

Predicting Knot Invariants using Machine Learning

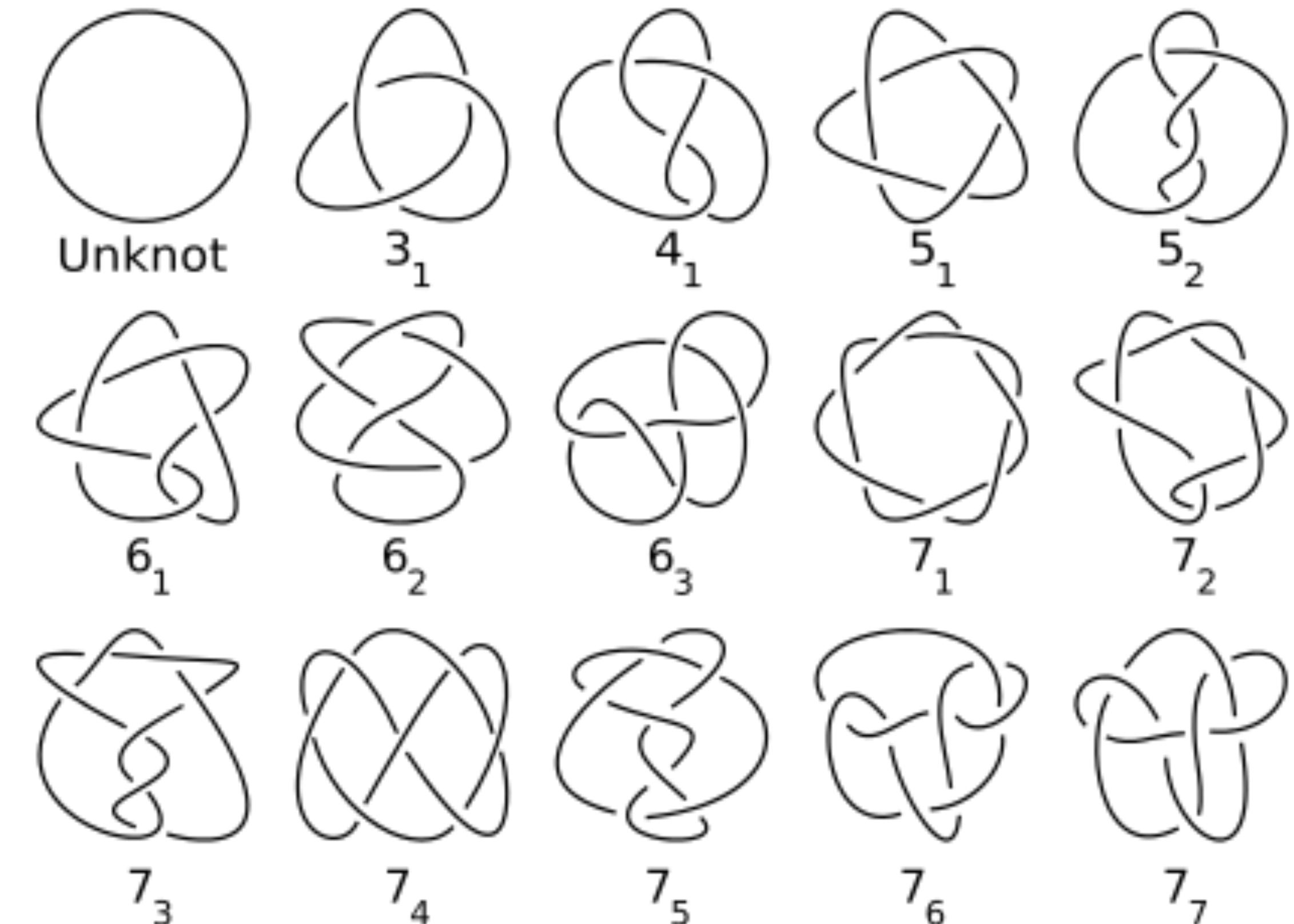
Ben Jacobsohn

Mentored by Professor Boyu Zhang

Knot Theory

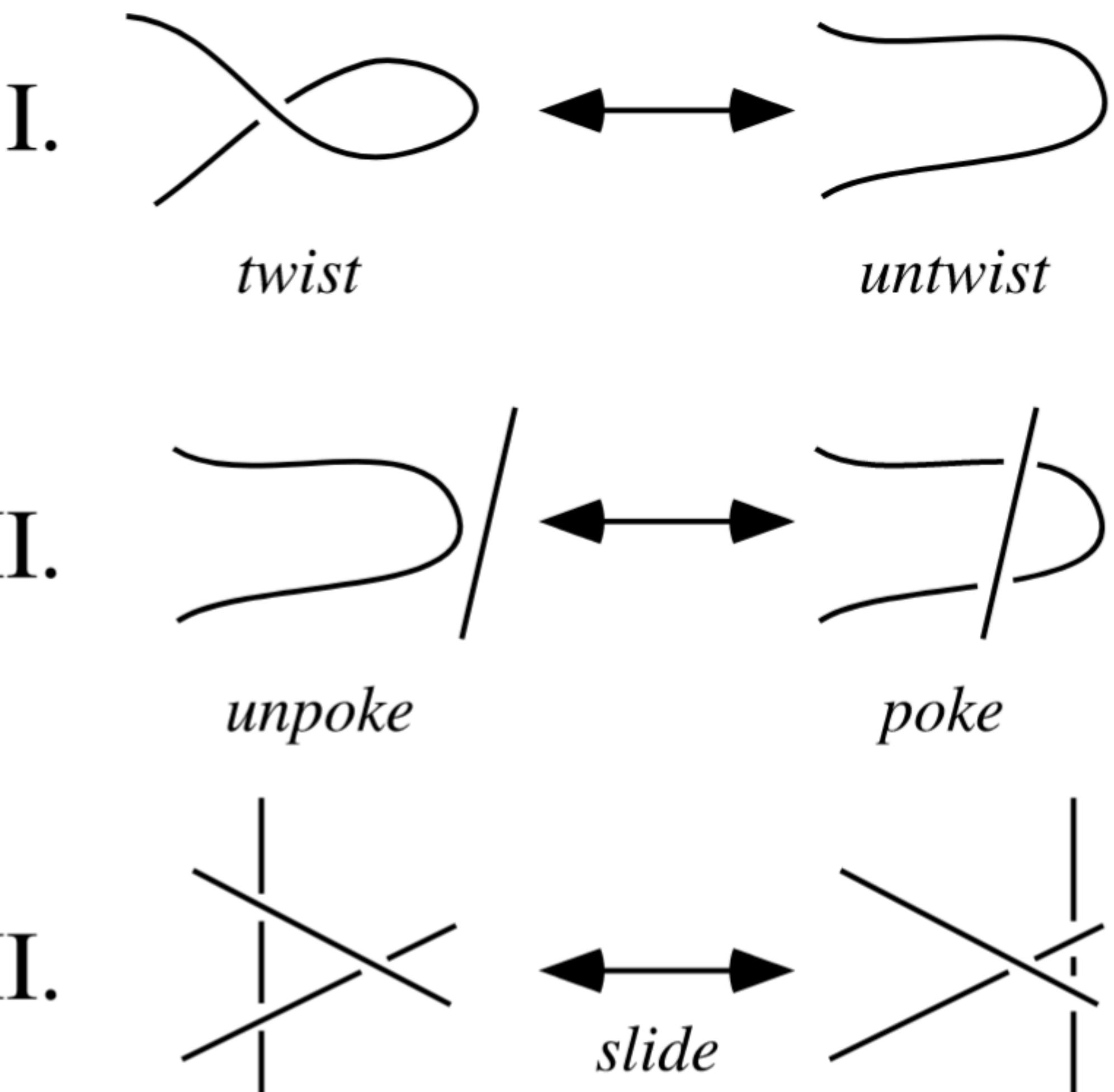
Mathematical Knots

- Smoothly embedded circle in \mathbb{R}^3
- No self-intersections
- Can smoothly map space between equivalent knots
- Equivalence is difficult to determine



Redemeister Moves

- Moves that preserve equivalence
- Can describe all knot deformations
- Can be done computationally



