Final Goal Calibration

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Initial Objective

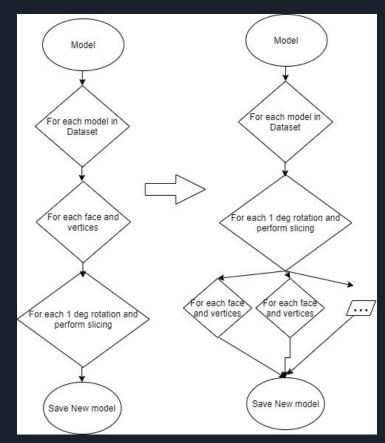
- We set out to implement the methodology described in the work Generative and Discriminative Voxel Modeling with <u>Convolutional Neural Networks</u>.
- A dataloader system was needed in order to convert the database's .off files into workable voxel objects.
- This would consist of the design of a machine learning model (a variational autoencoder) to train on the Princeton ModelNet database.
- The model would then be able to generate voxel objects given a random seed. We would use voxel-based convnets for object classification.
- A GUI would be developed that displays generated objects as well as providing a user interface to dynamically interpolate between models.



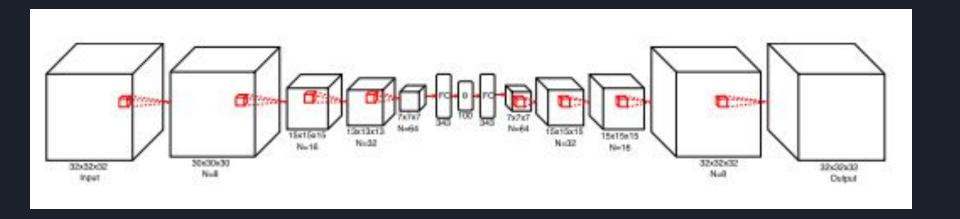
Dataloader (Feed Preprocessing)

- Code conversion improved performance in space complexity from ~10GB RAM occupancy as mentioned in the matlab code to ~2GB RAM
- Using OpenMP to Muti-thread the matrix rotation and slicing.

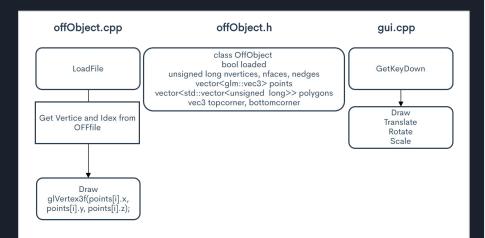
- Now we could do it online in training and testing.
- (Can be used for Real-time)



Variational Autoencoder Architecture



GUI code

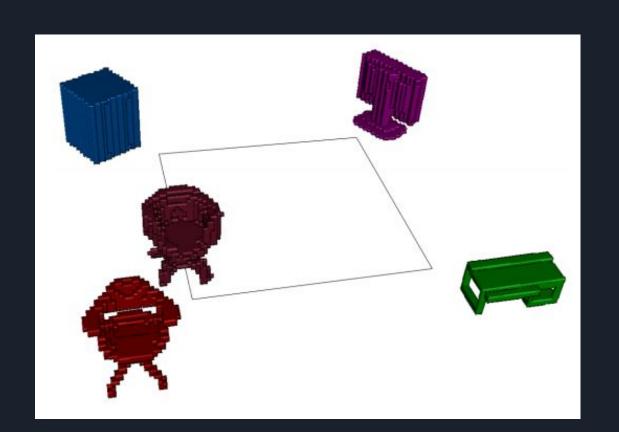


```
offObject.cpp* → × offObject.h
Miscellaneous Files

→ OffObject

             polygons.reserve(nfaces);
             for (auto i = 0; i < nfaces; ++i) {
    60
    61
                 if (!get_next_uncommented_line(infile, info)) {
                     return false;
    62
    63
    64
                 std::istringstream info stream(info);
    65
                 unsigned long n;
    66
                 unsigned long index;
                 info_stream >> n;
    68
                 polygons.emplace_back(n);
                 for (auto j = 0; j < n; ++j) {
    69
                     info_stream >> index;
    70
    71
                     polygons[i][j] = index;
    72
    73
    74
             infile.close();
    75
             return (loaded = true);
    76
    77
       pvoid OffObject::draw(DISPLAY TYPE displayType) const {
    78
    79
             if (displayType == DISPLAY_TYPE::POINTS) {
    80
                 glBegin(GL POINTS);
    81
                 std::for_each(points.begin(), points.end(),
                     [](const auto& point) { glVertex3f(point.x, point.y, point.z); });
    82
    83
                 glEnd();
    84
             else {
    85
                 GLenum mode = (displayType == DISPLAY_TYPE::POLYGONS) ? GL_POLYGON : GL_LINE_LOOP;
    86
    87
                 for (const auto& indexes : polygons) {
                     glBegin(mode);
    88
                     for (const auto& i : indexes) {
    89
                         glVertex3f(points[i].x, points[i].y, points[i].z);
    90
    91
                     glEnd();
    92
    93
    94
    95
```

GUI Display



Concerns to Address

- <u>C++ Tensorflow API</u> does not natively possess certain functionalities (such as 3D batch normalization), thus requiring some workarounds.
- Processing the database proved to be more involved than initially thought.
- Data loading in C++ requires O($n^2 + k$) (n > k) based on batchsize. Reduction in time complexity with rotating first and voxelizing to O($k^2 + n$).
- Advanced features of GUI, such as interpolation between objects, as well as voxception based classifiers, have proven to be rather complex and time-consuming to implement.

Project Adjustment and Final Deliverables

- We will have a <u>functioning dataloader</u> and VAE model, capable both of training on and generating these voxel objects.
- We intend to table the implementation of <u>Voxception classifiers</u>, though we may given time availability. (According to time available)
- Our GUI will <u>display models of the training set and generated objects</u>, as well as the progress of the VAE, but we do not intend to featured

Any Questions?