#### **SYDE 556/750**

#### Simulating Neurobiological Systems Lecture 9: Analysing Representations

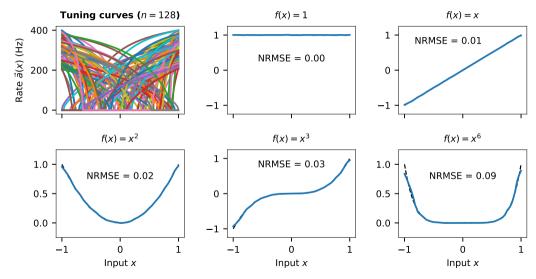
Andreas Stöckel

March 5, 2020

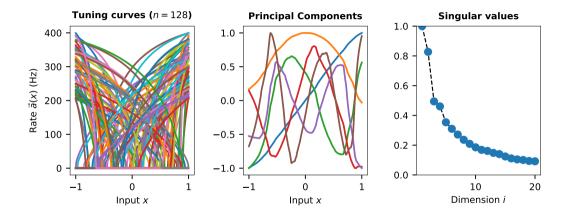




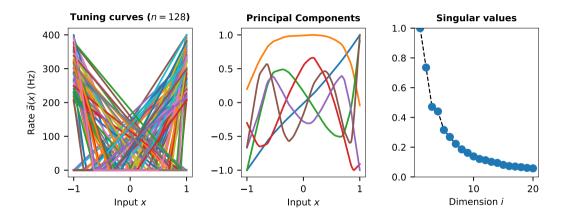
## Decoding Polynomials



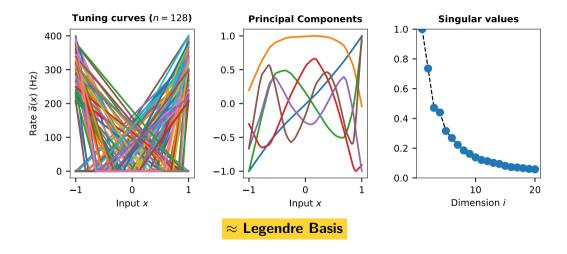
### LIF Tuning Curve Principal Components



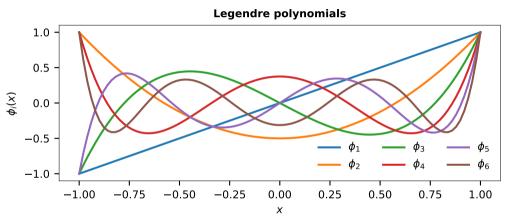
## ReLU Tuning Curve Principal Components



### ReLU Tuning Curve Principal Components

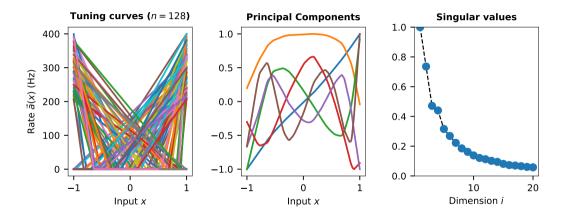


## Reminder: Legendre Polynomials

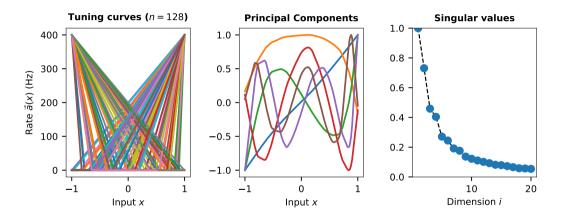


$$\varphi_i(x) = \frac{1}{2^i} \sum_{k=0}^n {i \choose k}^2 (x-1)^{i-k} (x+1)^k$$

