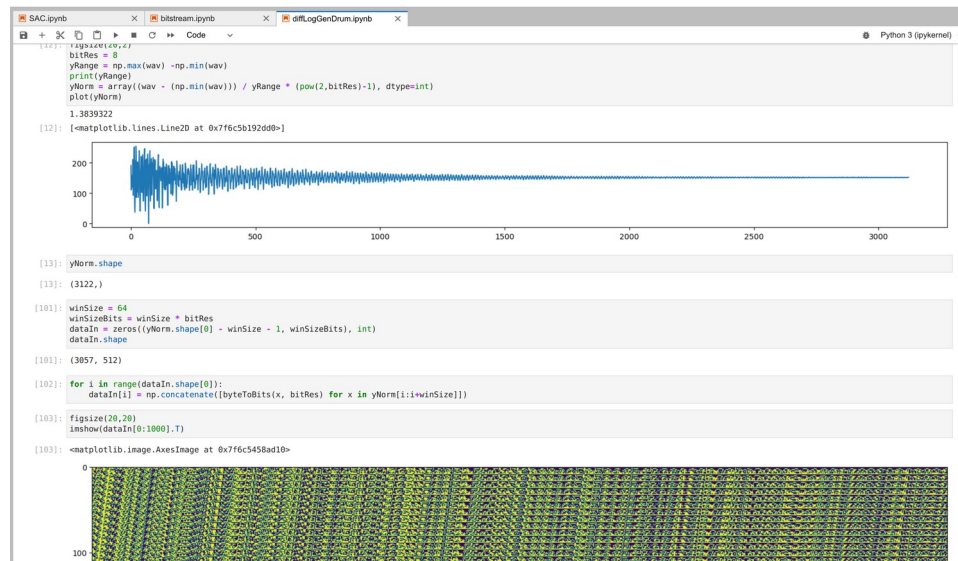
Expressive Potential of ML in Musical Instruments

- Gesture Processing
 - Mapping design
 - Sound synthesis
 - Signal Processing
 - Generative pattern / sound production
-
- Wider possibilities:
 - Generating organic complexity / encouraging serendipity
 - Creating adaptive (personalisable) instruments
 - Opening up complex DSP to musicians without technical experience
 - Shared agency
 - New creative possibilities e.g. latent space exploration, network bending

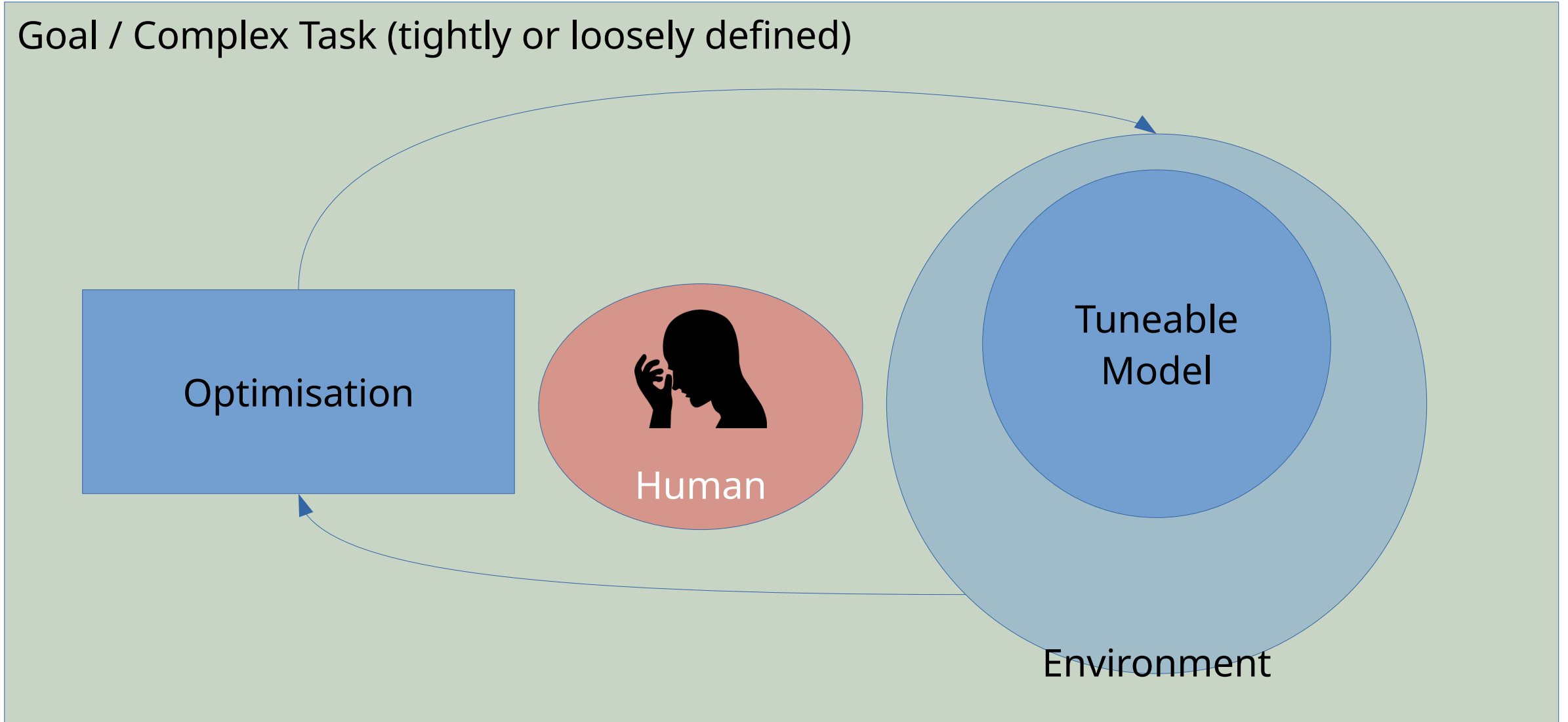


Problems with ML and Musical Instruments

- Disembodied split between instrument and machine learning toolsets
- Focus on static pre-trained models (hardware creates constraints)



Conceptualising Creative ML



ITMLs

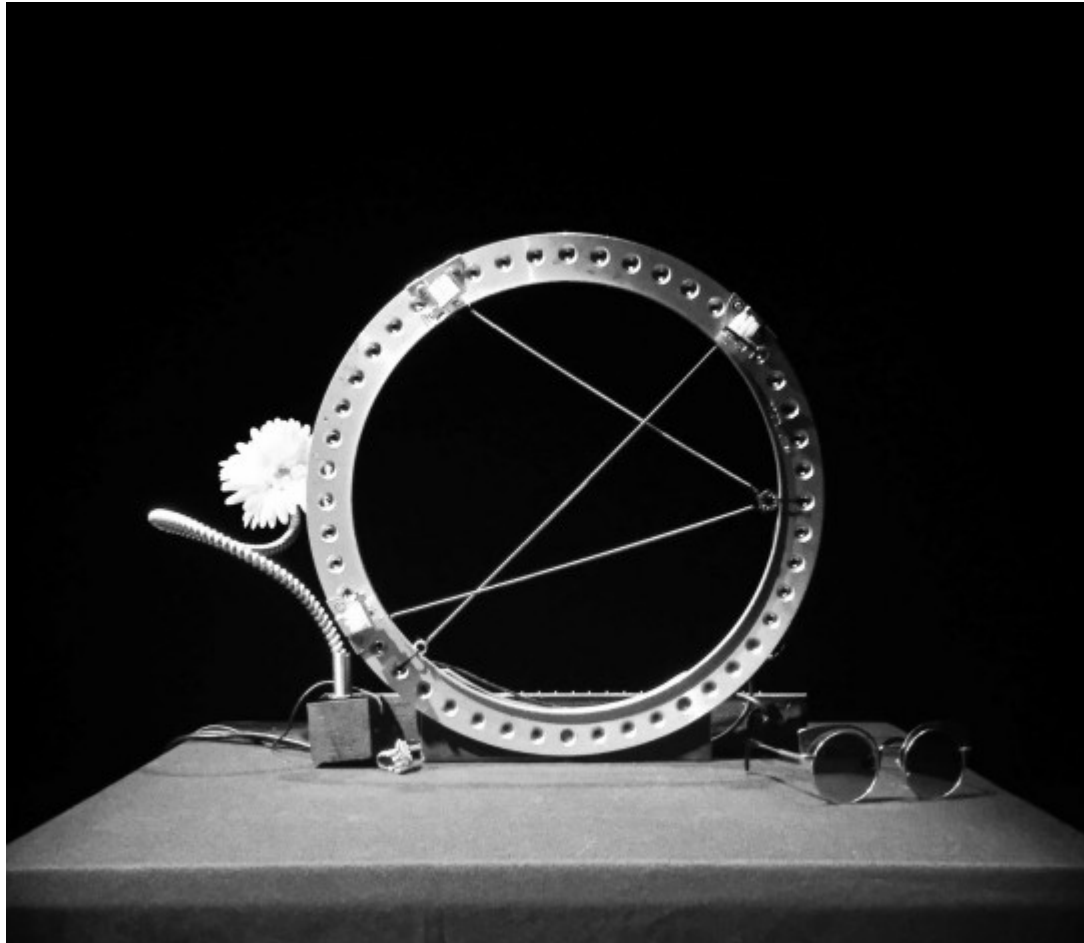
- Instruments with Tuneable Machine Learning
 - The entire process of training and running models is embedded within the instrument
 - Unified interface with the musical instrument
 - Using machine learning *musically*
 - Implications/challenges
 - Low-power portable computing
 - Simplified interface to machine learning
 - Instruments that adapt and evolve over their lifetime