Knot Invariants

- Properties that don't change under Reidemeister moves
- Not always unique
- Jones Polynomial
 - Recursive combinatorial definition
- Hyperbolic Volume
 - Knot complement with hyperbolic structure

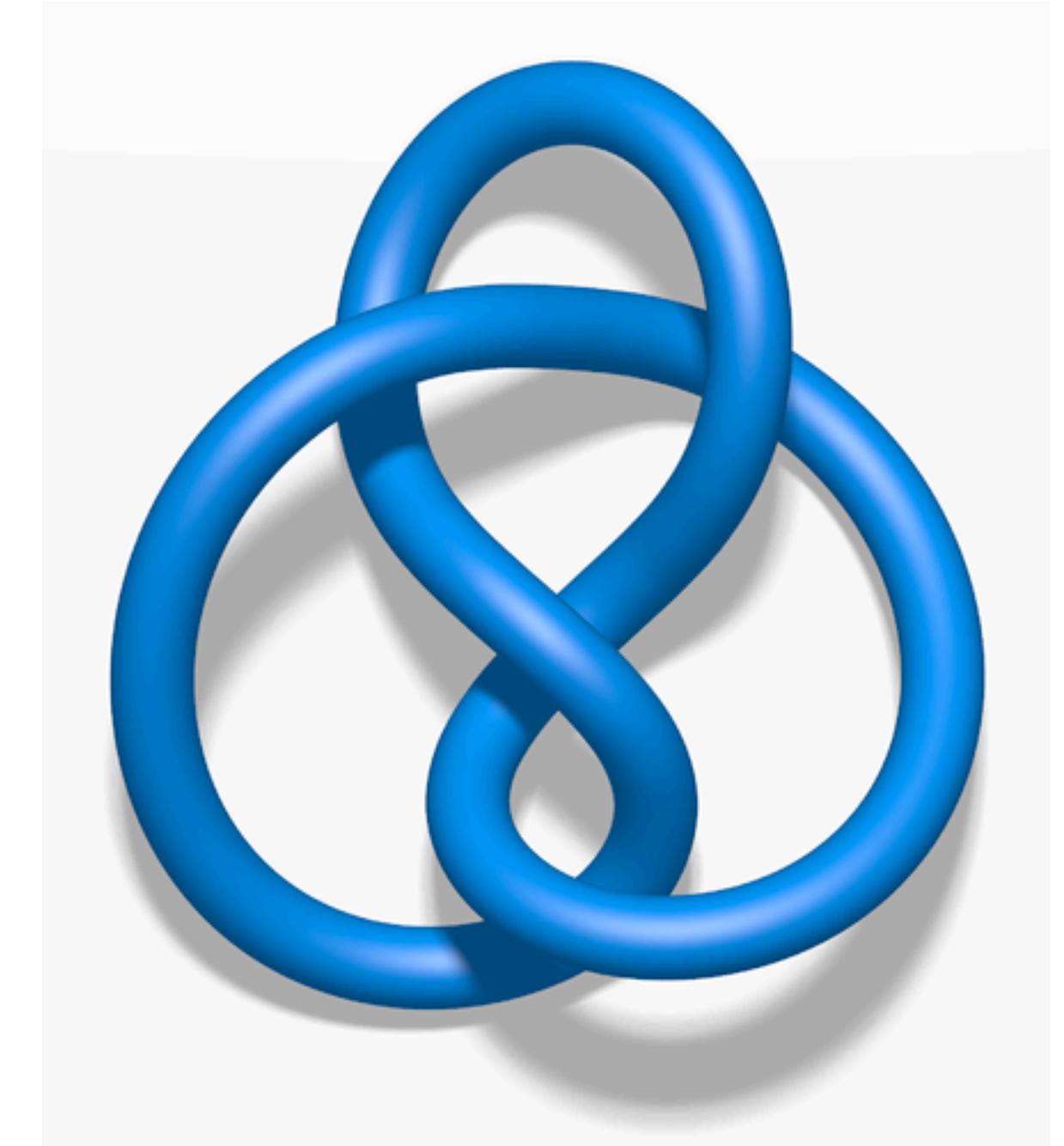
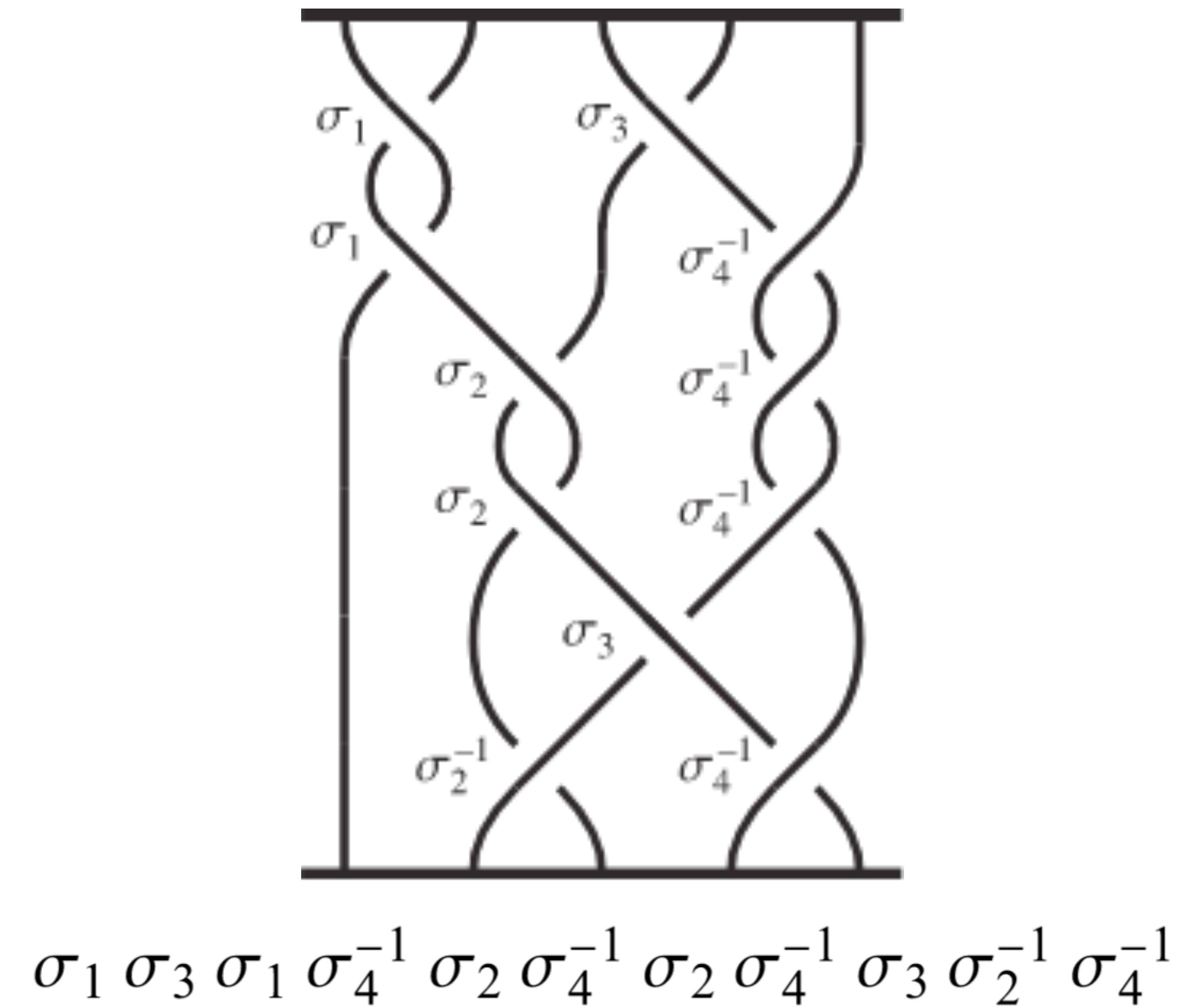
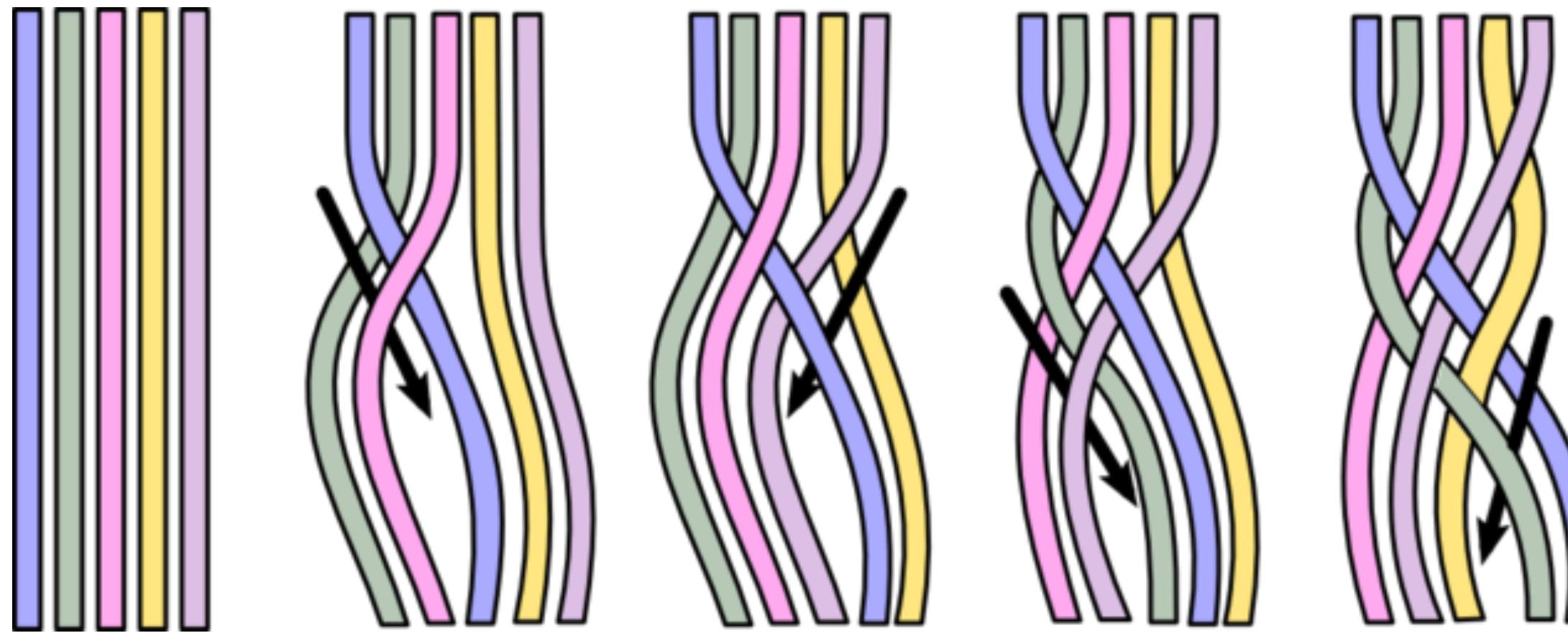


