patchDPCC: A Patchwise Deep Compression Framework for Dynamic Point Clouds

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Appendix

Algorithms of adjusting a P-patch

In this section, we describe two algorithms adjusting point number of a P-patch \mathbf{P}^P to align with a transformed I-patch \mathbf{P}^{I_t} . In the cases of the P-patch having less or more points than the transformed I-patch, we need copy points to or delete points from \mathbf{P}^P . As described in the methodology section, we fuse two patches and build an octree over their points before the copy or deletion. Suppose that we have m leaf nodes in the octree, and each is a point set

$$\mathbf{T}_i = \mathbf{S}_i^I \cup \mathbf{S}_i^P, i = 1, \dots, m,$$

where \mathbf{S}_i^I and \mathbf{S}_i^P are sets of points belonging to the transformed I-patch and the P-patch, respectively. We then follow Algorithm 1 to copy points from the transformed I-patch to the P-patch. It is worth noting that we use three priority queues to ensure that the copy operations are not consecutively performed in neighboring leaf nodes.

Algorithm 1 Find points in a transformed I-patch to copy.

Input: \mathbf{P}^{I_t} , \mathbf{P}^P , m, r, $\{(\mathbf{T}_i, \mathbf{S}_i^I, \mathbf{S}_i^P): i=1,\ldots,m\}$, nearest_pts (\cdot) returns the nearest point pair from two point sets, $\mathrm{dist}(\cdot)$ returns the distance of two octree nodes.

```
\triangleright Q_* are priority queues
  1: Initialize Q_0, Q_1, Q_2
 2: \mathbf{U} \leftarrow \emptyset, \mathbf{R} \leftarrow \emptyset
                                                       \triangleright \mathbf{R} stores the points to be copied
  3: s^I \leftarrow 0, s^P \leftarrow 0
 4: for i \leftarrow 1 to m do
                if |\mathbf{S}_i^I| > 0 and |\mathbf{S}_i^P| > 0 then
 5:
                        s^I \leftarrow s^I + |\mathbf{S}_i^I|, s^P \leftarrow s^P + |\mathbf{S}_i^P|
  6:
                        \operatorname{push}(Q_0,(\mathbf{T}_i,\mathbf{S}_i^I,\mathbf{S}_i^P))
  7:
                else if |\mathbf{S}_{i}^{I}| > 0 then
  8:
                        \mathbf{U} \leftarrow \mathbf{U} \cup \mathbf{S}_{i}^{I}
 9:
10:
                end if
11: end for
12: if s^{I} - s^{P} > |\mathbf{P}^{I_{t}}| - |\mathbf{P}^{P}| then
                for i \leftarrow 0 to 2 do
13:
                        \mathbf{while}\ (\mathbf{T}, \widetilde{\mathbf{S}^I}, \mathbf{S}^P) \leftarrow \mathrm{pop}(Q_i)\ \mathbf{do}
14:
                                \qquad \qquad \triangleright \text{ Pop the element with the most points in } \mathbf{S}^I \\ p_{min}^I, p_{min}^P \leftarrow \text{nearest\_pts}(\mathbf{S}^I, \mathbf{S}^P) 
15:
16:
```