Research Questions

- **Musical Practice** When musicians play ITMLs, what are the approaches they develop to playing and tuning the instruments?
- **Processes in Musical Machine Learning** What are musical, technical and ethical considerations that emerge when designers and musicians work with ML models that can be adapted during the ongoing life of an instrument?
- **Design and Technology** Considering the new creative possibilities opened up by instruments with tuneable ML, what design techniques and principles can support their development? Furthermore, how can the complexity of contemporary ML be encapsulated into simpler interfaces, that integrate with the playable feel (or ergonomics) of an instrument?



Example Instrument: Spire/Wekinator (Sonami, Fiebrink)

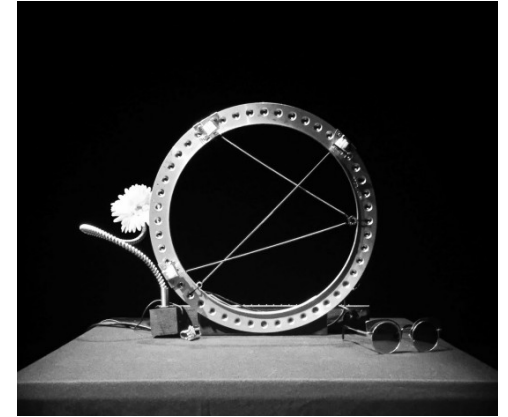


Fiebrink, R., & Sonami, L. Reflections on eight years of instrument creation with machine learning. NIME 2020.

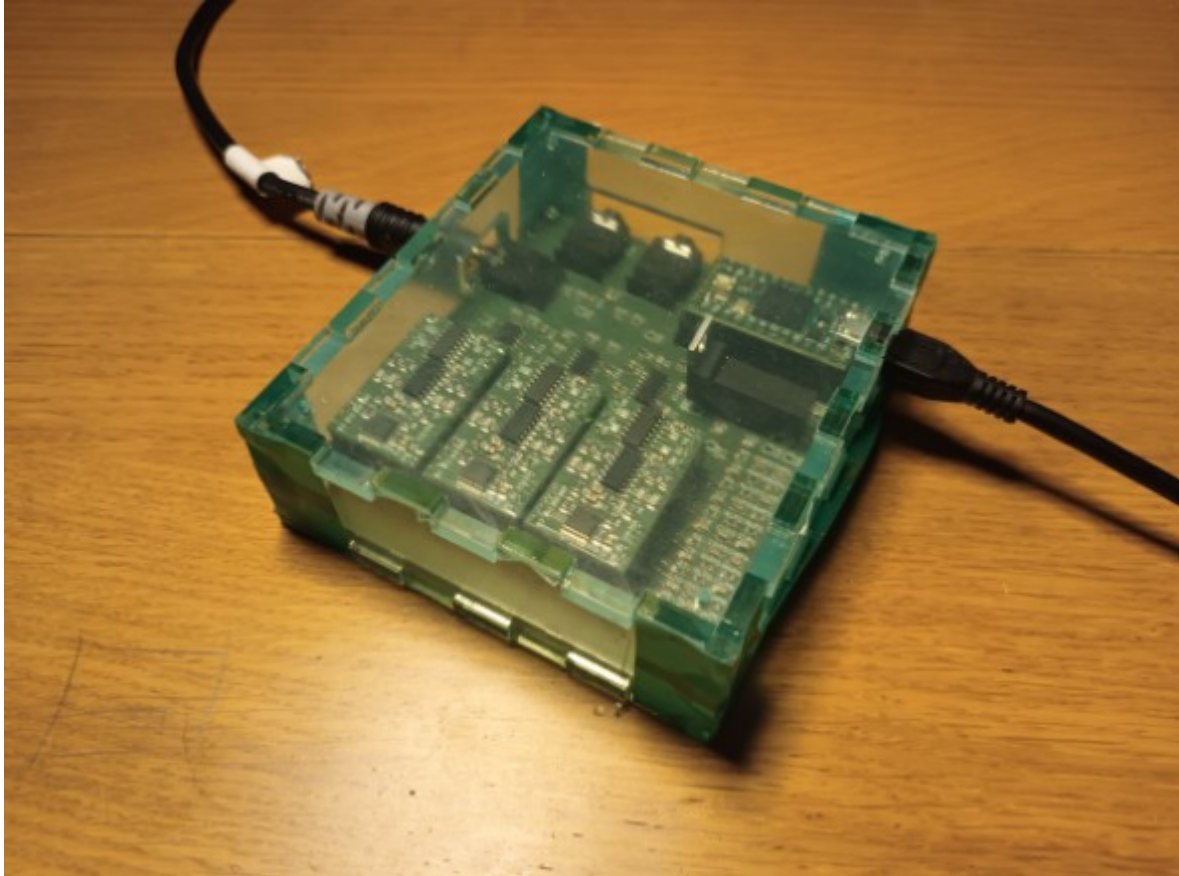
Example Instrument: Spring Spire/Wekinator (Sonami, Fiebrink)

Q: Can you say more about how you think about mapping in your instruments and compositions? How does ML impact on the task of creating a mapping?

A: Mapping is the backbone of a composition. While it has been argued that in electronic instruments, controllers, sound generators and links (mappings) are independent units, I believe those to be tightly correlated. The choice of control inputs dictates the gestures, hence possible mappings and resulting musical events. While people using more standard controllers such as faders and buttons might apply control as an afterthought, mapping strategies have been an essential component in elaborating pieces with the *lady's glove* and now with the *Spring Spire*. Mapping bridges the physical world to the sonic world. ML offers a way to easily configure mappings with a wide variety of behaviors, thus allowing the composer to focus on the sounds and compositions. I cannot envision at this point in time any other ways to map.



Example Instrument: Digital No-Input Mixer (Elia & Overholt)



Example Instrument: Digital No-Input Mixer (Elia & Overholt)

the FluCoMa project [12], [24]. It works by manually assigning 2d coordinates to a number of interesting points from parameter space, and then training a Multilayer Perceptron to learn a non-linear mapping from 2d coordinates to parameter values in the instrument's higher-dimensional parameter space. The network can then be fed with 2d coordinates to provide instrument control parameters using the learned non-linear mapping.