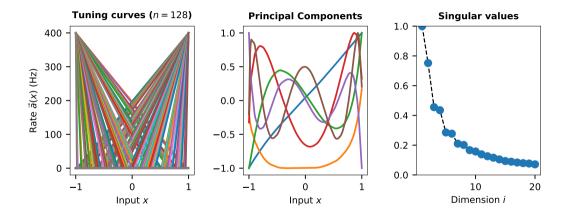
# Modifying the Basis – Same Maximum Rate (I)



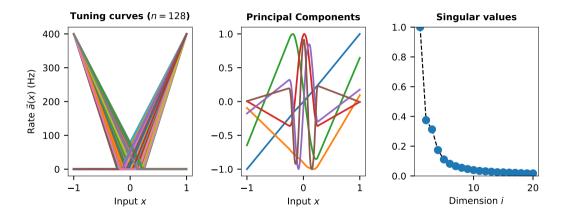
# Modifying the Basis – Same Maximum Rate (I)



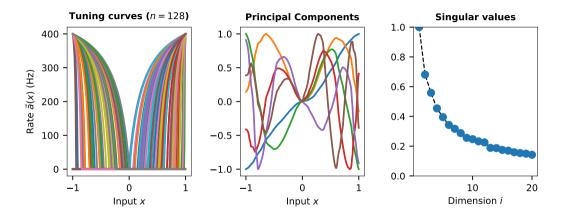
## Modifying the Basis – Equidistant *x*-Intercepts (II)



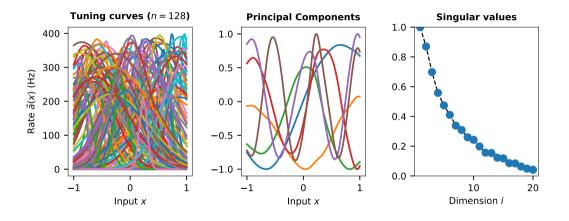
# Modifying the Basis – Limited *x*-Intercepts (III)



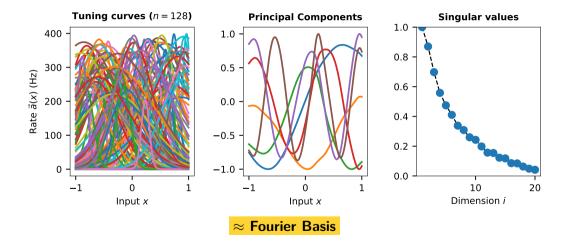
## Modifying the Basis – Symmetric Tuning Curves (IV)



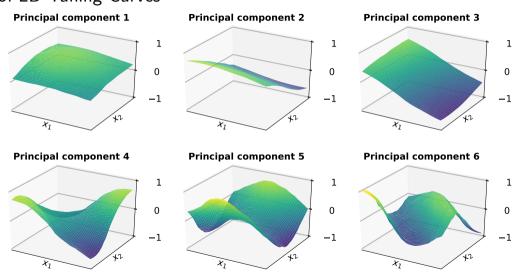
## Gaussian Tuning Curve Principal Components



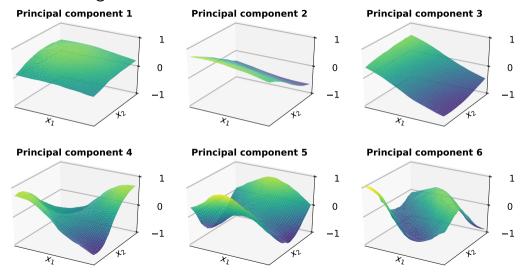
## Gaussian Tuning Curve Principal Components



## PCA of 2D Tuning Curves



#### PCA of 2D Tuning Curves



**Combination of 2D Polynomials** 

#### Conclusions

- ► Can use **PCA** to find the basis functions underlying neural representations
- ► Singular values inversely proportional to noise
- Basis function shape depends on
  - x-intercept distributions
  - Neuron response curve G[J]
- Finding optimal tuning curves for representations
  - ⇒ Full network optimization (must use gradient descent)

## Image sources

#### Title slide

Maurice Denis: Homage to Cézanne, 1900

From Wikimedia.