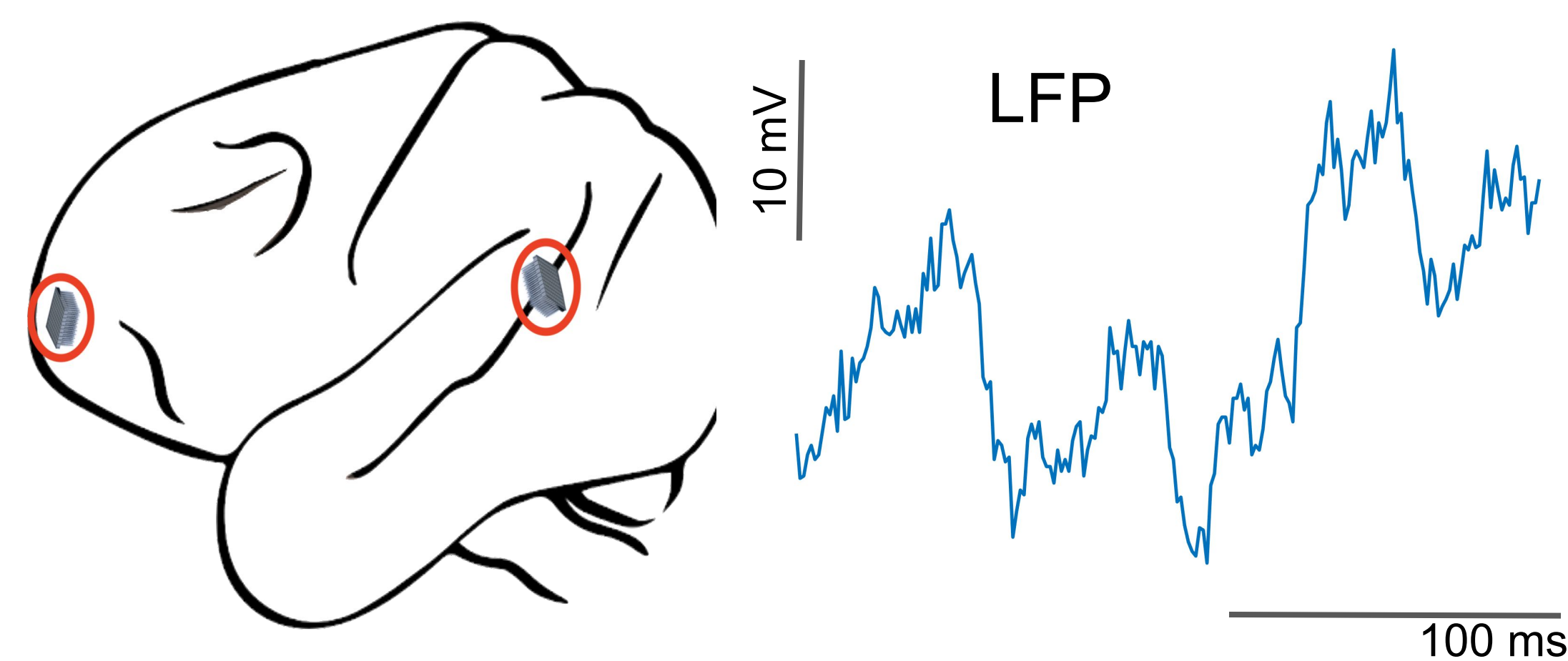


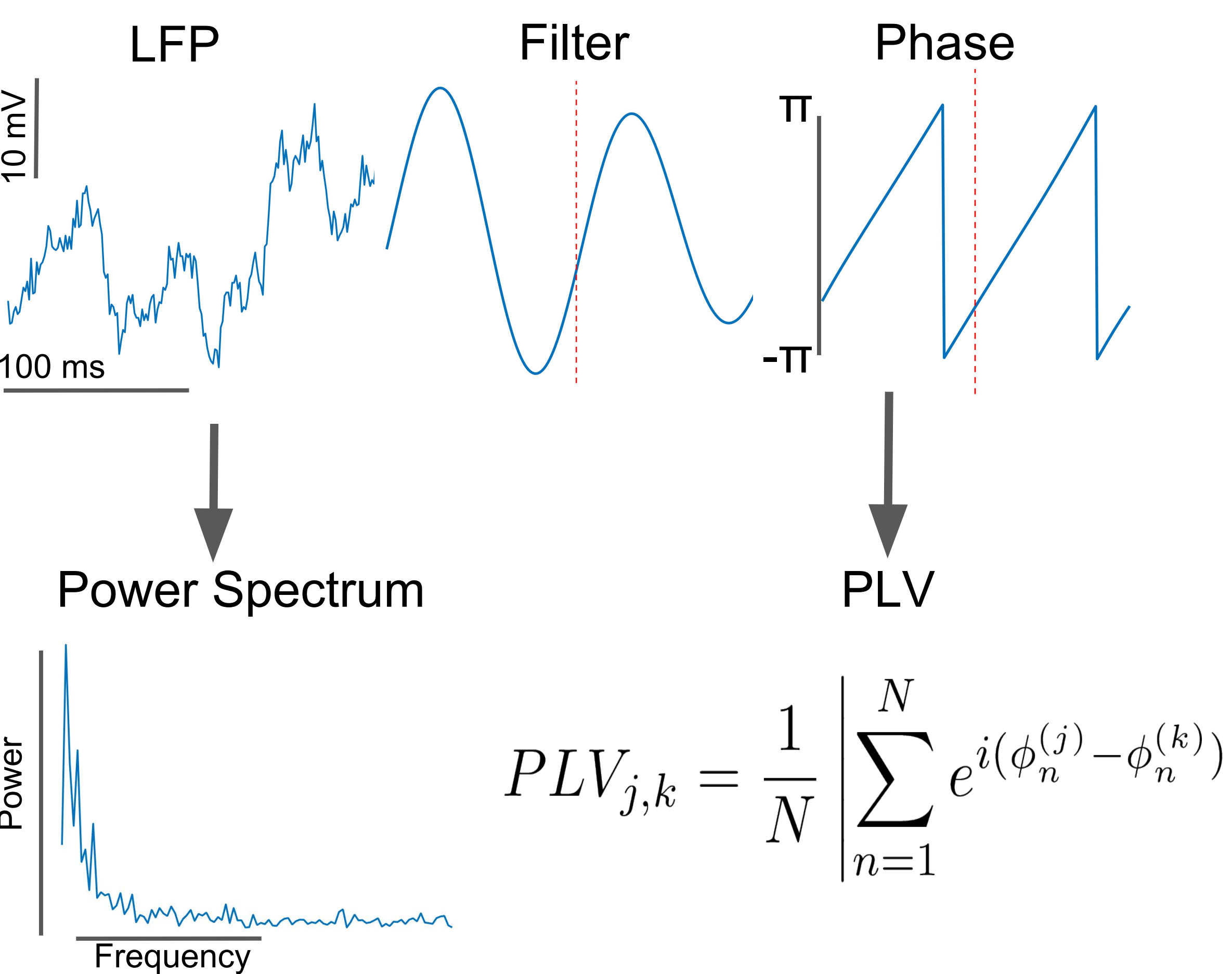
## Introduction

- Phase coupling measures how synchronous neural oscillations are. It can be an indicator of how well two regions are communicating.
- Phase coupling, signal strength, and distance are all related to neural communication, however, the relationship is not understood.
- Goal: Explore the relationships between phase coupling, signal strength, and distance.**
- Data: Local field potentials (LFP) are recorded simultaneously within V4 and prefrontal cortex (PFC) using UTAH arrays over 1742 trials.



## Methods

- To measure phase coupling, we used phase locking value (PLV) which lies between 0 and 1.
- LFPs are bandpass filtered to 15Hz and 10Hz. Phases were then found using the Hilbert transformation and at a fixed time point.