Methods

- Long-term participatory design with practicing musicians
- Probe instruments
- Workshops
 - Sound and Music Computing, Aalborg Universitat Copenhagen (PhD Course running in January)
 - Intelligent Instruments Lab, Reykjavik
- Autobiographical design
- Industrial collaborations
 - Schlappi Engineering
 - Emute Lab Instruments



Embedded/Embodied Models and TinyML

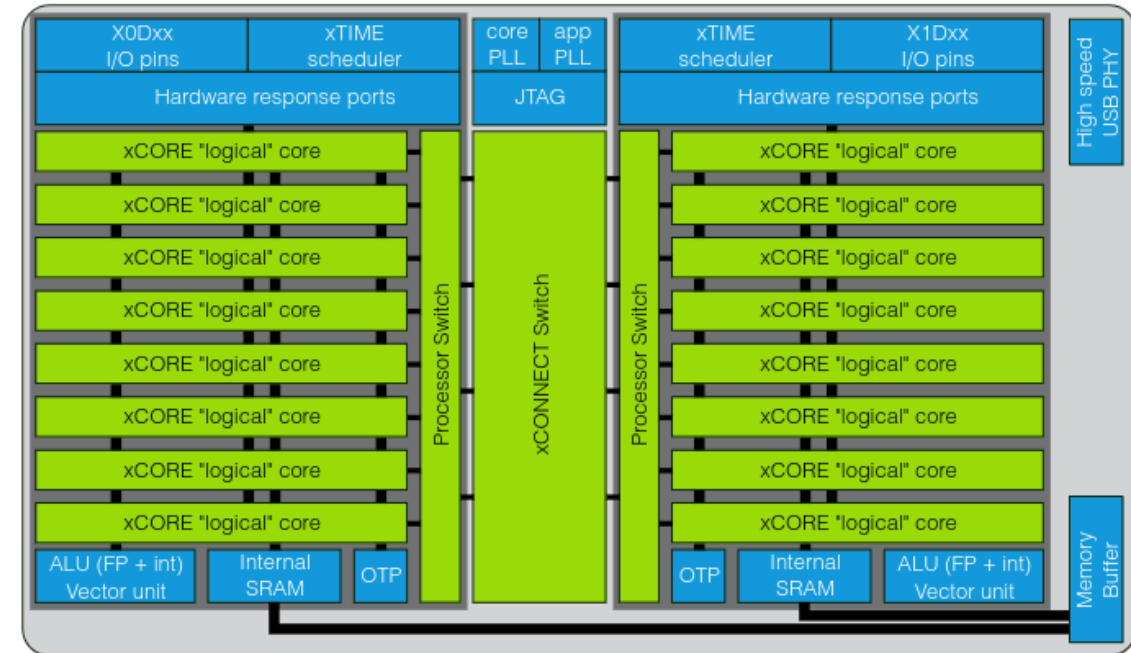
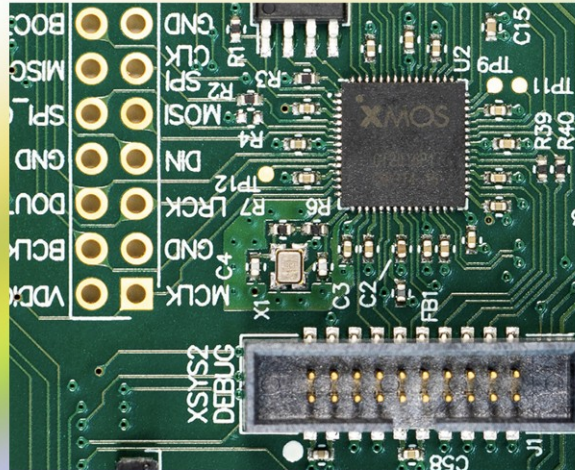


Question: What's the value of hardware/hybrid instruments where machine learning is self-contained?

Hardware Platforms: XMOS



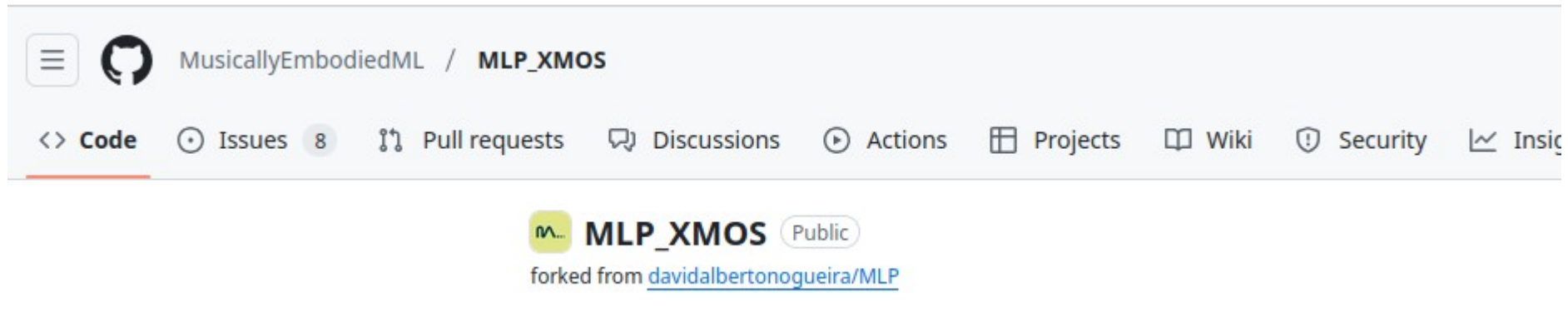
Bringing technology to life



Hardware Platforms: RP2350



Software: MLP Library for Microcontrollers



Workshop Report

CHIME Annual One-day Music and HCI Conference 2024



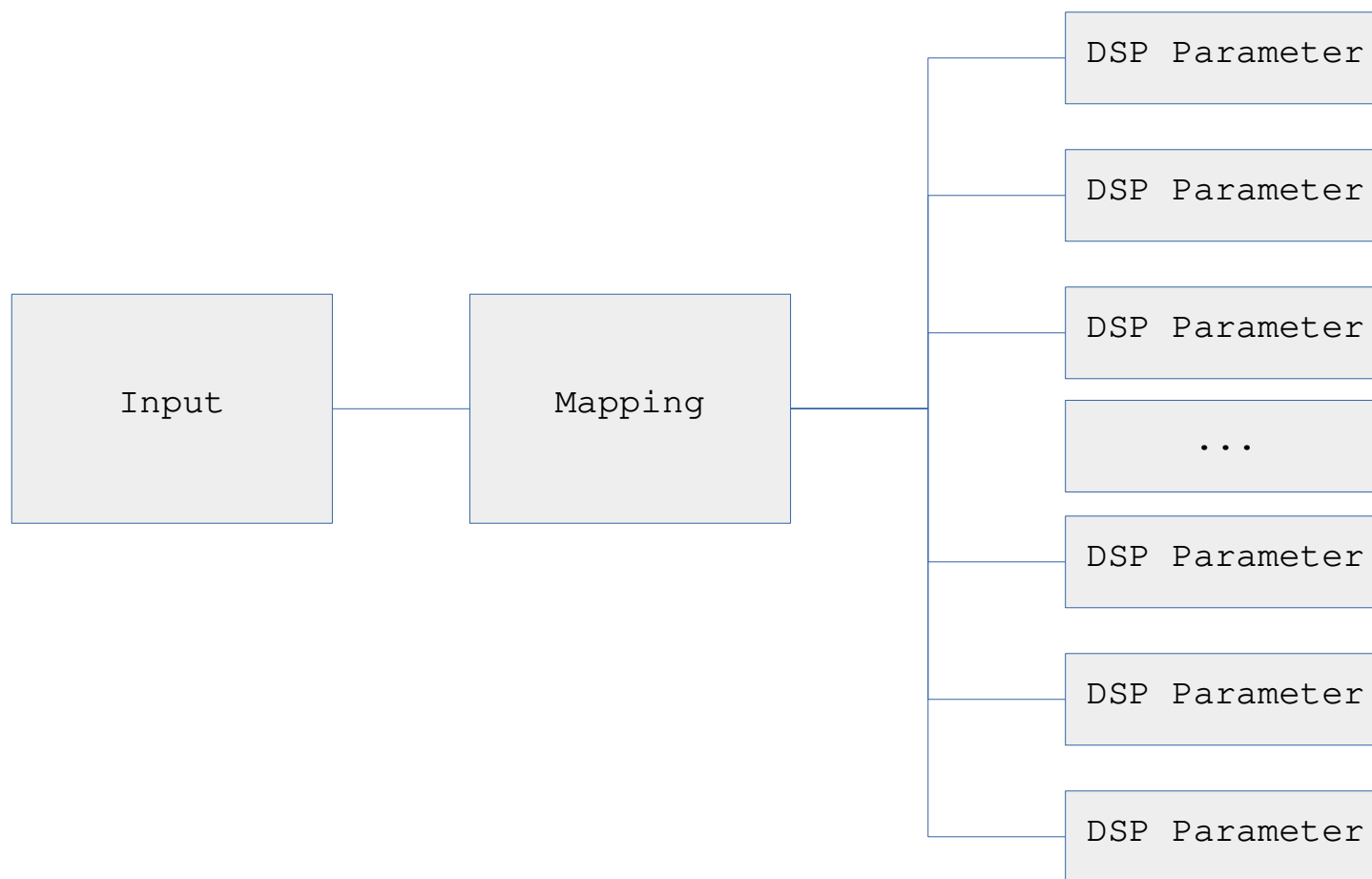
The Open University, in Milton Keynes, UK