Figure-Eight Knot
Hyperbolic Volume:
2.0298832
Jones Polynomial:
 $V(t) = t^2 - t + 1 - t^{-1} + t^{-2}$

Braid Words

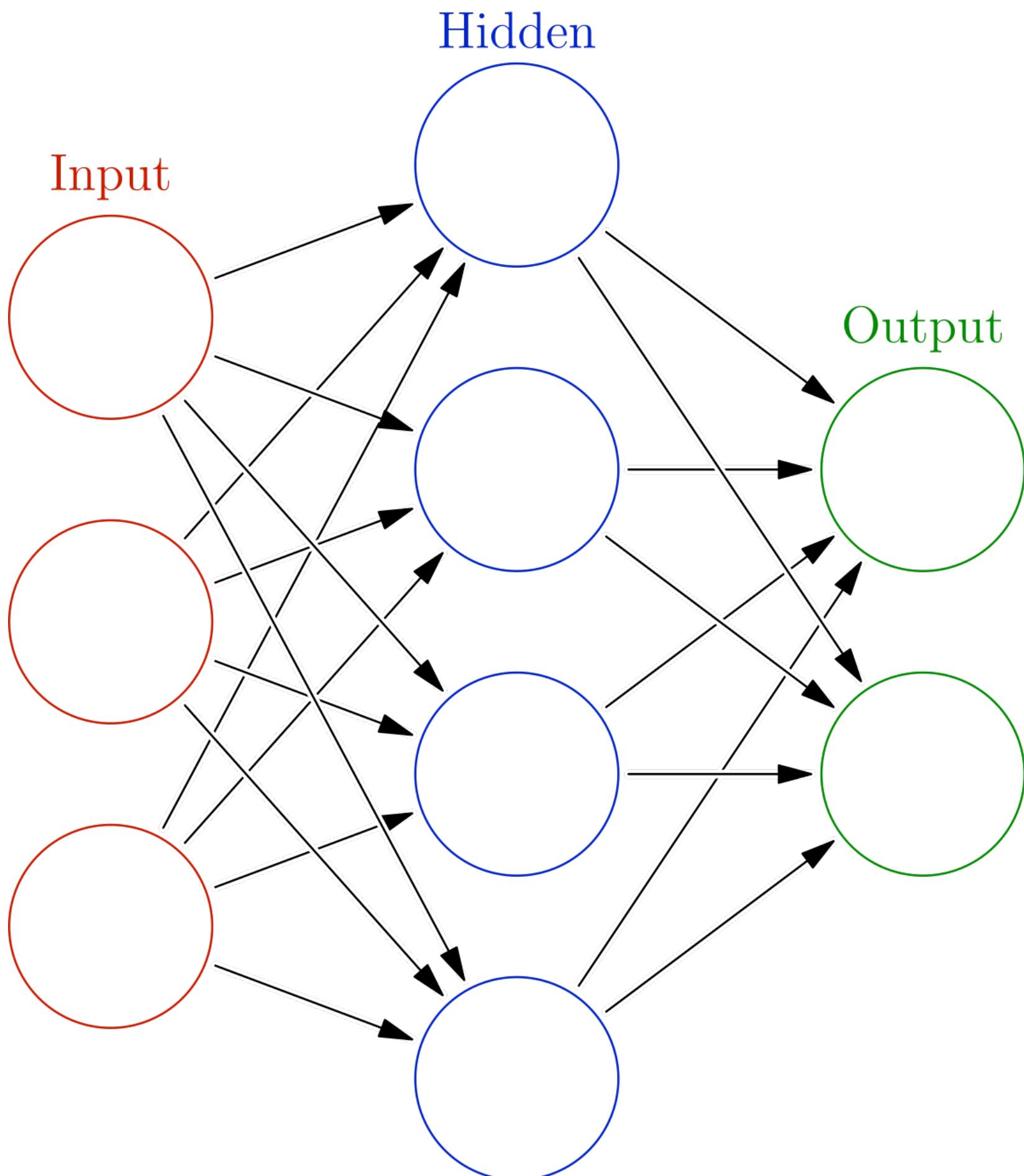
- Can represent every knot
- List of crossings between strands
- Connect ends of strands to form knot



Machine Learning

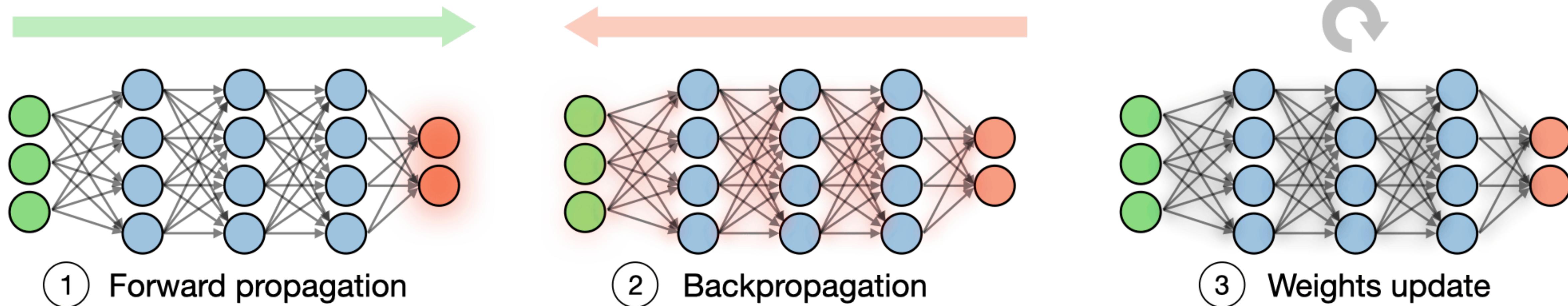
Neural Networks

- General function approximation
- Layers of neurons connected based on weights
- Weights updated based on network performance



