

# Introduction to Neuromorphic Engineering

## Assignment 1

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### Usage

#### Requirements

- NumPy
- Matplotlib

#### How to Use

- Python Assignment.py -h -n -m -p
- N: Length of Stimulus
- M: Length of window
- P: probability of spike

### Implementation

- **Stimulus:** Length N stimulus has been generated by assuming gaussian distribution
- **Spikes:** Binomial Distribution has been assumed to generate spikes in an interval of length n with probability p
- **Filter:**
  - M length windows has been sliced from stimulus at the position of spikes
  - Average of all such windows will give us filter
- **Smooth Stimulus:** Stimulus has been convoluted with filter
- **Histogram**
  - Histogram of filtered stimulus is generated with (n/10) bins ' $P(Sf)$ '.
  - For generation of  $P(Sf|r)$  histogram values of filtered stimulus has been chosen at the position of spikes.
  - Probability of response given Stimulus then calculated as follow

$$P(r|Sf) = P(Sf|r) \cdot P(r) / P(Sf)$$

$P(r)$  is the rate of spike which is taken equal to p

## Results

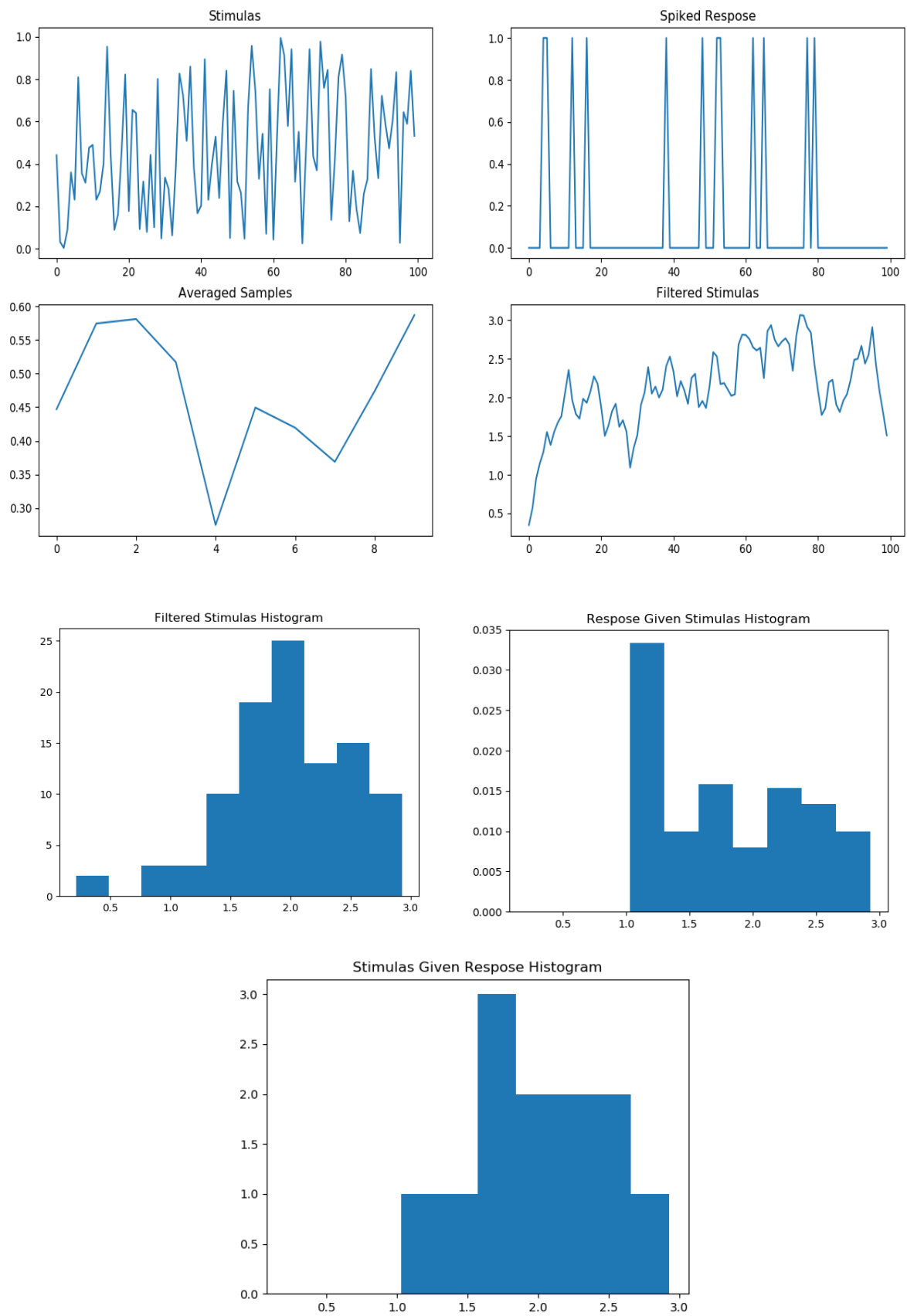


Figure 1  $n = 100$ ,  $m = 10$ ,  $p = 0.1$

For  $n = 100$ ,  $m=20$ ,  $p=0.1$

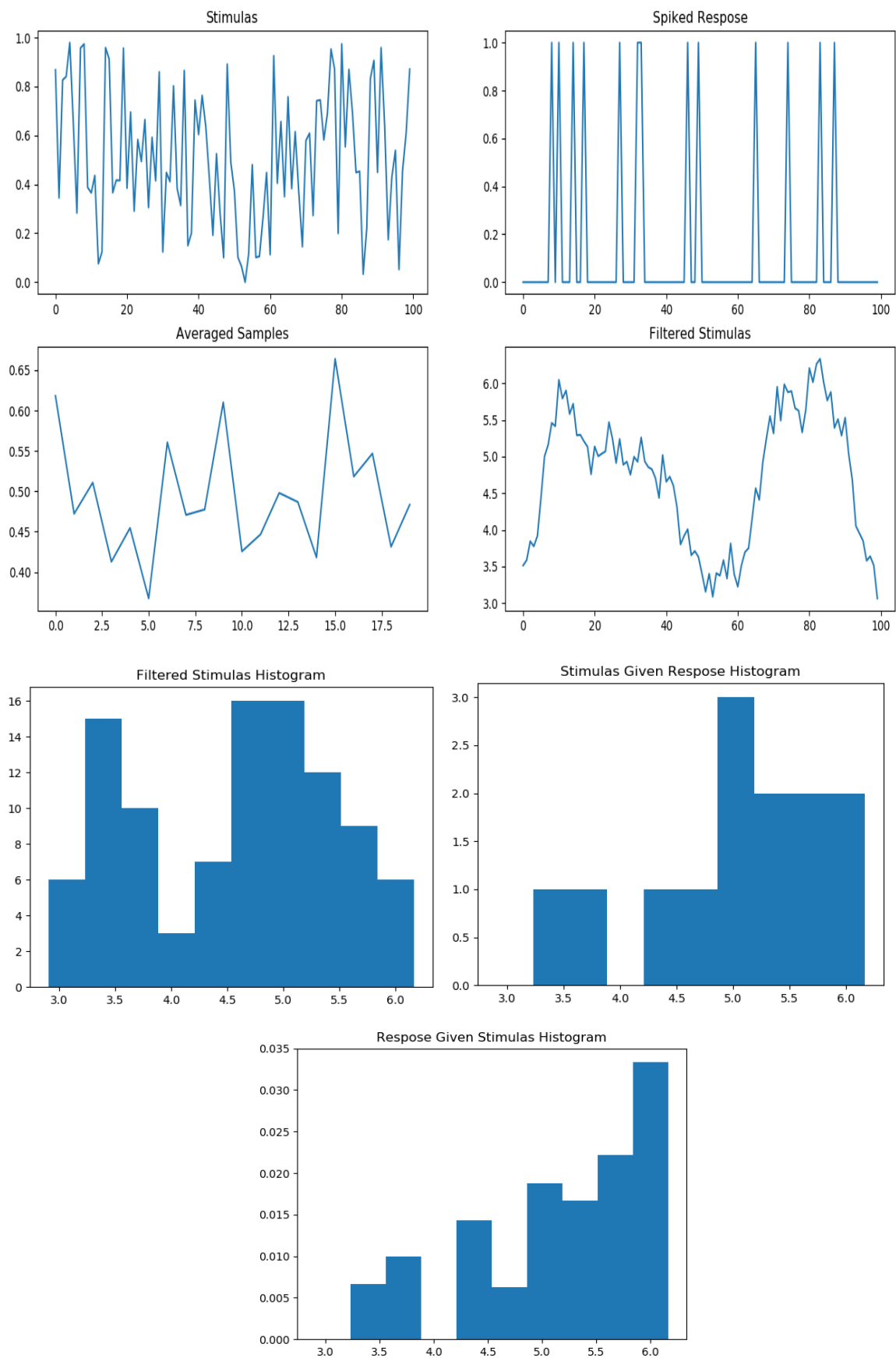


Figure 2  $n = 100$ ,  $m = 20$ ,  $p = 0.1$

For  $n = 1000$ ,  $m=10$ ,  $p=0.1$

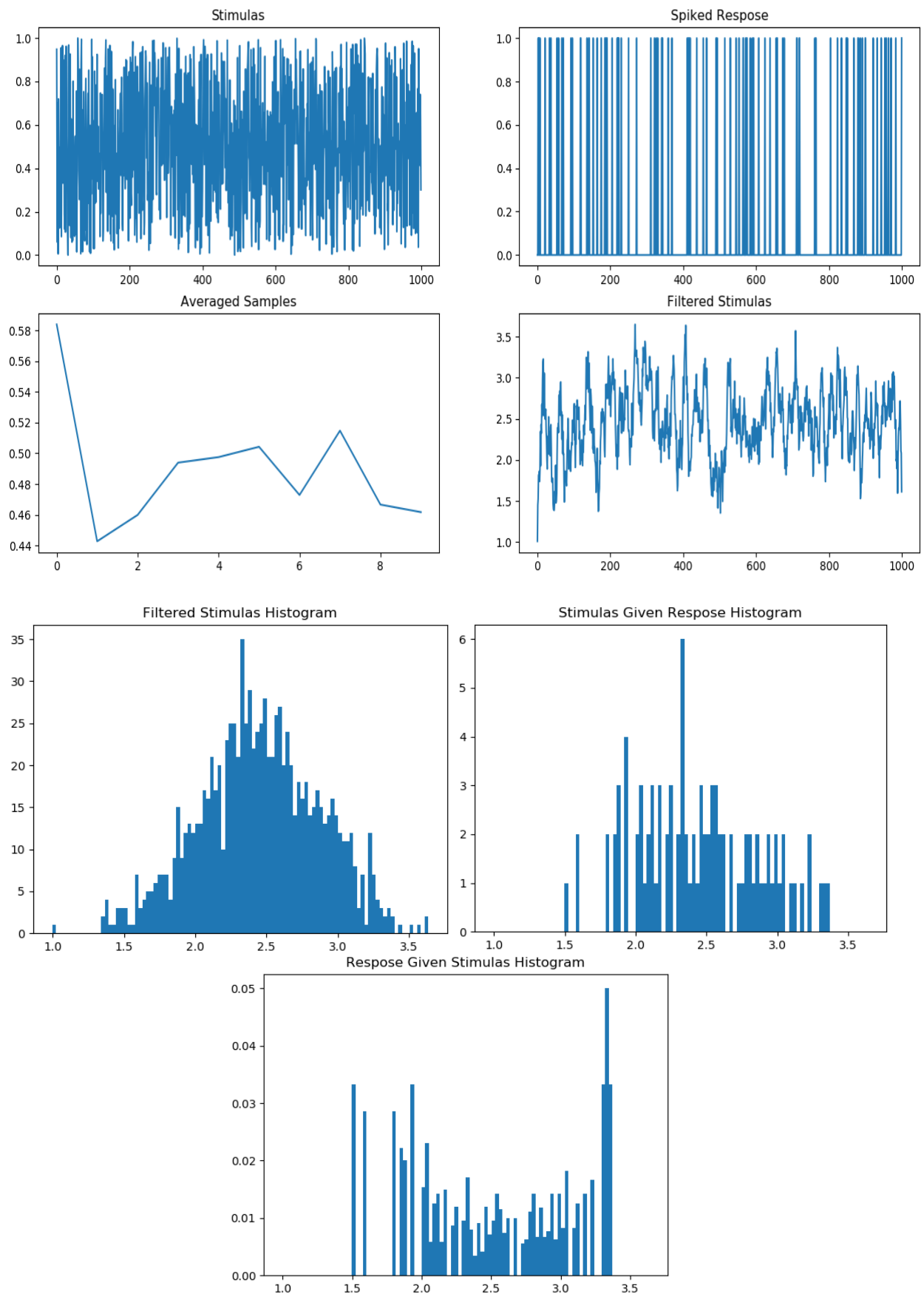