One-day Conference: **Monday 2 December 2024,**

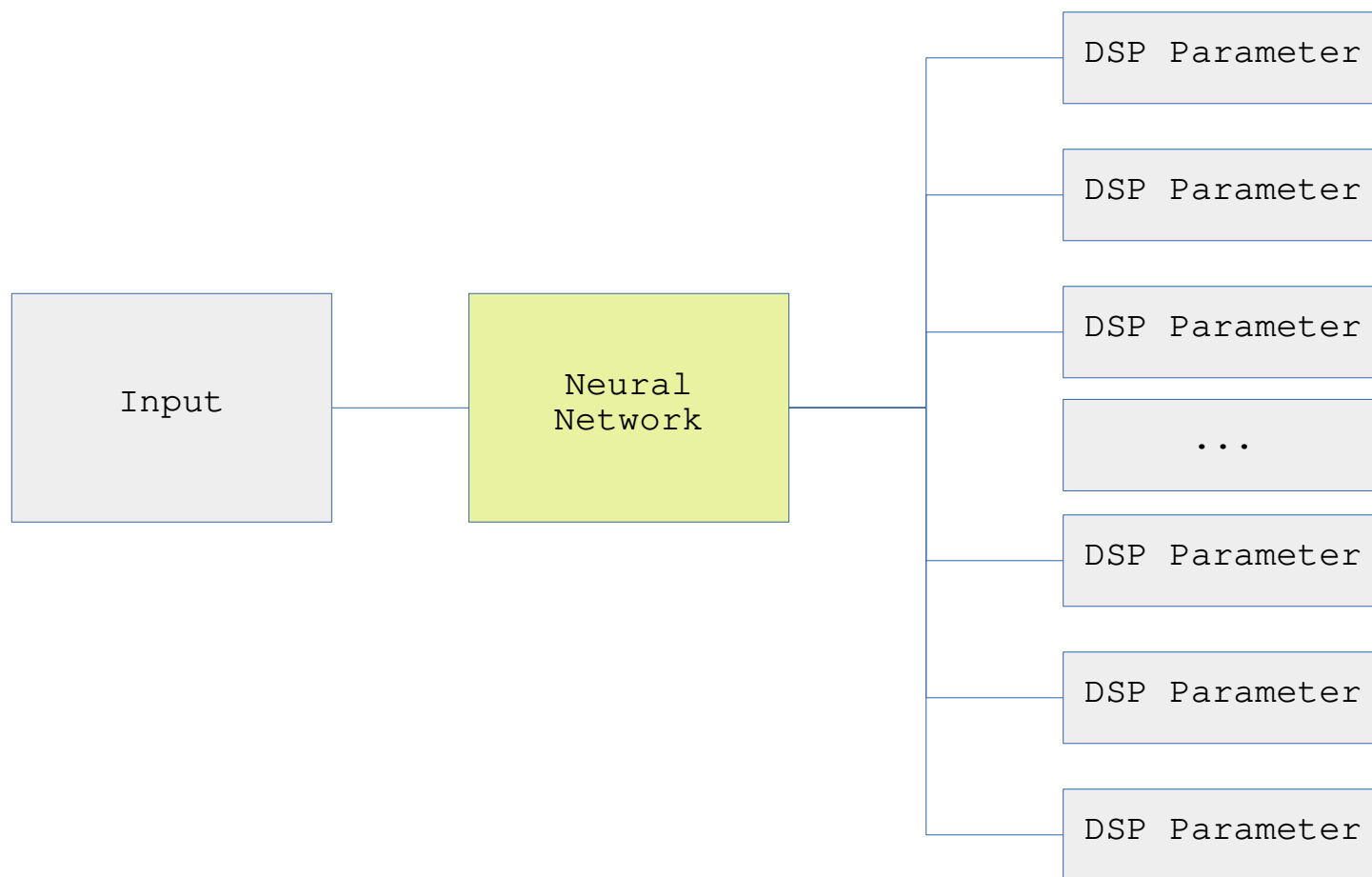
Workshops: **Sunday 1 December 2024.**

Question: How do we embed training and inference into musical instruments in musical ways? What interactive workflows are valuable?

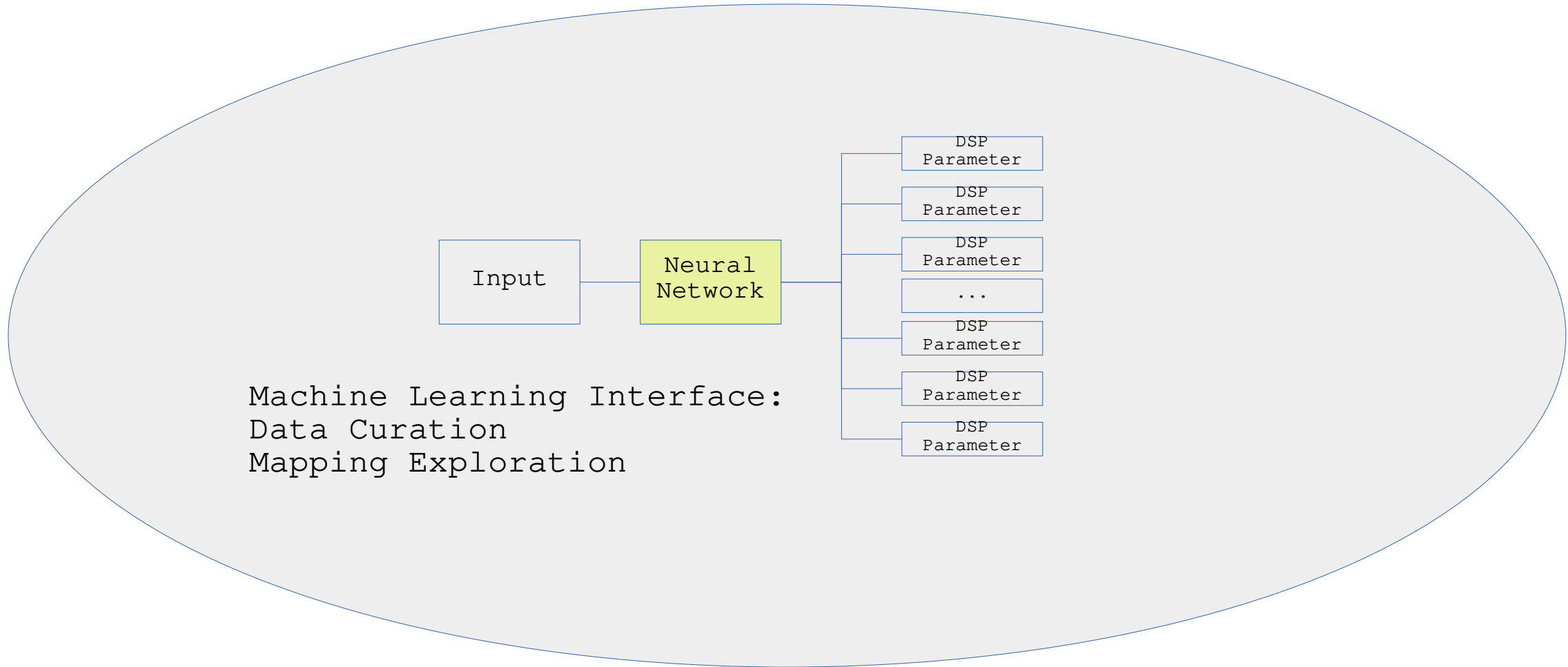
Testing Instruments with Parameter Space Mapping