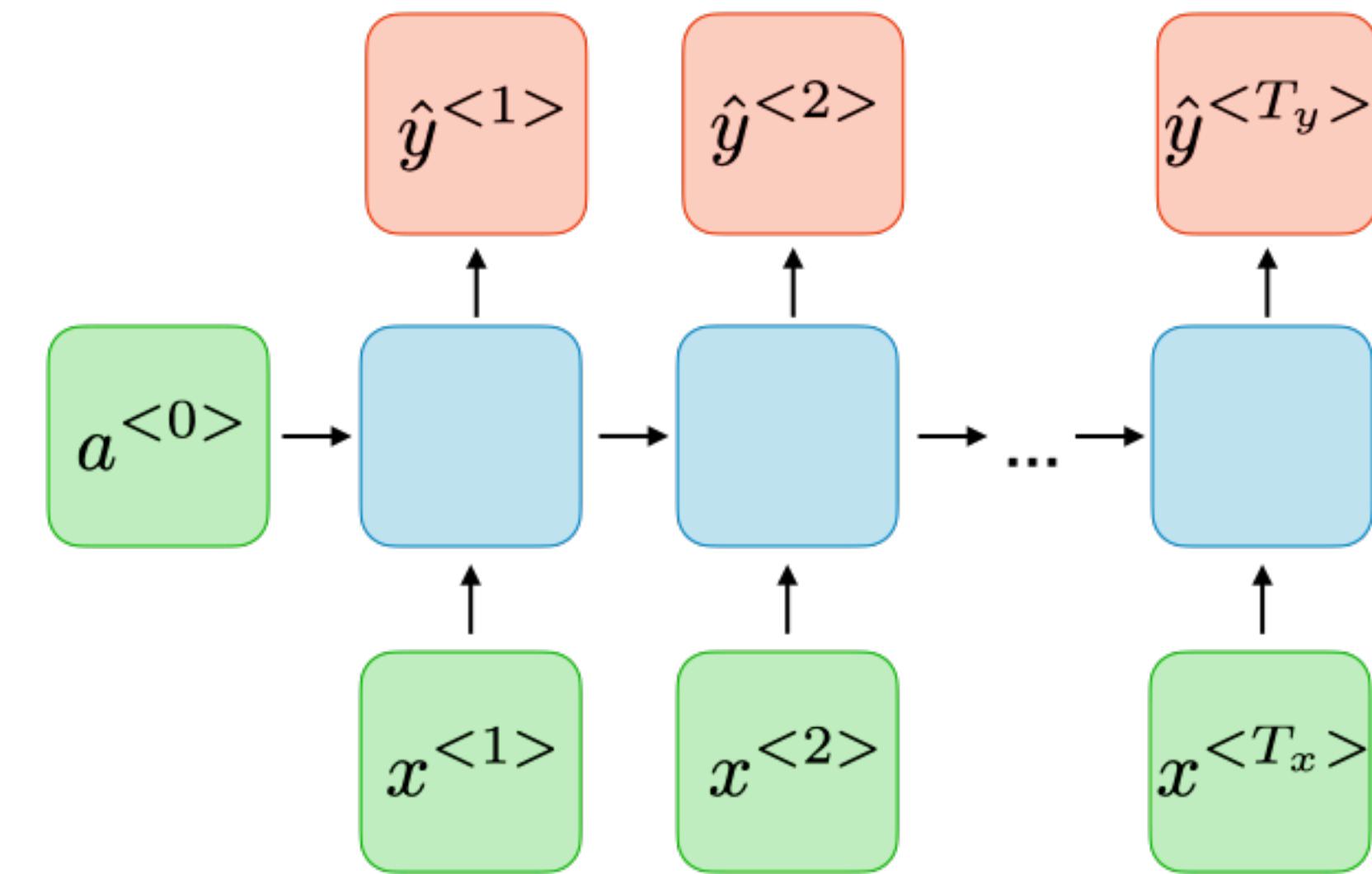
Training a Neural Network

- Basic form: supervised learning
- Start with labelled dataset
- Apply network to make prediction
- Backpropagate errors

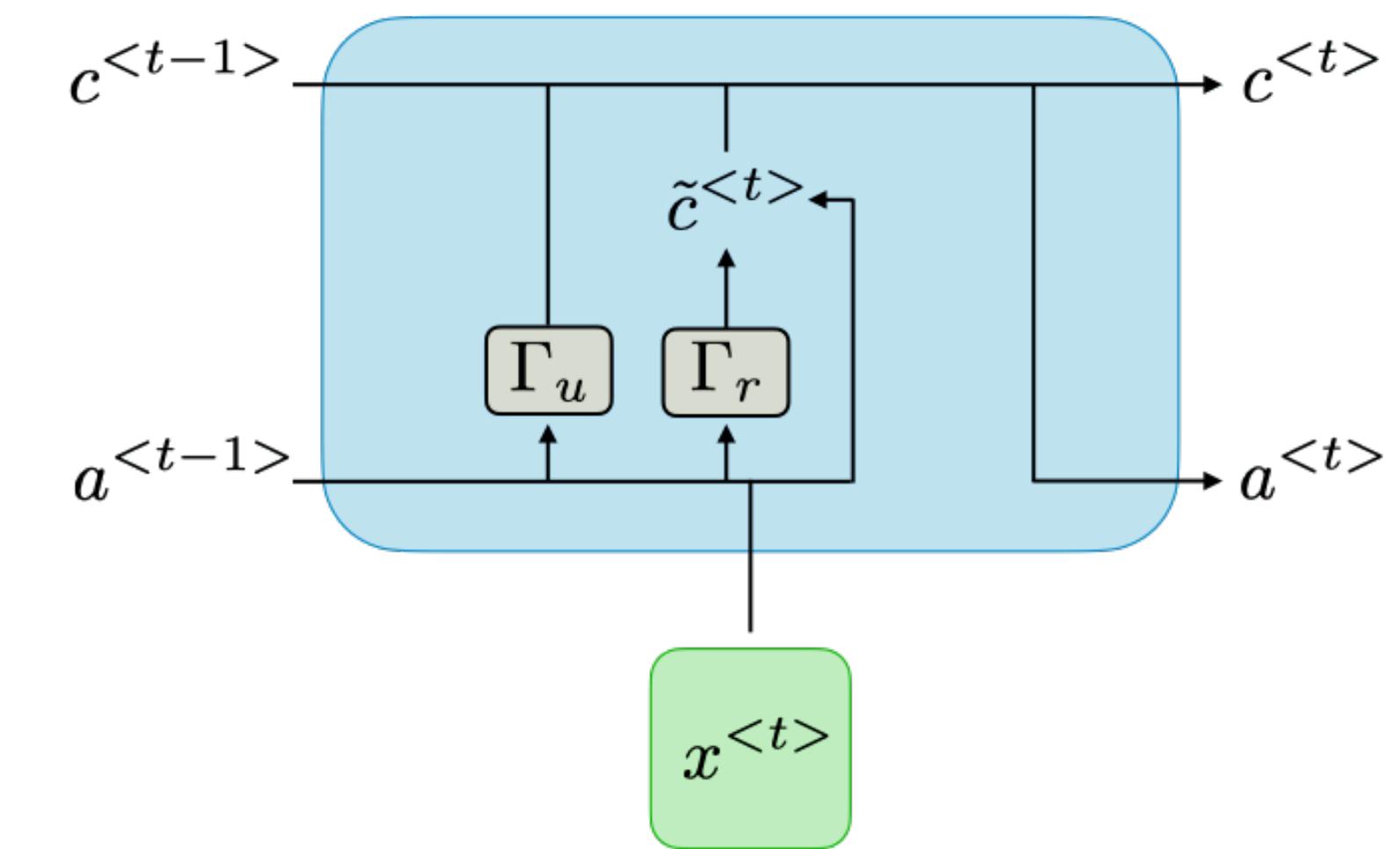


Recurrent Networks

- Used for sequences of data
- Activations depend on previous steps
- Simple recurrent net: activation from current and previous time step
- Can handle long-term dependencies
 - Gated Recurrent Unit (GRU)
 - Long Short-Term Memory (LSTM)



