

NEUROPIXEL DATA TO 3D VISUALISATION

BME 4509 Group project
Team: Imperial
PixelMinds



IMPERIAL

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Scan for more info from our
website including our report



With heartfelt thanks to Professor Claudia Clopath and Dr. Albert Albessa Gonzalez for their support.

FUTURE POTENTIAL



- 1 Boost computational power for full-brain spike mapping.
- 2 Enable pipeline to process raw Neuropixels data.
- 3 Support additional modalities like calcium imaging and widefield data.
- 4 Integrate Allen 3D viewer with AR platforms for real-time visualization.
- 5 Develop AR tools for experiment planning, e.g. probe trajectory preview.



DISCUSSION

The results indicate successful mapping of neuronal spikes using spatial localisation using **Euler-based methods with the AMSGRad optimiser**.

It can be observed that recordings in **primary motor cortex (MOp)** and **secondary somatosensory cortex (SSp)** are largely non-selective visual and action neurons, active prior to movement onset.

Such result is expected as the regions integrate sensory and motor-related signals in movement planning and initiation.

MOTIVATION

Up to now, we still don't know:

- ? How neural networks transform perception and thought into behaviour.
- ? How neurons function and communicate, and how this helps us understand brain disorders.

+ High-resolution of neuron architecture

- Time consuming and labour intensive



Neuropixel probes capture detailed 2D data, but creating a 3D brain activity map still requires slow, manual tissue slicing, staining, and signal sorting.



Building on the framework of Steinmetz et al. and modern spike-localization methods we aimed to;

Automatically label neurons based on what stimuli they respond to.

Reconstruct 3D spike locations from the probe's 2D recordings

Map those neurons onto a standard 3D mouse-brain atlas

1

AIM



IDLE Shell 3.11.4

```
</> Session 1-Spike filtering & preprocessing
→ Step 1: Statistical selection
Compute stimulus-response p-value
If  $p < \text{threshold}$  → mark as responsive
```

→ Step 2: Half-Gaussian filtering
Remove low-firing neurons and noise

>>> Output 1: Significant neuronal clusters



```
</> Session 2-Neuronal classification
→ Step 3: Build Toeplitz matrix to map spikes to functions (visual, choice, action)
→ Step 4: Use ElasticNet to assign spikes to kernels
→ Step 5: Test for significant kernel types
```

>>> Output 2: Identified neuronal functions



```
</> Session 3-3D visualisation of spike
→ Step 6: Map filtered significant neuronal functions from Session 2 to the dataset
→ Step 7: Extract PTP values
→ Step 8: Train the algorithm to obtain the optimised 3D global coordinates by Euler-based method equipped with AMSGrad (Adam variant) regularizer.
```

>>> Output 3: 3D spike visualisation in Brainrender at CCF



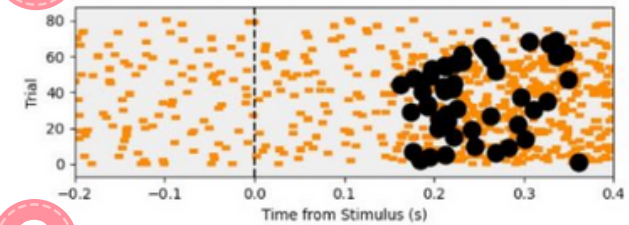
Python was selected as the primary software tool for its flexibility and the richness of its open-source library ecosystem.

2

SPIKING DATA
PIPLINE

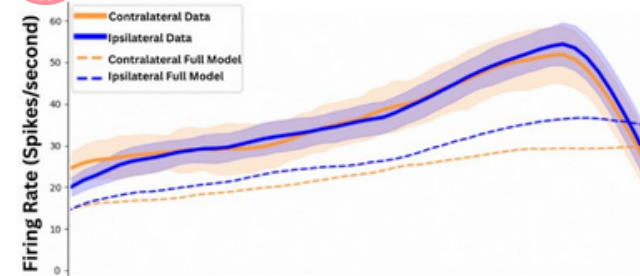
1

Contralateral

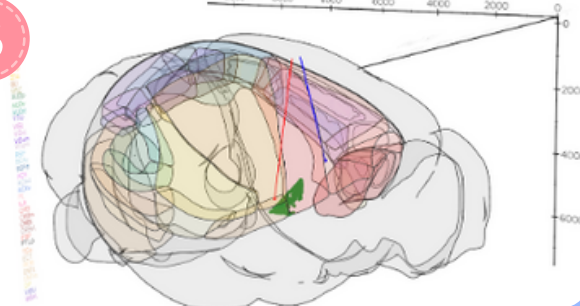


2

Model With Action Predictors



3



RESULTS

1

The spike activity increases around 0.2s, which suggests the neurons are involved in action execution.

2

If the variance is greater than 2%, the neuron is classified as action-selective neurons.

3

The visualisation of action and vision neurons at MOP and SSP

3

VISUALIZATION
OUTPUTS