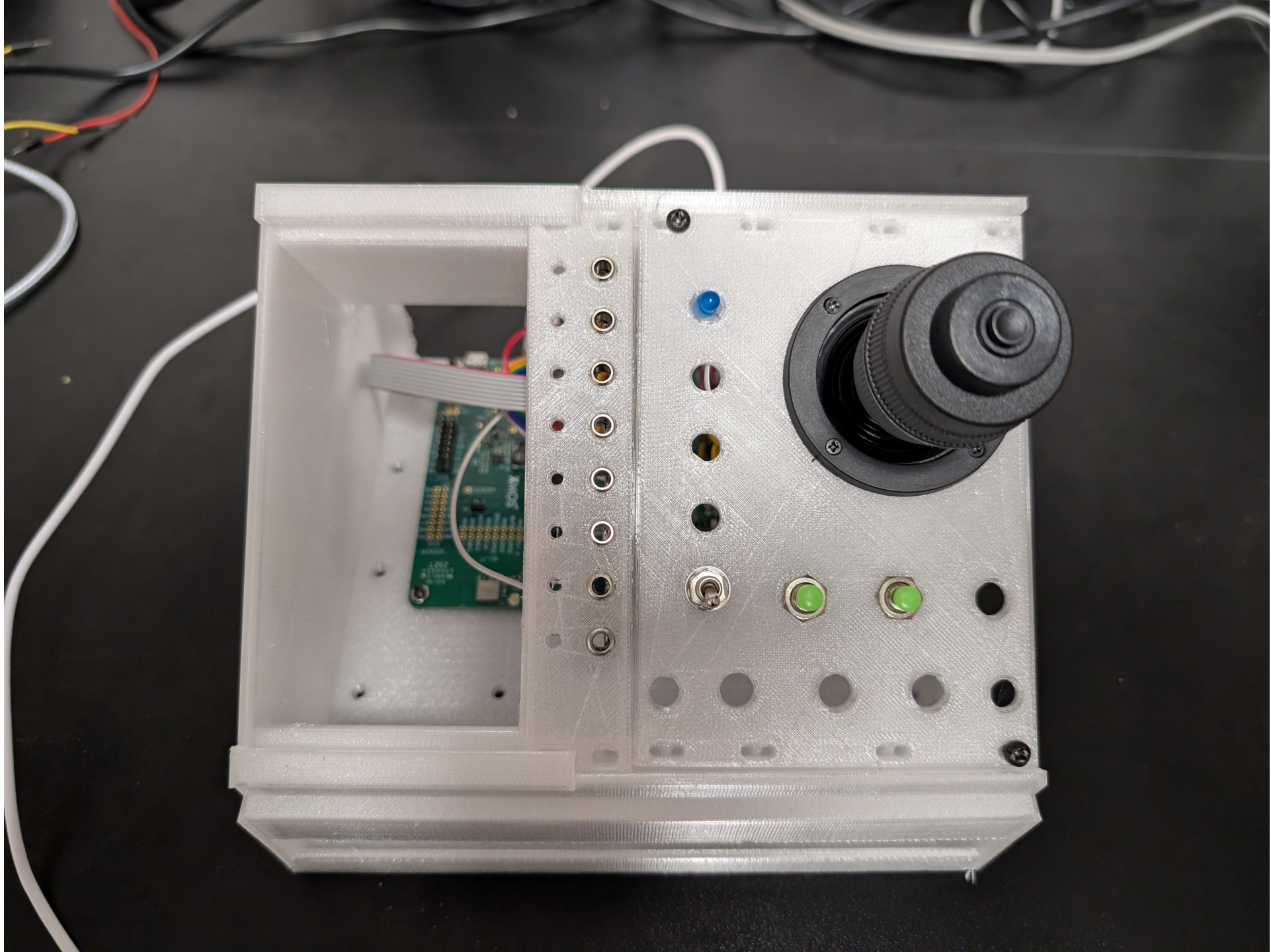
Testing Instruments with Parameter Space Mapping