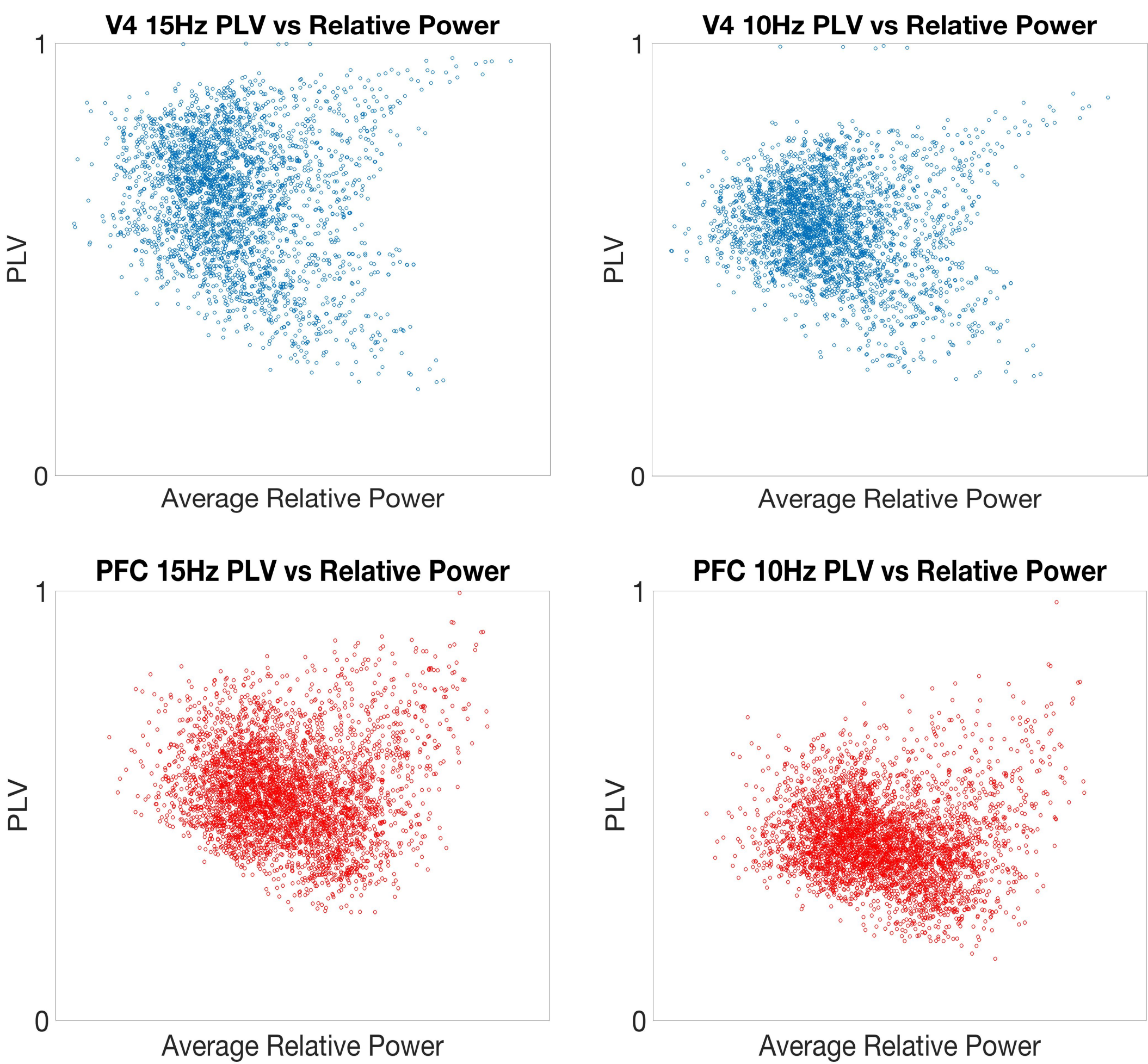


- Frequency analysis involved the use of the Fast Fourier Transform, an algorithm that implements the Discrete Fourier Transform to find the power of each frequency in the LFP.
- “Bad” channels were eliminated if there were no spiking activity present in the channels during the trial periods.

