

Voronoi Graph Neural Networks

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Machine Learning Overview

- ▶ Point Estimation
- ▶ Generative Models
 - ▶ Beta-Binomial
 - ▶ Dirichlet Multinomial
 - ▶ Naive Bayes
- ▶ Classifiers
 - ▶ Logistic Regression
 - ▶ Softmax Regression
- ▶ Dimensionality Reduction
 - ▶ Singular Value Decompositions
- ▶ Mixture Models
- ▶ Kernels and Sparse Vector Machines
- ▶ Hyperparameters

Voronoi Diagrams and Tesselations

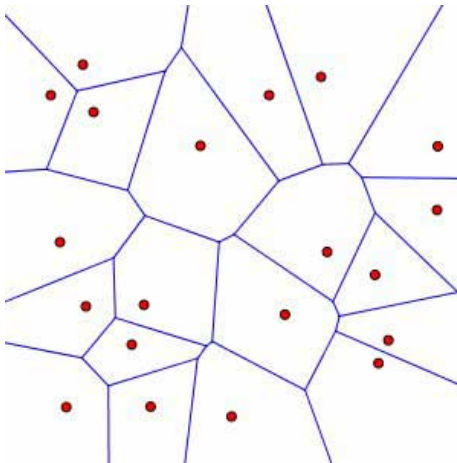


Figure: Voronoi Diagram

Neural Networks

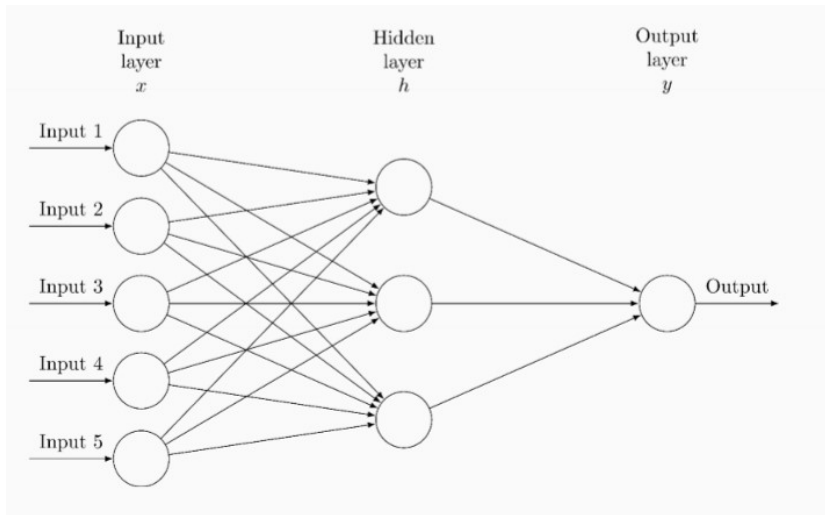


Figure: The layers in a neural network

Graph Neural Networks

- ▶ Combine graph node information with neighbouring nodes
- ▶ Aggregating the information
- ▶ Feeding it into a neural network
- ▶ Non-linear activation function

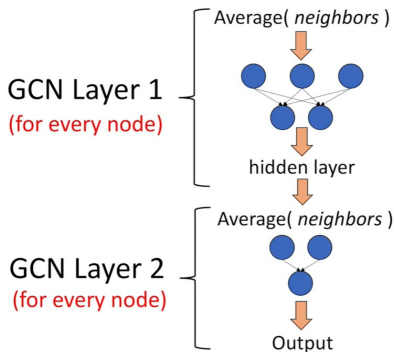


Figure: GCN Layer Outline

Voronata

From Proteins to Voronoi:

- ▶ Computing the Voronoi diagrams of given complex 3D shapes representing biological macromolecules (proteins and RNA)
- ▶ Using van der Waals radii instead of proteins or nucleic acid atoms directly

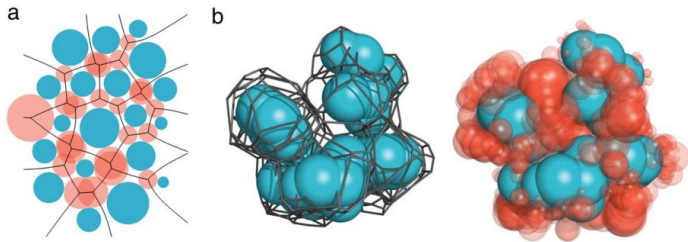


Figure: Voronoi cells and corresponding empty tangent spheres and their vertices.