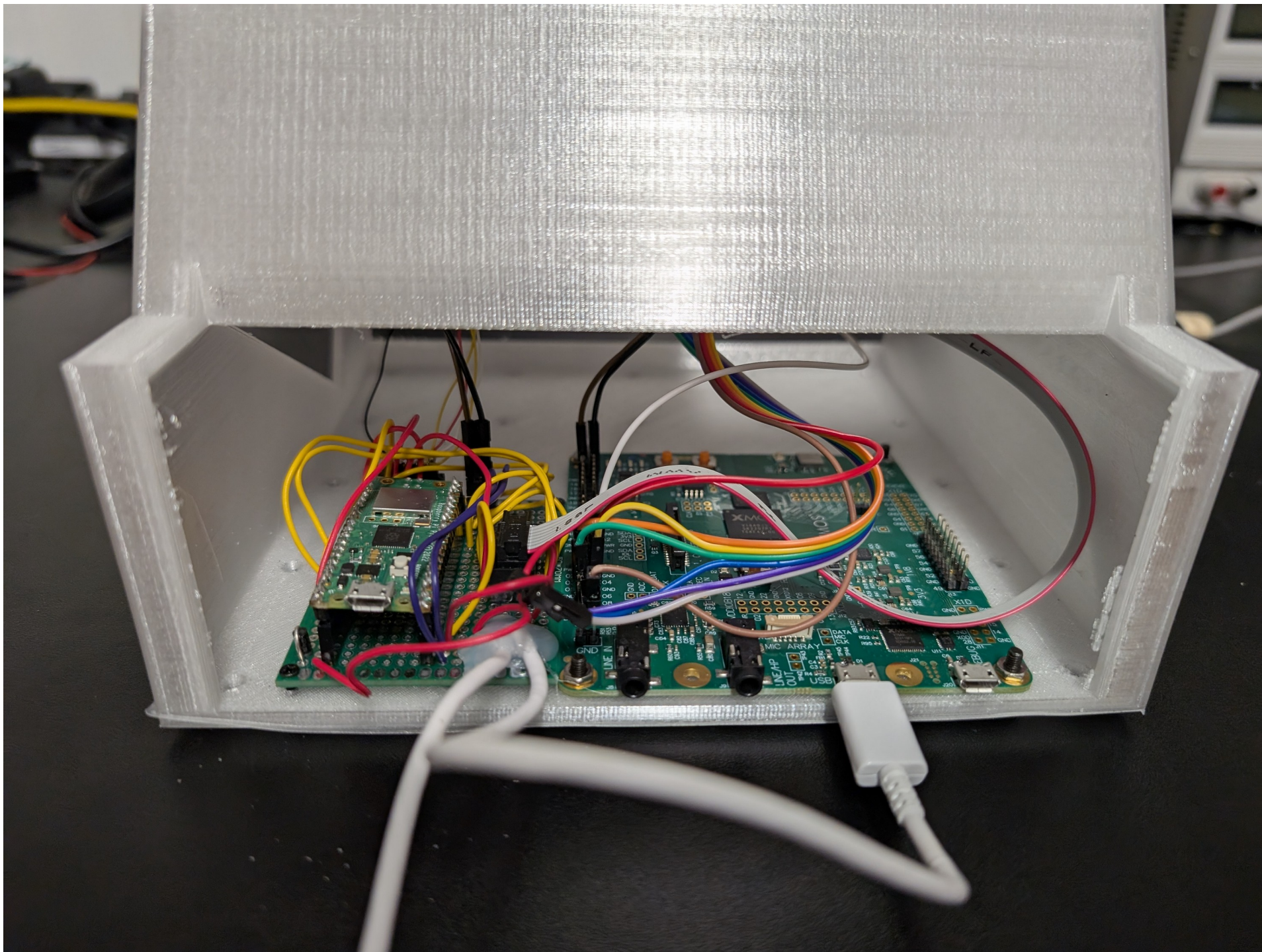


Testing Instruments with Parameter Space Mapping

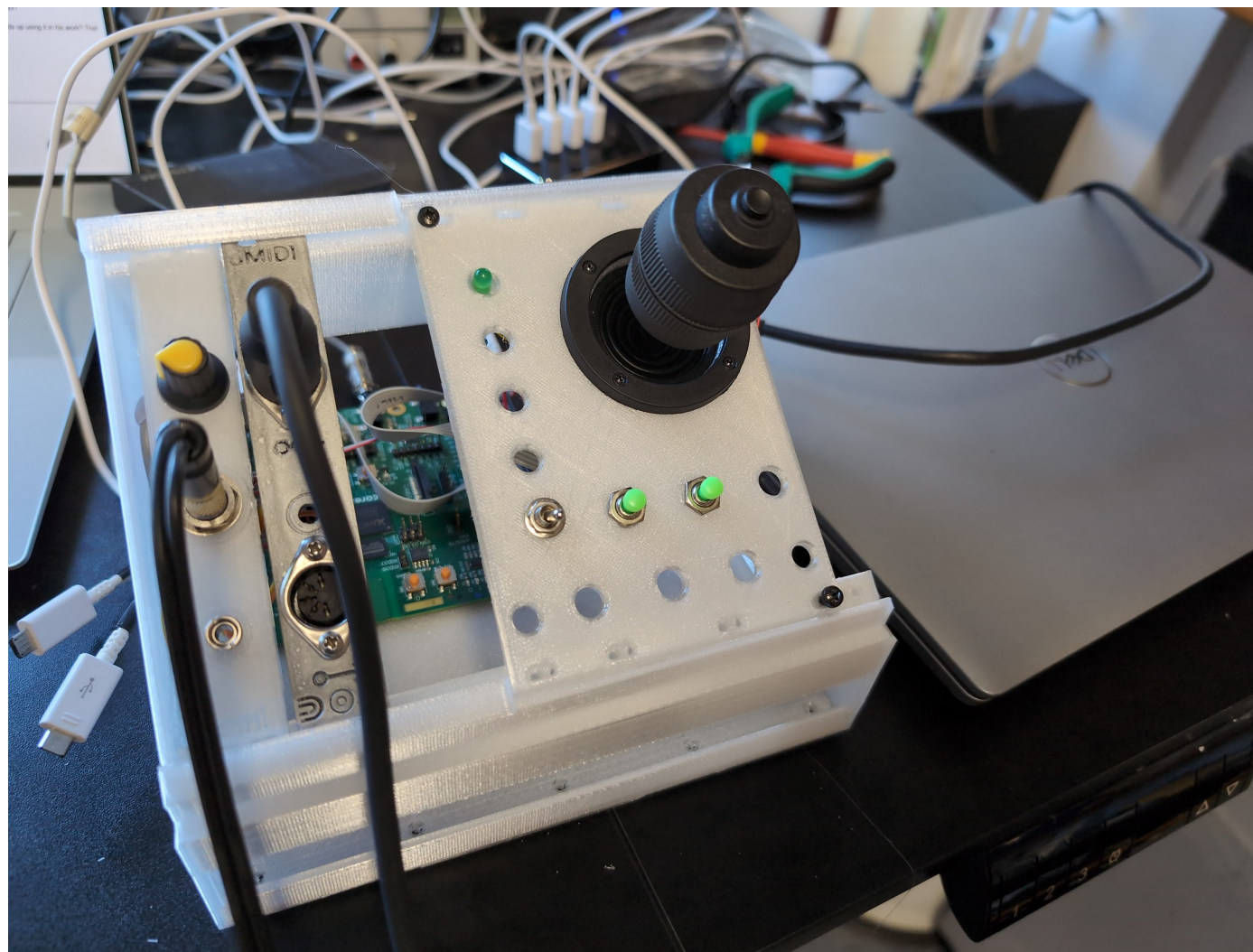


Research Instruments: Three Variations on a Theme

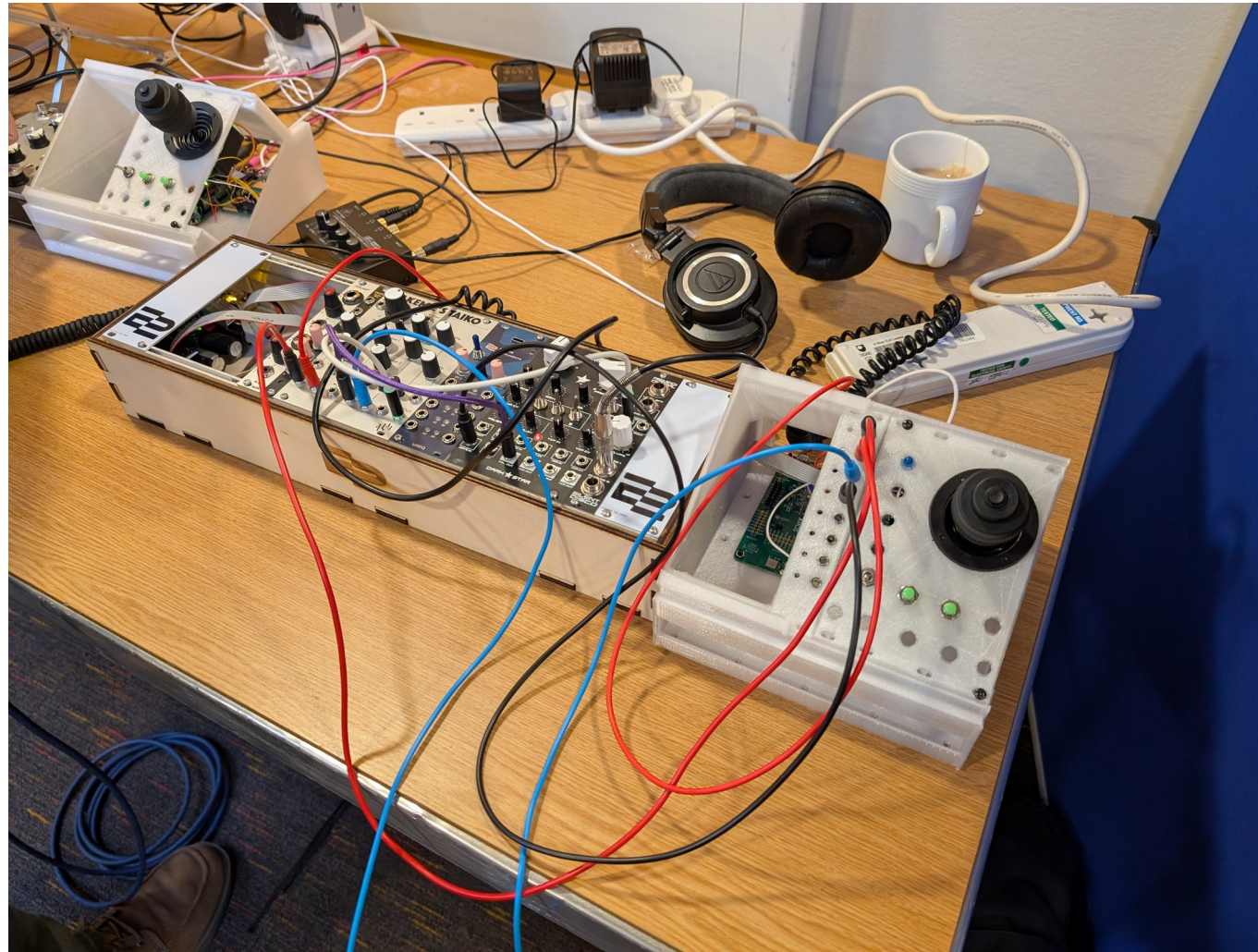