## Results

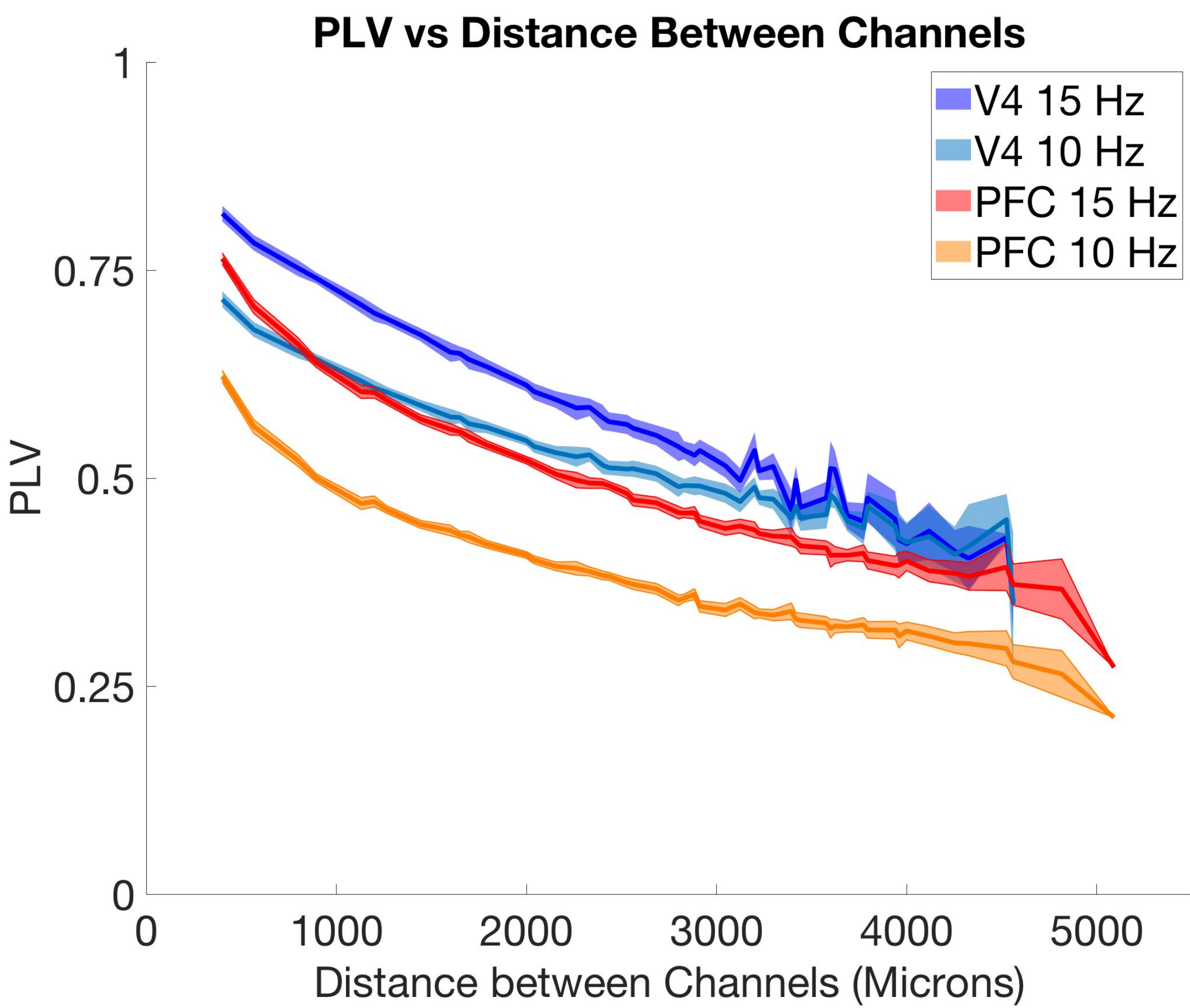
**PLV decreases as distance increases in both brain regions and both frequencies.**

- On average, PLV of the 10Hz signal is lower than its 15Hz counterparts.
- Plot depicts the mean PLV value and its standard error at each distance.
- Regions are plotted in a similar color scheme V4 (blue) and PFC (red/orange).