“Unrolled” recurrent network



Gated Recurrent Unit

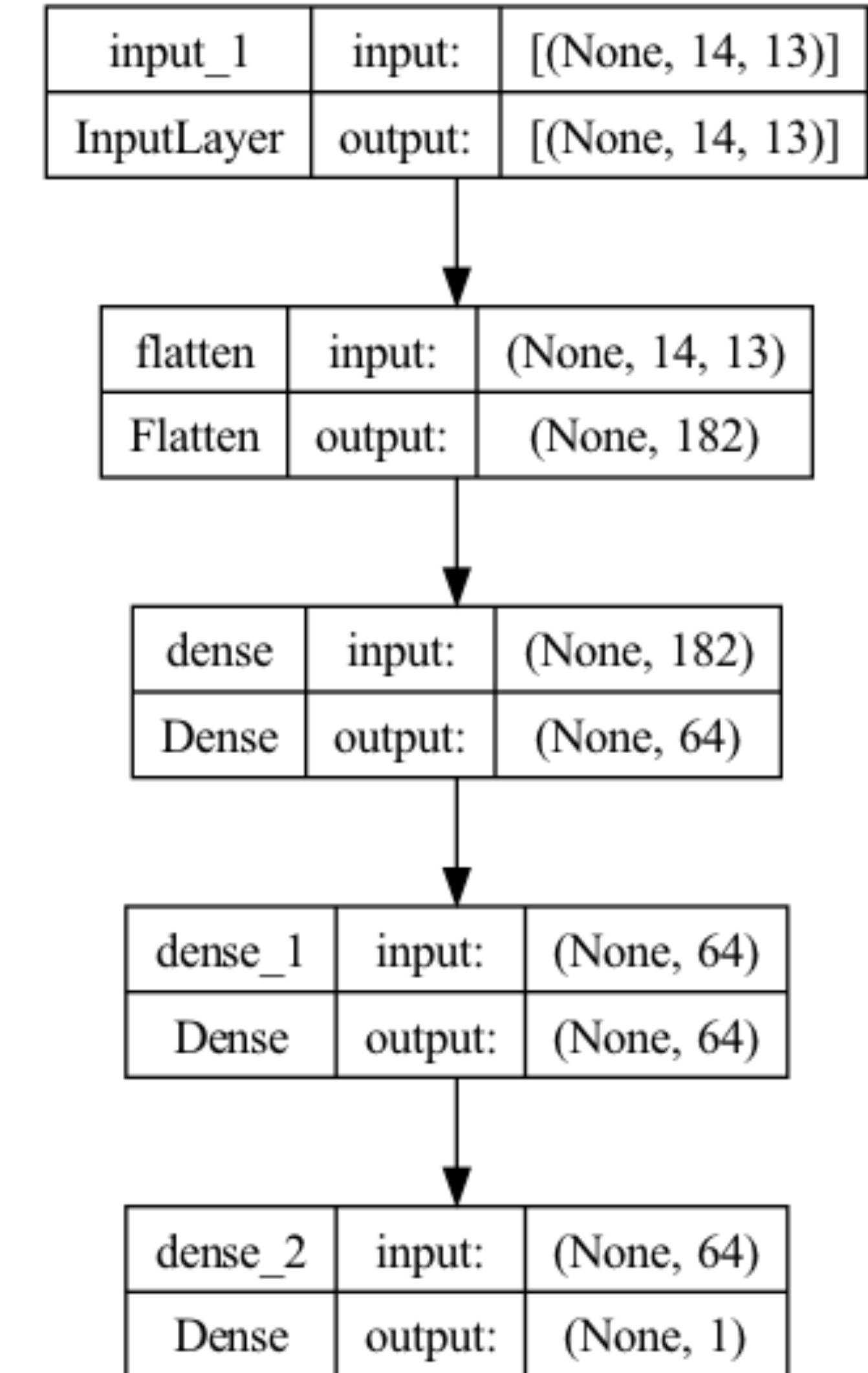
Applications to Knot Theory

- *A Neural Network Approach to Predicting and Computing Knot Invariants*
 - Predict knot quasipositivity from braid words
 - Also used other classification data
- *Deep Learning the Hyperbolic Volume of a Knot*
 - Predicted hyperbolic volume from Jones polynomial
 - Based on volume conjecture

Experiments

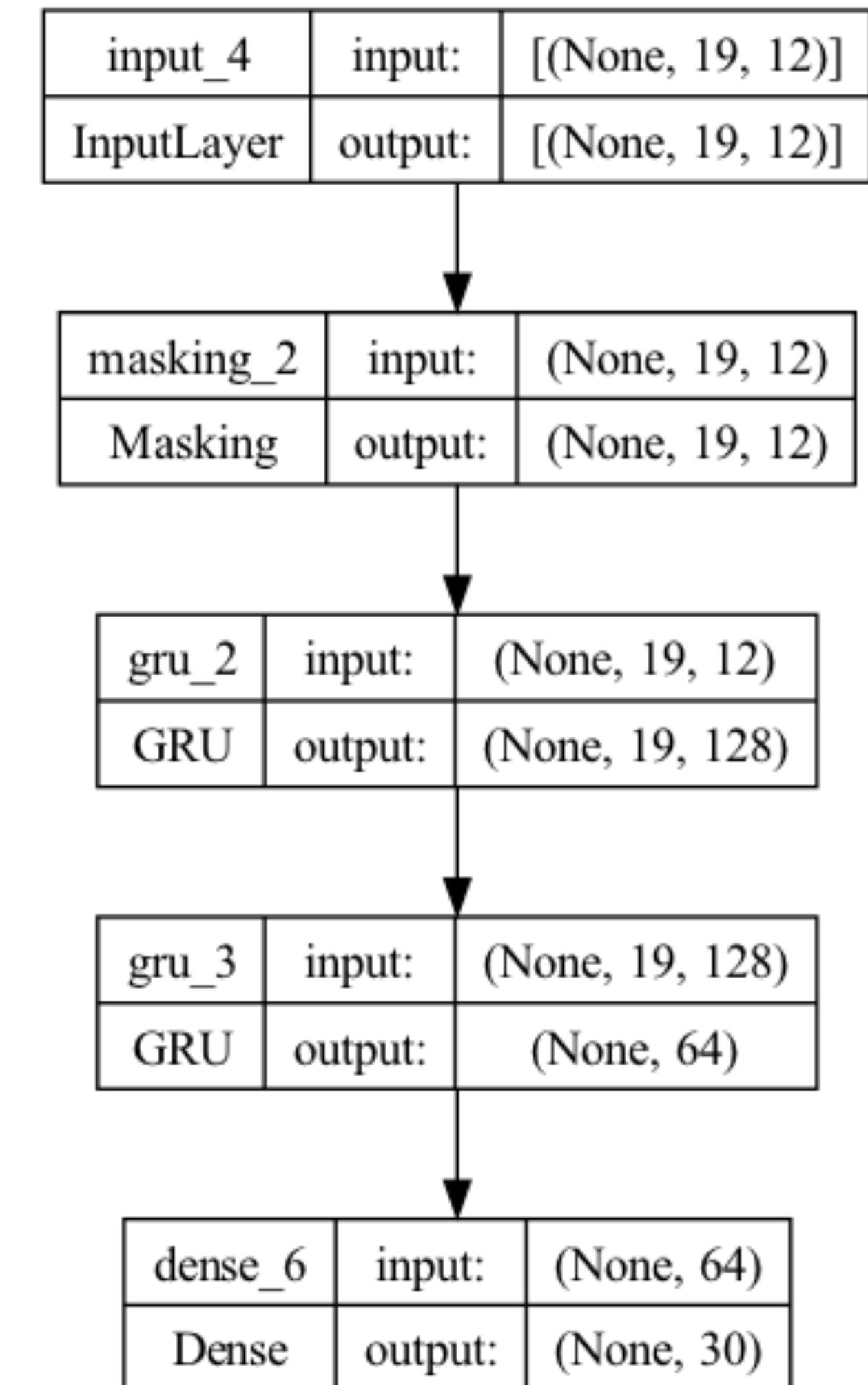
Predicting Knot Quasipositivity

- Recreated results from Knot Invariants paper
- Started with equivalent knots encoded as braid words
- Used a simple neural network with two linear layers
- Predicted knot quasipositivity (binary classification)
- 90% accuracy with just braid word
- 96% accuracy including braid positivity property