Algorithm

- ▶ Two balls are considered to be neighbours if their Voronoi cells intersect.
- ▶ The intersection of four cells is a Voronoi vertex

Procedures

- ▶ Find valid quadruples of spheres
- ▶ 4 balls with at least one tangent non-intersecting sphere
- ▶ Gift-wrapping algorithm
- ▶ Find a valid neighbour of a triple in a halfspace
- ▶ Select group centers in a bounding spheres hierarchy
- ▶ Algorithm takes advantage of macromolecular structures; majority of valid atom triples have two tangent planes

VoroCNN

- ▶ Using Voronata
- ▶ Represent interatomic interactions as a Voronoi tessellation of protein atoms
 - ▶ Voronoi cell = atom
 - ▶ Voronoi cell face = interatomic contact
- ▶ Weighted unordered multi-graph representation
- ▶ Key: implicitly retains spatial relationship of atoms
- ▶ 3D tessellation can be reduced to a graph without significant information loss
- ▶ Interaction graphs used to construct a trainable neural network

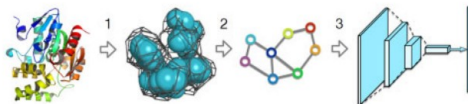


Figure: VoroCNN methods using Voronata

Nodes:

- ▶ Represent atoms in protein structure
- ▶ Each node has a vector of features describing the atom using one-hot encoding

Edges

- ▶ Contact Edges
 - ▶ Represent spatial relationship b/w atoms
 - ▶ Two atoms are in contact if their Voronoi cells have non-zero contact surface
- ▶ Covalent Edges
 - ▶ Correspond to covalent bonds between atoms in contact
- ▶ Weighted matrices equal to the respective contact surface of atoms if there is a contact/covalent edge, and 0 otherwise

- ▶ Convolutional layer contains trainable tensors
- ▶ Transformed feature matrix combining neighboring graph node information
- ▶ Non-linear activation function, ELU
- ▶ Downsampling operations
 - ▶ Reduce data representation and extract hierarchical features
 - ▶ VoroCNN uses prior topology and structure information for downsampling
 - ▶ Atoms are grouped into residues; primitive covalent edges representing the peptide chain of the protein

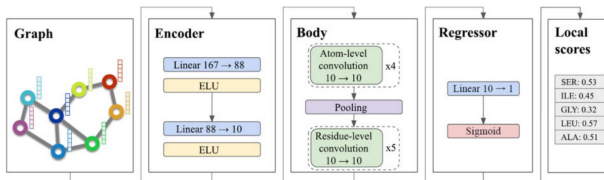


Figure: Architecture of the VoroCNN network

Neural Network Architecture:

- **Encoder**
 - Two linear layers with ELU activation function
 - Reduces the size of the feature vectors
- **Body**
 - Contains aforementioned convolutional layers and graph downsampling
- **Regressor**
 - Predicts local CAD scores, using MSE loss function, Adam optimizer and L2 regularization

Tag-VGNN

- ▶ Tactile Voronoi Graph Neural Network to achieve tactile pose estimation
- ▶ GNN used to model distribution between shear motions of tactile markers
- ▶ Voronoi diagram adds contact depth-related features
- ▶ 28% increase in pose estimation compared to vanilla GNN

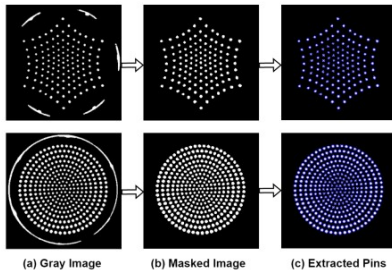


Figure: Tactile data preprocessing

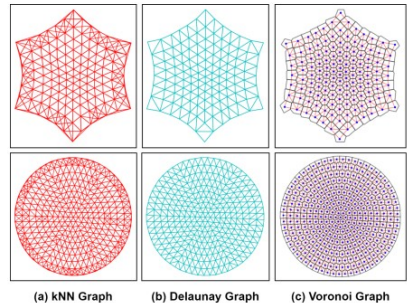


Figure: Different graph representations

Conclusion

Key Takeaways:

- ▶ Voronoi diagrams and tessellations must be constructed in a manner that maintains the structure and information of the data
- ▶ Voronoi diagram constructions are heavily dependent on the context of the data and the constrictions that exist

Next steps:

- ▶ Identifying dataset for initial testing of performance
- ▶ Deciding Voronoi computation method
- ▶ Applications and use cases
- ▶ Expected results