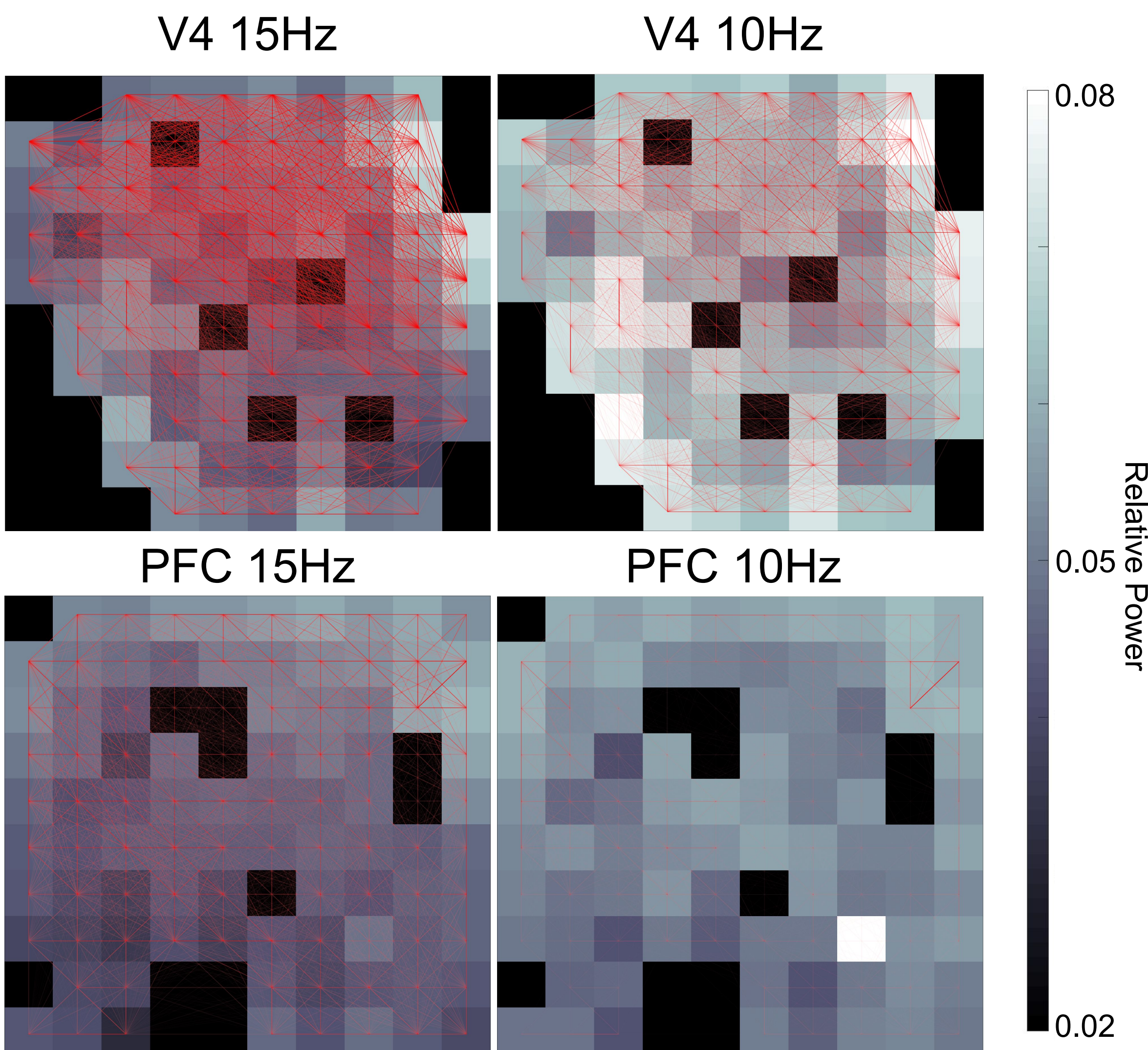
**15Hz signal exhibits, on average, higher phase locking than the 10Hz signal in both regions.**

- Red lines that are more opaque indicate higher PLV values and thus stronger phase locking
- Brighter pixels indicate stronger relative power and signal fidelity