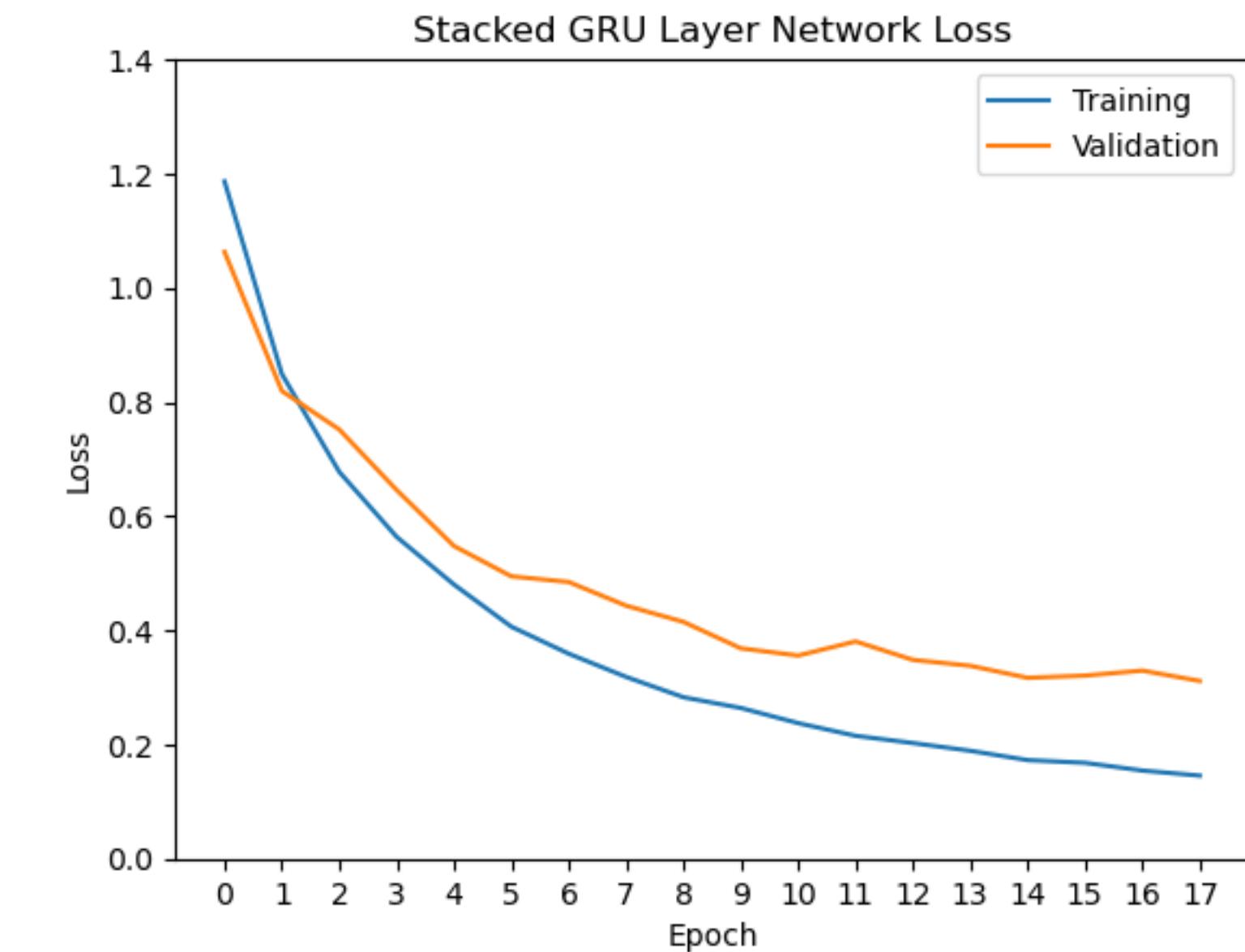
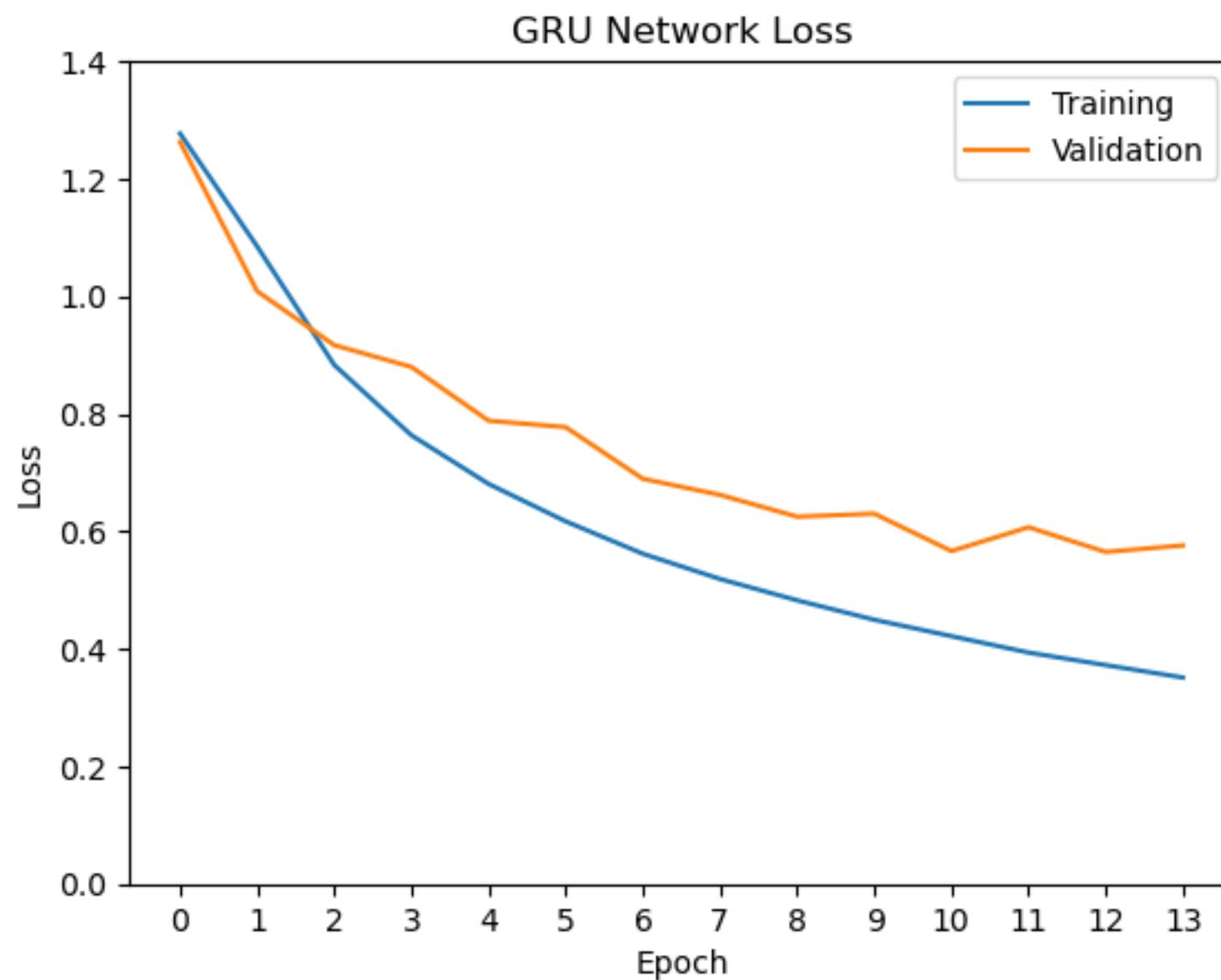
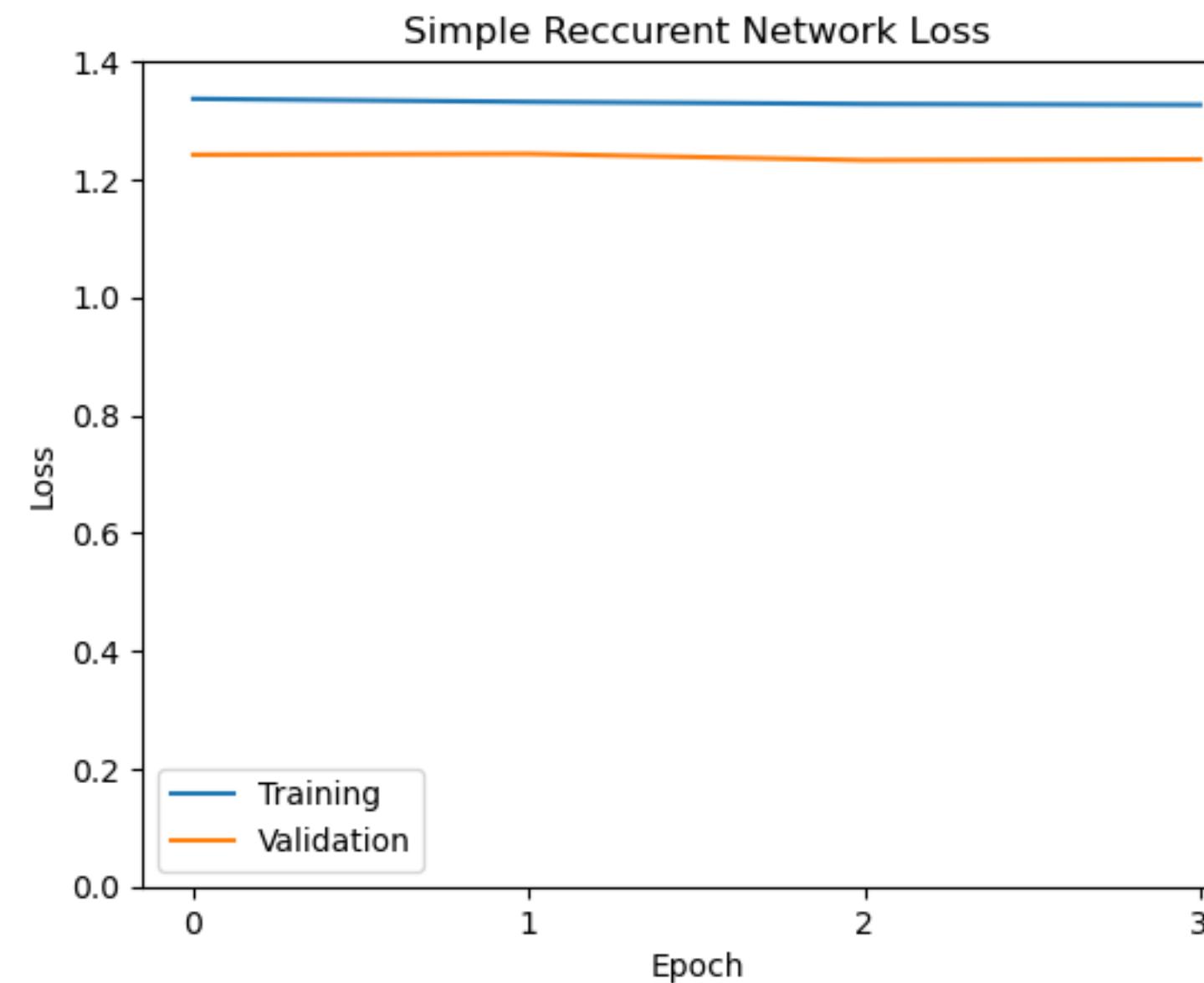
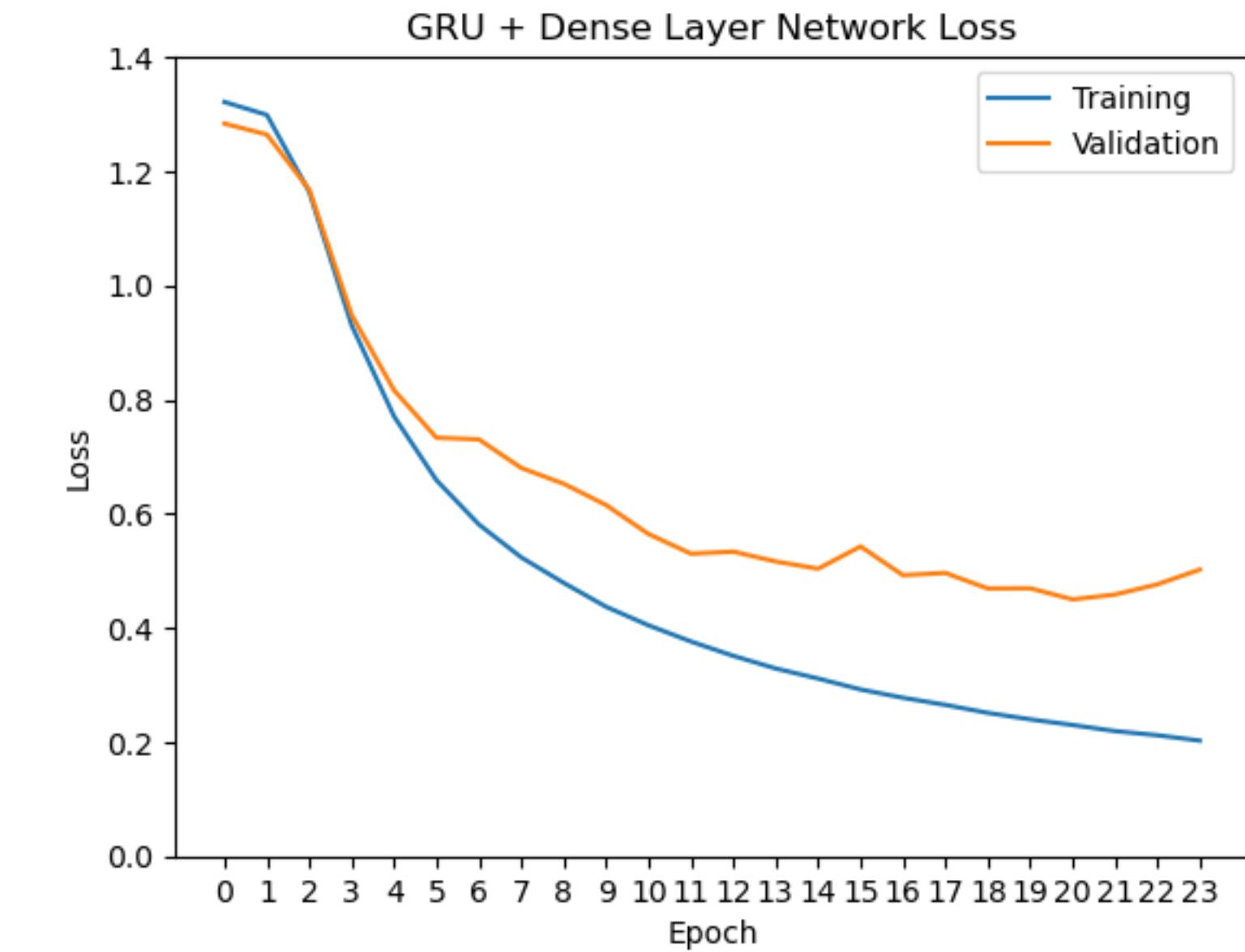
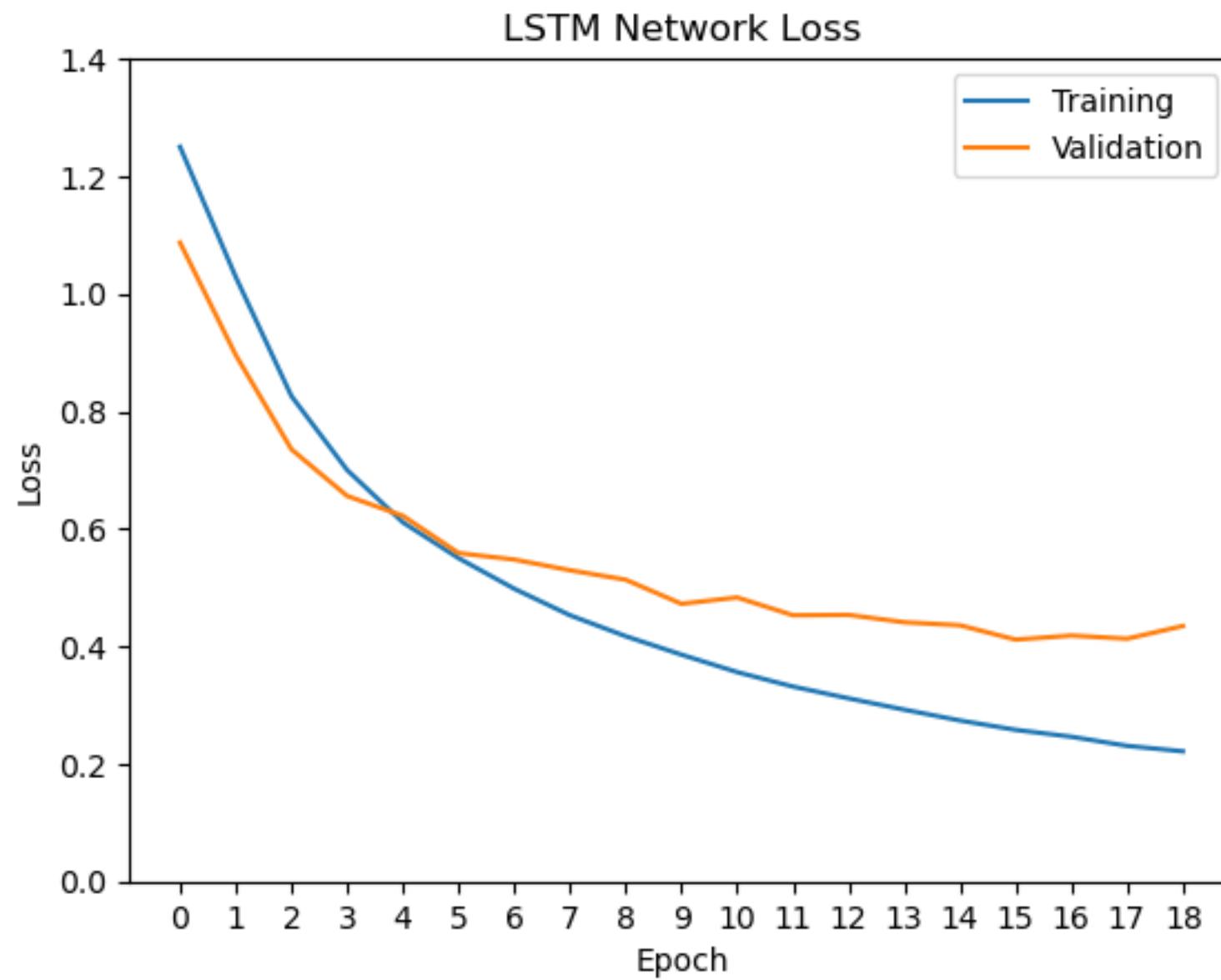
