

# Signals and Machine Learning

Summary

# Signal Representation

# Main Topics

- Signals and Media
  - Types of Signals / Nature of Signals
- Signal Representation
  - Representations / Schemes / Domains
- Sampling and Reconstruction
  - Regular and Irregular Sampling / Reconstruction Methods / Aliasing

# Neural Networks for Signals

# Main Topics

- Neural Network Architectures
  - Data NN / Coordinate NN
- Signal Specification
  - Input Dimension / High and Low Dimension
- Neural Nets in Spectral Domain
  - Neural Tangent Kernel / Fourier Feature Mapping
- Compressive Sensing
  - Sampling Sparse Signals / Sensing and Reconstruction

# Geometric Deep Learning

Summary

# High-Dimensional Learning

# Main Topics

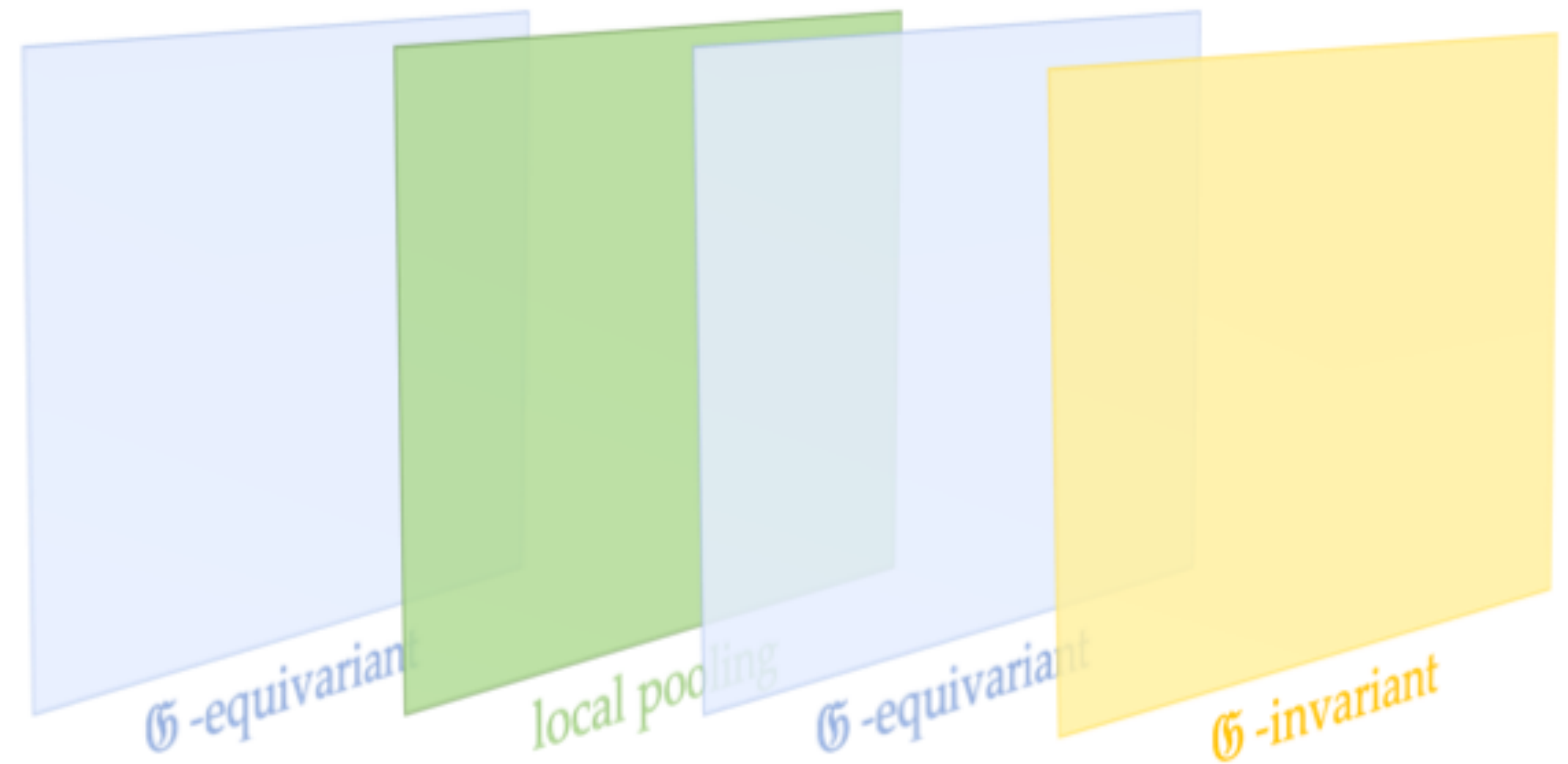
- Statistical Learning
  - Data Distribution / Approximation Model / Error Metric
  - ML Algorithm / Empirical Risk Minimization
  - Decomposition of Error (Approximation / Statistical / Optimization)
- The Curse of Dimensionality
  - Universal Approximation Theorem (Dimension Dependent)
  - Classic Regularity Fails in High Dimensions
- Addressing the Curse
  - Geometric Function Spaces (Low Dimensional Structure)
  - Novel Notions of Regularity



# Geometric Deep Learning Theory

# Basic Concepts

- Symmetry Prior
- *Invariant Functions*
- *Equivariant Functions*
- *Scale Separation Prior*
- Geometric Blueprint



# The 5 Gs

# Main Topics

- Graphs (*Sets*) [GNNs, Deep Sets, Transformers]
  - Arbitrary Ordering / Permutation Invariance / Neighbors / Message Passing
- Grids (*Images / Sequences*) [CNNs, LSTMs]
  - Fixed Neighborhood / Linear Local Aggregation / Convolution / Shift-Equivariance
- Groups (*Homogeneous Spaces*) [Group CNNs]
  - Symmetry Groups / Matching Transformations
- Gauges & Geodesics (*Meshes*) [Intrinsic CNNs]
  - Manifolds / Riemannian Metric / Parallel Transport / Gauge -Equivariant