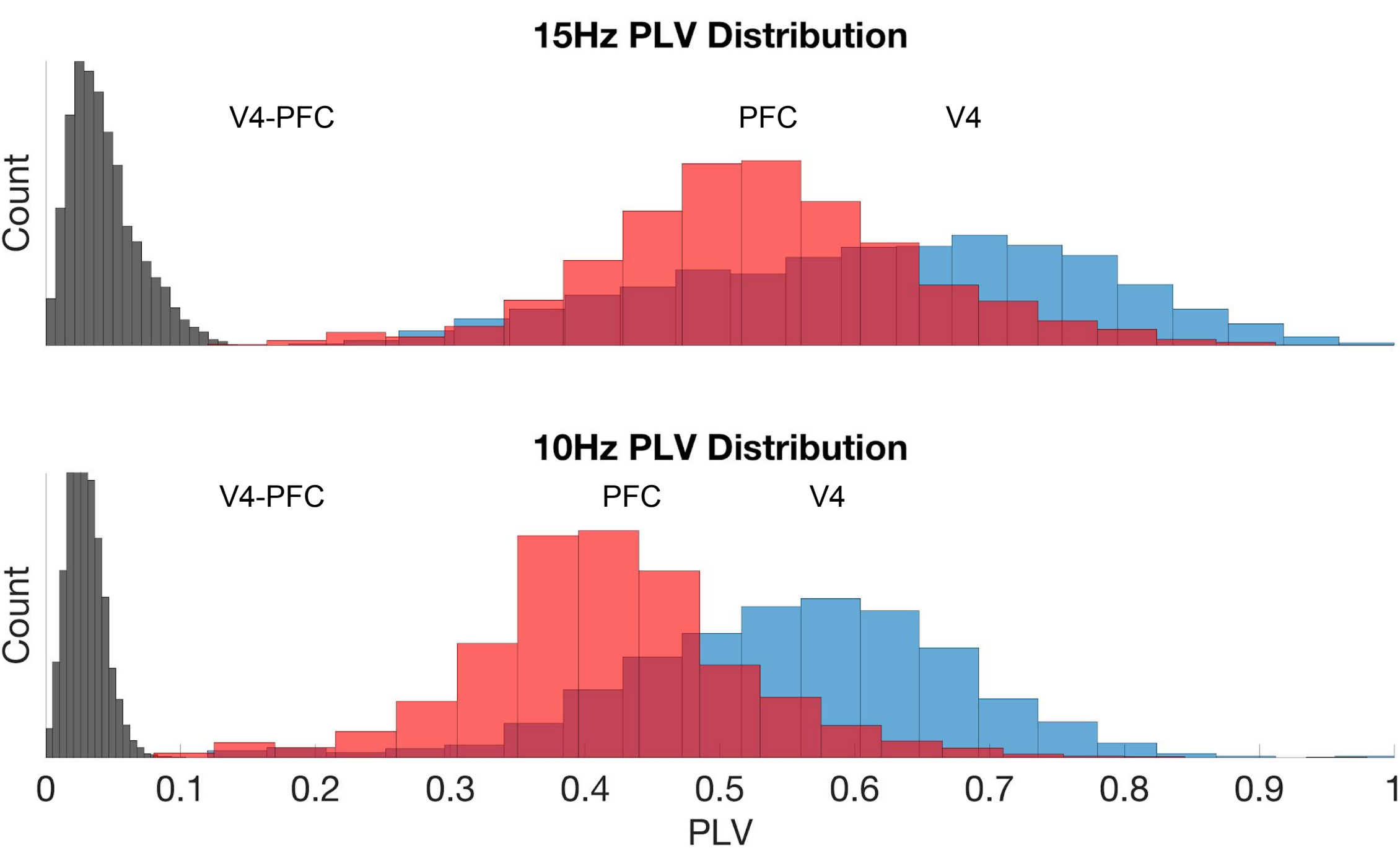


**No direct relationship between PLV values and relative power of the target signal strength.**

- Each signal in each region had a differently shaped scatter plot for PLV vs relative power of the signal
- No large changes across experimental conditions
- Phase locking value is plotted versus the average relative power of its channels



**Phase coupling between V4 and PFC was low.**



## Discussion

PLV decreases as distance increases, has no direct relationship with relative power of a signal, and is higher in the 15Hz signals. V4 and PFC lack phase coupling in the two recorded regions.

- Verification of the existence of the signals was done by comparing a filtered signal to a smoothing at the target frequencies
- Poor communication between V4 and PFC because their signals were poorly coupled
- V4 is more phase coupled to itself, perhaps it must conduct computations on the signals to find various features of the stimulus

## Future Plans

- Explore the relationship between coherence and PLV. PLV is only based on phase values while coherence also includes the power of the signal
- Explore how well gamma waves phase lock between these regions especially during attention experiments
- Check more brain regions to see if the trends still appear

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### Author Affiliations

- Department of Mathematics & Department of Molecular and Cell Biology, UC Berkeley
- Department of Neuroscience, University of Texas at Austin
- Machine Learning Department, Carnegie Mellon University
- Department of Statistics and Data Science, Carnegie Mellon University
- Center for the Neural Basis of Cognition, Carnegie Mellon University and University of Pittsburgh