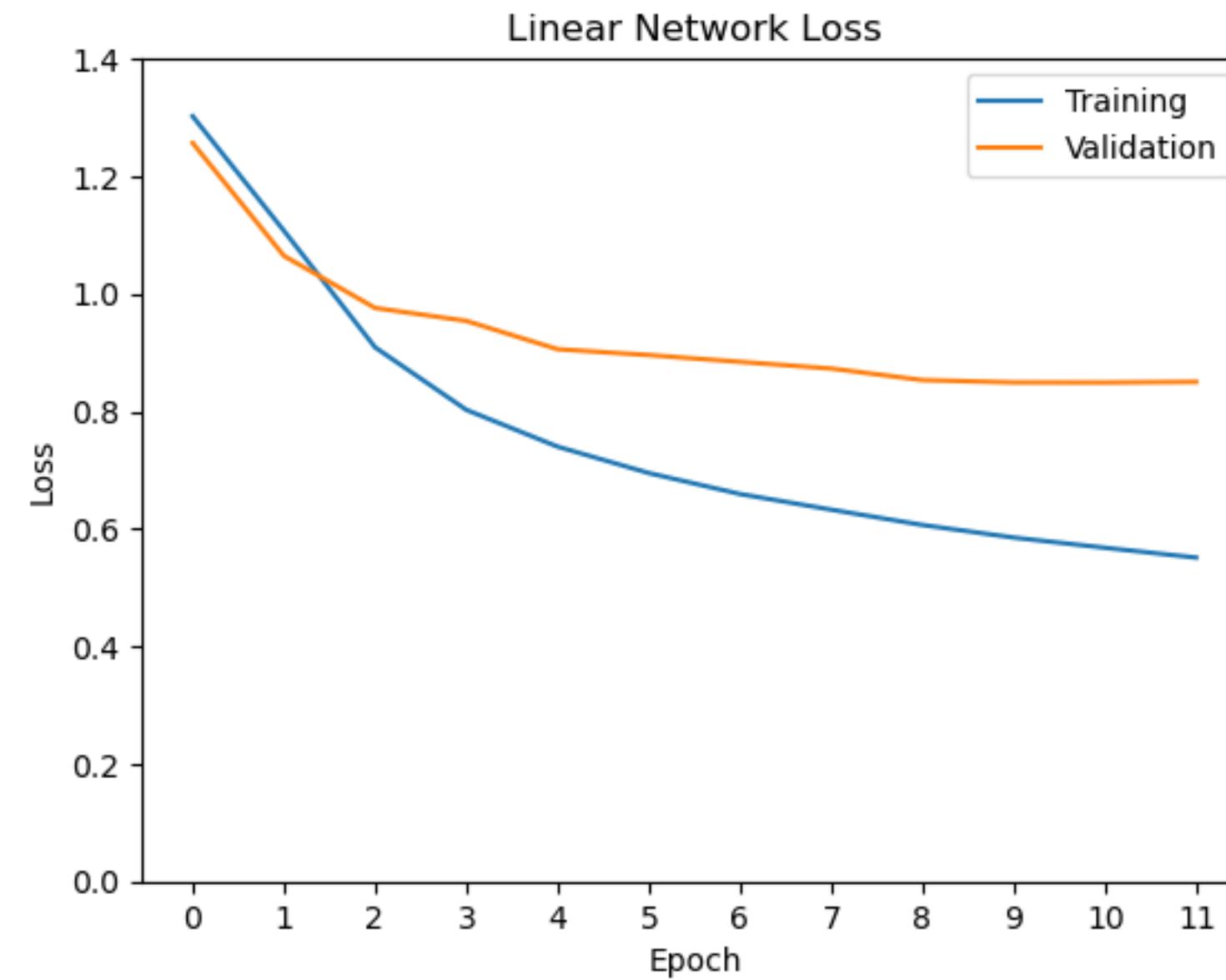


Predicting the Jones Polynomial

- Started from encoded braid words
- Predicted list of coefficients for polynomial
- Tested different architectures
 - Linear
 - Recurrent
 - LSTM
 - GRU and variants



Experimental Results



Future Research

- Test other network architectures for Jones polynomial
 - Custom recurrent layer
 - Attention / transformer network
- Predict hyperbolic volume
- Predict knot equivalence