FM Synth



Multi-euclidean Sequencer

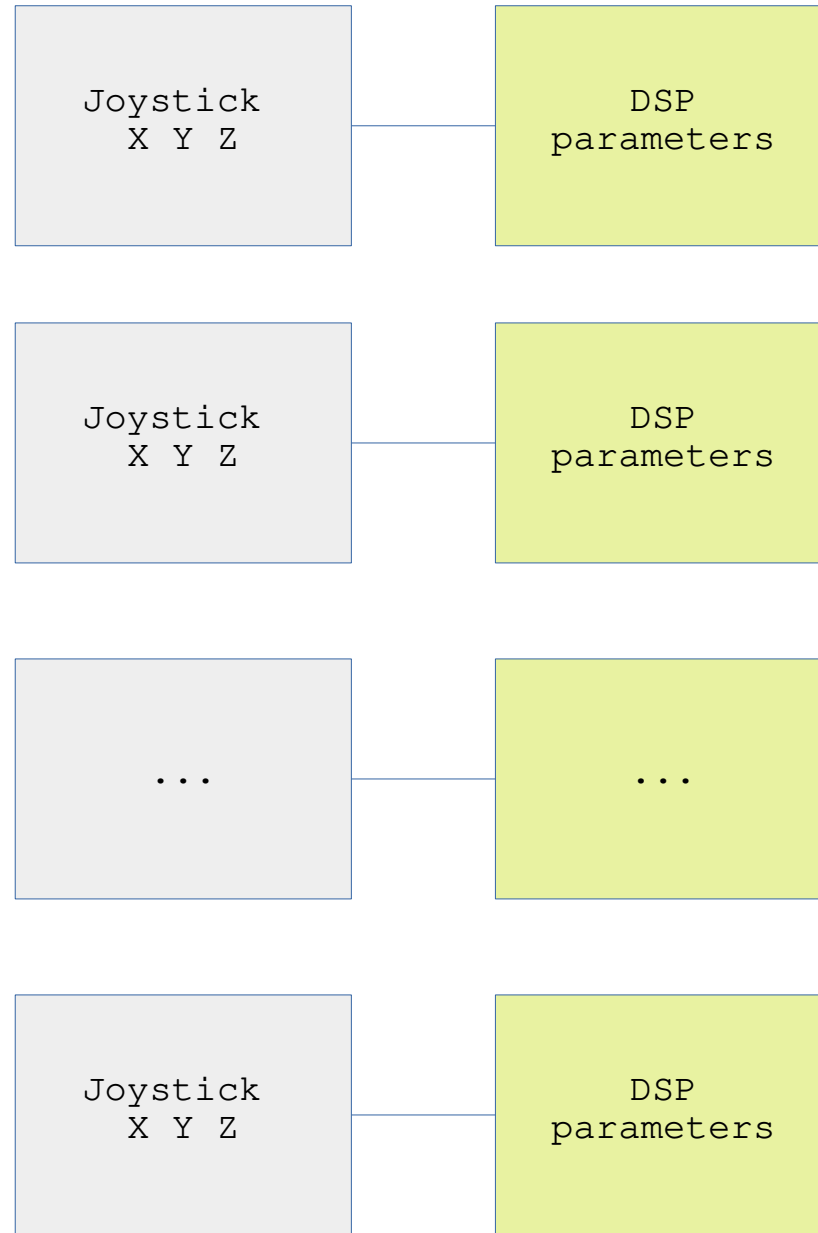


Multi-Fx

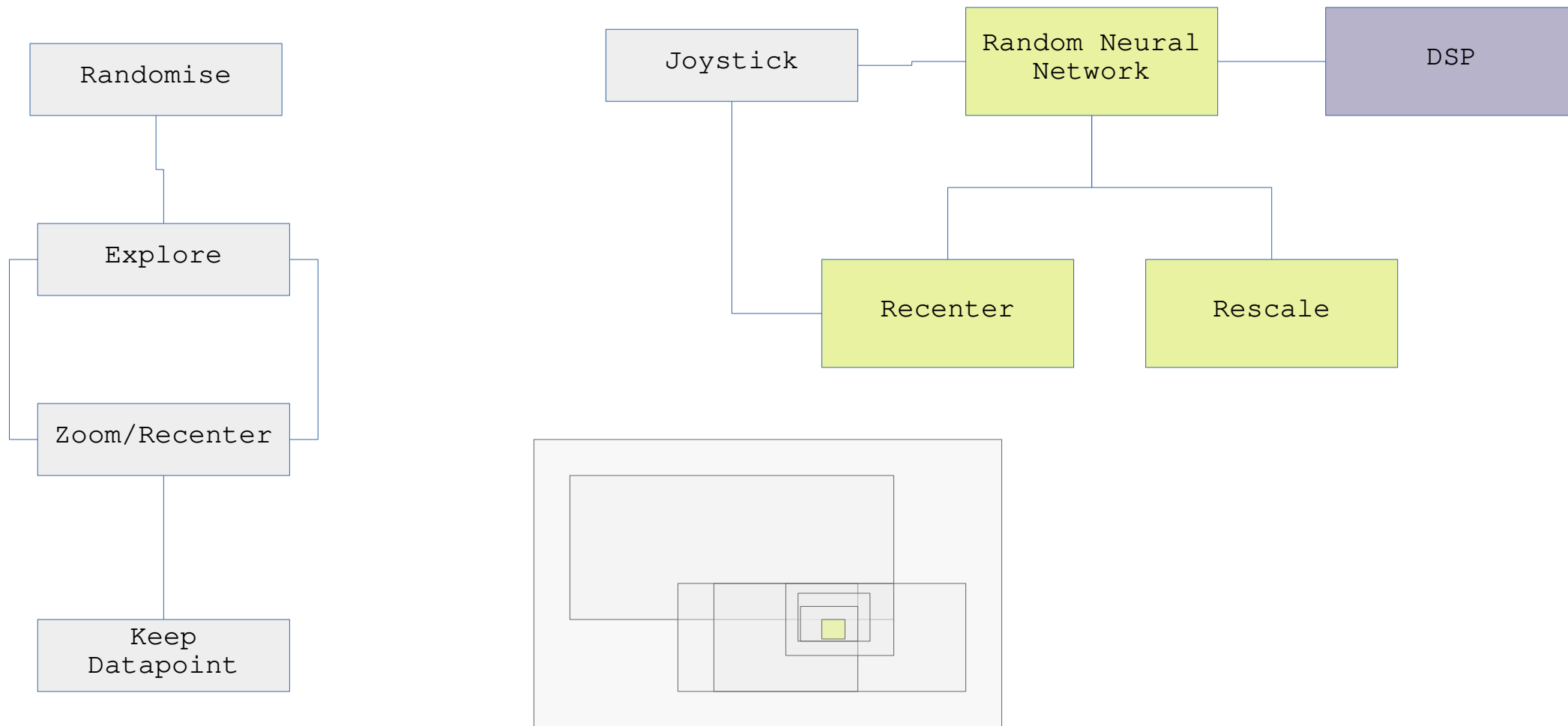


Question: how do we train these models in a *musical* way?