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\begin{aligned} \mathbf{R} \leftarrow \mathbf{R} \cup \{p_{min}^I\} \\ \mathbf{if} \ |\mathbf{R}| \geq |\mathbf{P}^{I_t}| - |\mathbf{P}^P| \ \mathbf{then} \end{aligned}
17:
 18:
 19:
                                                   return R
                                          end if
20:
                                         \begin{aligned} &\mathbf{S}^I \leftarrow \mathbf{S}^I - \{p_{min}^I\} \\ &\mathbf{S}^P \leftarrow \mathbf{S}^P \cup \{p_{min}^I\} \\ &\text{push}(Q_2, (\mathbf{T}, \mathbf{S}^I, \mathbf{S}^P)) \end{aligned}
21:
22:
23:
                                          if i < 2 then
24:
                                                    for (\mathbf{T}', \mathbf{S}'^I, \mathbf{S}'^P) \in Q_i do
25:
                                                              if \operatorname{dist}(\mathbf{T},\mathbf{T}') < 2r then
26:
                                                                        delete(Q_i, (\mathbf{T}', \mathbf{S}'^I, \mathbf{S}'^P))
27:
                                                                        \operatorname{push}(Q_{i+1}, (\mathbf{T}', \mathbf{S}'^I, \mathbf{S}'^P))
28:
                                                              end if
29:
30:
                                                    end for
31:
                                          end if
32:
                               end while
33:
                      end for
34: else
                     for (\mathbf{T}, \mathbf{S}^I, \mathbf{S}^P) \in Q_0 do
35:
36:
                               \mathbf{R} \leftarrow \mathbf{R} \cup \mathbf{S}^I
37:
                      end for
                      \begin{array}{c} \textbf{while} \ |\mathbf{R}| < |\mathbf{P}^{I_t}| - |\mathbf{P}^P| \ \textbf{do} \\ p^U, p^P \leftarrow \text{nearest\_pts}(\mathbf{U}, \mathbf{P}^P) \\ \mathbf{R} \leftarrow \mathbf{R} \cup \{p^U\}, \mathbf{U} \leftarrow \mathbf{U} - \{p^U\} \end{array} 
38:
39:
40:
41:
                      end while
42:
                      return R
43: end if
```

Similarly, Algorithm 2 is designed to delete excessive points in the P-patch.

Each time a point has been copied between two patches, we want to search the neighboring leaf nodes within 2r and push them to the queue with a lower priority. This is achieved by employing the radius search of KD-tree, whose time complexity is $\mathcal{O}(logN)$. N is the point number of a patch. As most N points are copied from the transformed I-patch, the complexity of the Algorithm 1 is $\mathcal{O}(NlogN)$. As for Algorithm 2, searching the neighbors is also the bottleneck, and its time complexity is $\mathcal{O}(NlogN)$ as well.

Algorithm 2 Find points in a P-patch to delete.

Input: \mathbf{P}^{I_t} , \mathbf{P}^P , m, r, $\{(\mathbf{T}_i, \mathbf{S}_i^P): i = 1, \ldots, m\}$, randpick (\cdot) randomly picks a point from a point set, $\mathrm{dist}(\cdot)$ returns the distance of two octree nodes.

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```
1: Initialize Q_0, Q_1, Q_2
                                                                         \triangleright Q_* are priority queues
 2: R ← ∅
                                                    \triangleright \mathbf{R} stores the points to be deleted
  3: for i \leftarrow 1 to m- do
                if |\mathbf{S}_i^P| > 0 then
  4:
                        \operatorname{push}(Q_0,(\mathbf{T}_i,\mathbf{S}_i^P))
  5:
  6:
                end if
  7: end for
  8: for i \leftarrow 0 to 2 do
                while (\mathbf{T}, \mathbf{S}^P) \leftarrow \text{pop}(Q_i) do
  9:
                       Pop the element with the most points in \mathbf{S}^P
p^P \leftarrow \text{randpick}(\mathbf{S}^P)
\mathbf{R} \leftarrow \mathbf{R} \cup \{p^P\}
if |\mathbf{R}| \ge |\mathbf{P}^P| - |\mathbf{P}^{I_t}| then
return \mathbf{R}
10:
11:
12:
13:
14:
                        end if
15:
                      end if \mathbf{S}^P \leftarrow \mathbf{S}^P - \{p^P\} push(Q_2, (\mathbf{T}, \mathbf{S}^P)) if i < 2 then for (\mathbf{T}', \mathbf{S}'^P) \in Q_i do if \mathrm{dist}(\mathbf{T}, \mathbf{T}') < 2r then
16:
17:
18:
19:
20:
                                               delete(Q_i, (\mathbf{T}', \mathbf{S}'^P))
21:
                                               \operatorname{push}(Q_{i+1},(\mathbf{T}',\mathbf{S'}^{P}))
22:
23:
                                        end if
                               end for
24:
25:
                        end if
                end while
26:
27: end for
```

Visualization results

The visualization results are shown in Figure 1. Our approach clearly maintains the geometric details in the original point clouds and reconstructs visually appealing results.



Figure 1: Comparing visual results on MPEG 8i dataset using different methods(a-e), and the ground truth(f)