Figure 3  $n = 1000$ ,  $m = 10$ ,  $p = 0.1$

For  $n = 1000$ ,  $m=50$ ,  $p=0.1$



Figure 4  $n = 1000$ ,  $m = 50$ ,  $p = 0.1$

For  $n = 1000$ ,  $m=50$ ,  $p=0.5$

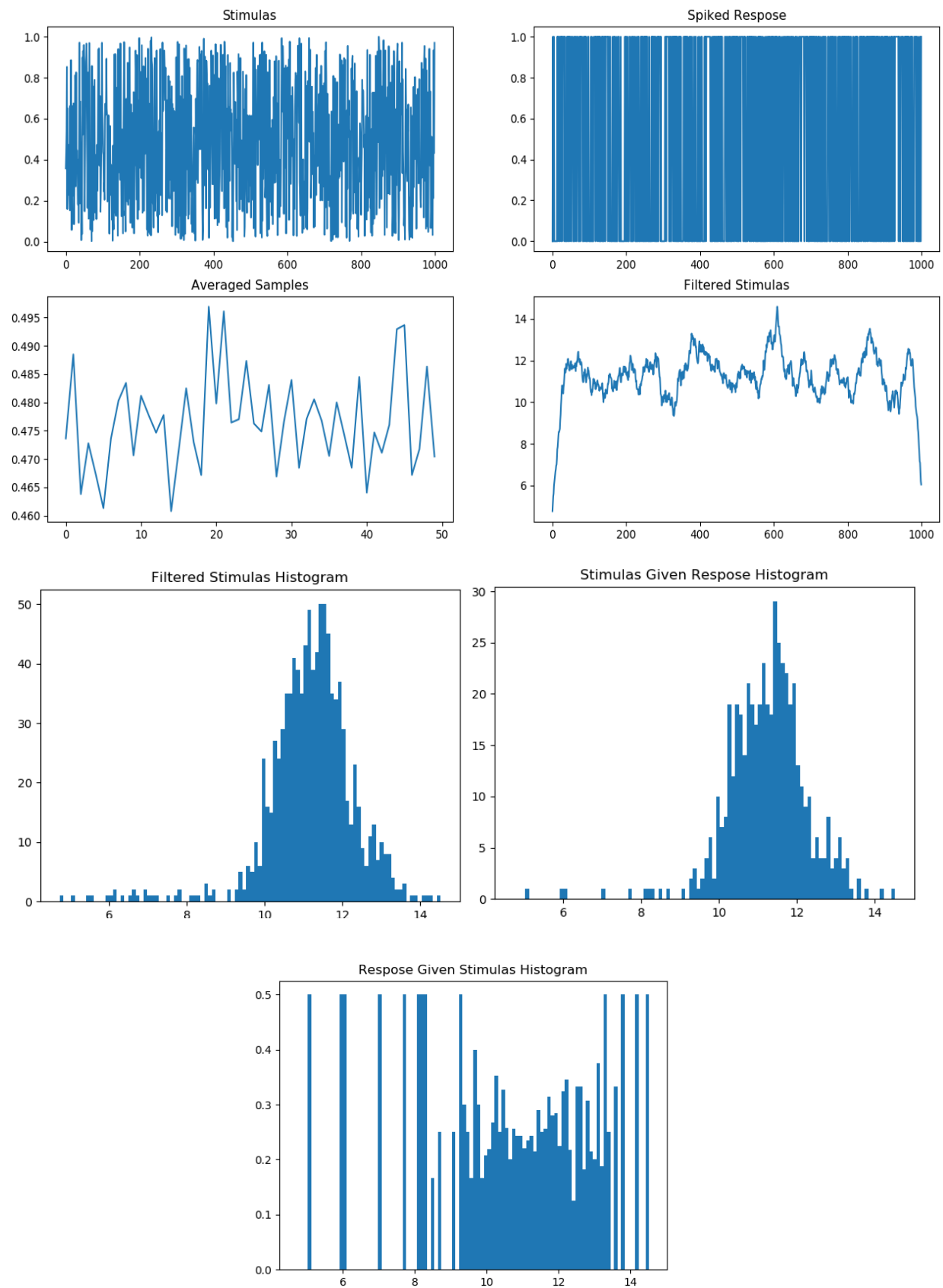


Figure 5  $n = 1000$ ,  $m = 50$ ,  $p = 0.5$

## Discussion:

- PCA can be used for multiple filters instead of using average of all samples, we can select suitable component, so different components can act like different filters