Data



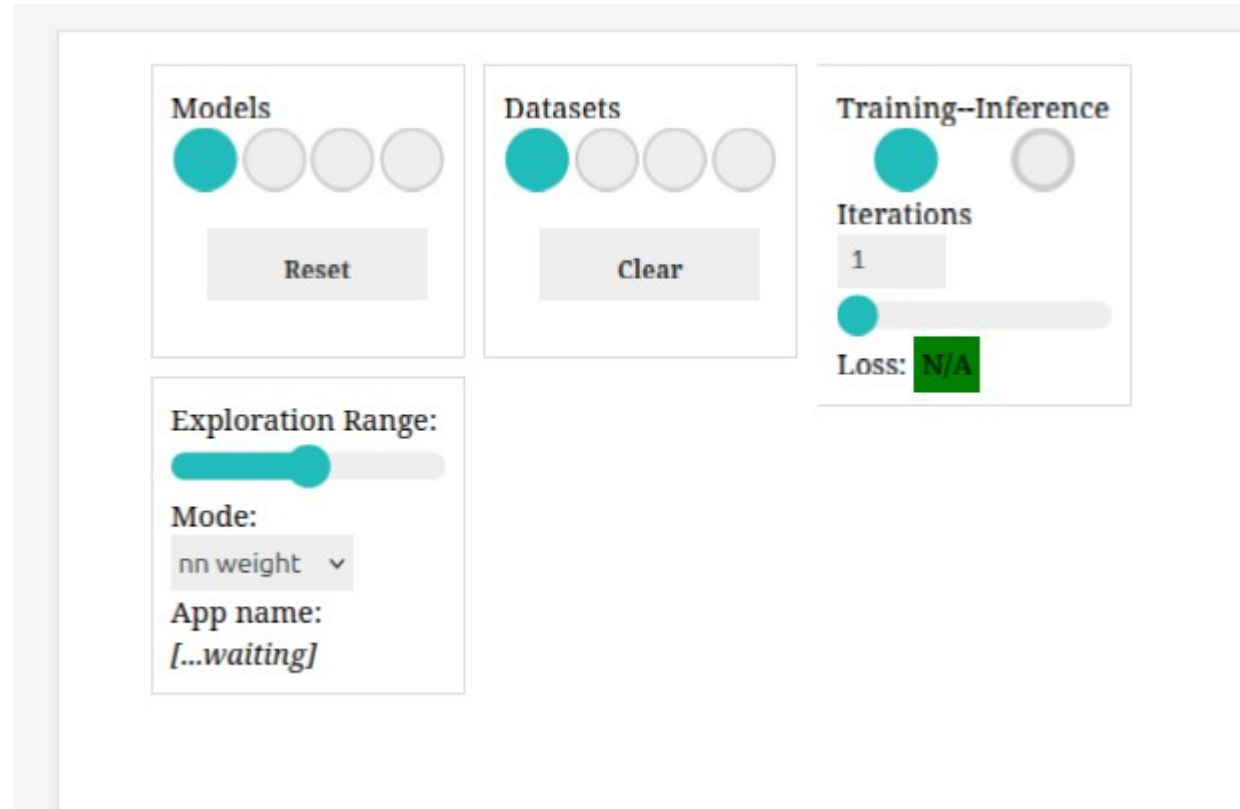
Sculpting Large Parameter Spaces: Neural Network Mapping Mode



Training Processes

Features:

- Switch between training models
- Store data in multiple datasets
- Train models with combinations of datasets
- Structured exploration of parameter spaces
- Influence training quality



themes

```
graph LR; themes --- shared_agency[shared agency]; themes --- levels_surprise[levels of surprise]; themes --- temp_control[temperature control]; themes --- encouraging_sparsity[encouraging sparsity]; themes --- scaffolding[scaffolding]; themes --- visual_feedback[visual feedback]; themes --- labelling[labelling]; themes --- aff_machine_learning[affordances of machine learning]; themes --- abstractions[abstractions?]; scaffolding --- visual_feedback; scaffolding --- labelling; levels_surprise --- temp_control; aff_machine_learning --- abstractions
```

shared agency

levels of surprise

temperature control

encouraging sparsity

scaffolding

visual feedback

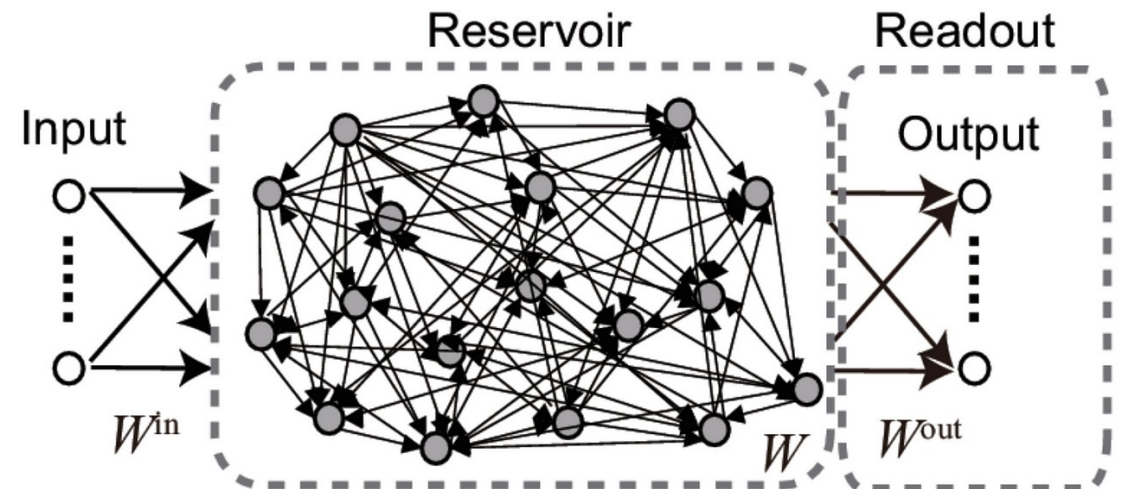
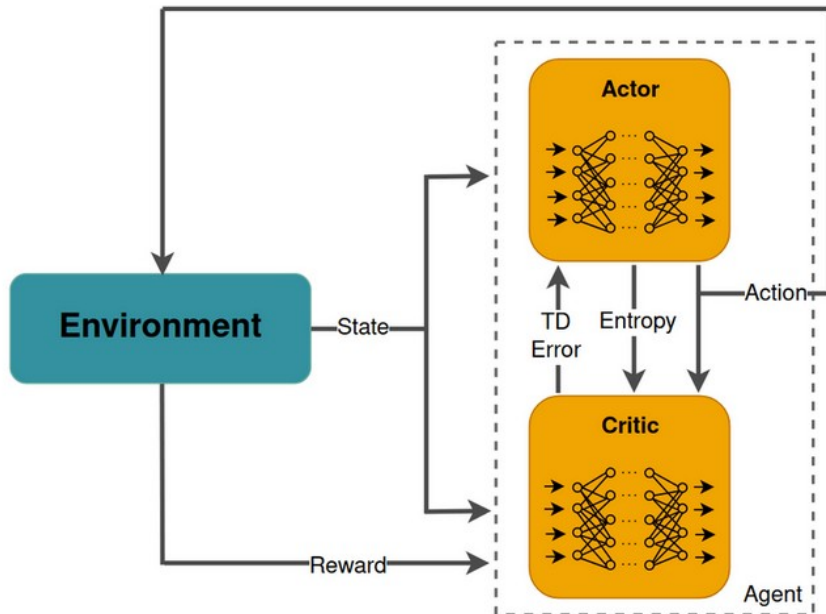
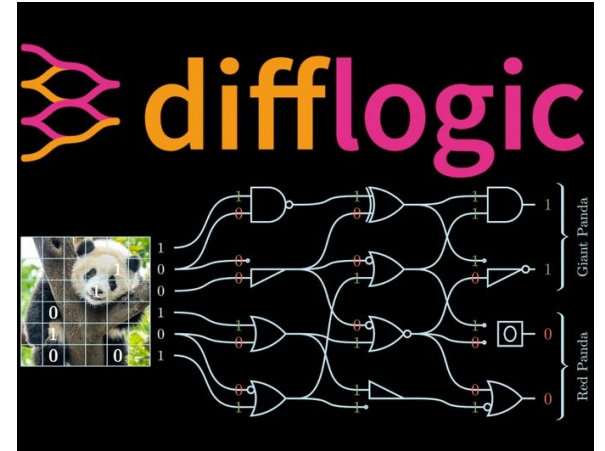
labelling

affordances of machine learning

abstractions?

Future Machine Learning Approaches

- Soft Actor-Critic Reinforcement Learning
- Reservoir Computing
- Differentiable Logic Networks



Future Instruments: ML Multi-fx in Production



Future Instruments: ML in the Xiasri (autobio design)



Questions?

<https://users.sussex.ac.uk/~ck84/meml/>

<https://www.emutelabinstruments.co.uk/>

<https://luuma.net/>

<http://www.emutelab.org/>



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