A Implementation Details

Datasets used contain videos with various codecs, which are directly decoded into RGB frames and motion vectors. We implement MVCGC based on “pyav“, a Pythonic binding for the FFmpeg libraries used for decoding RGB frames and motion vectors on-the-fly. We sample a 32-frame clip from each decoded video with a spatial resolution of $128 \times 128$ pixels. We apply clip-wise random crops, horizontal flips, Gaussian blur, graying, and color jittering in the spatial dimension while random sampling a continuous 32-frame clip in the temporal dimension for data augmentation. Motion vectors are processed as follows: those of I-frames are directly set as all-zero matrices; and the forward motion vectors in B-frames are used, whose information resembles those in P-frames. A third zero channel is stacked to the two-channel motion vector, and motions exceeding 15 pixels are truncated. The final values are projected to $[0, 255]$ as RGB images, and are only spatially augmented (i.e., crops and flips). A non-linear projection head is attached above each encoder during the pre-training stage and is removed for downstream evaluations.

B Architecture

We use the S3D architecture [Xie et al., 2018] for all experiments following the previous work [Han et al., 2020]. At the pre-training stage, S3D is followed by a non-linear projection head, which consists of two fully-connected (FC) layers and is removed when evaluating downstream tasks. The detailed dimensions are shown in Table 1.

<table>
<thead>
<tr>
<th>Stage</th>
<th>Detail</th>
<th>Output size: $T \times H(W) \times C$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Backbone</td>
<td>followed by average pooling</td>
<td>$1 \times 1^2 \times 1024$</td>
</tr>
<tr>
<td>Projection</td>
<td>FC-1024→ReLU→FC-128</td>
<td>$1 \times 1^2 \times 128$</td>
</tr>
</tbody>
</table>

Once pre-trained, we replace the non-linear projection head with a single linear layer for classification tasks. The detailed dimensions are illustrated in Table 2.

Table 1: The encoder architecture of both RGB and motion vector streams at the pre-training stage. “FC-1024” and “FC-128” denote the output dimension of each fully-connected layer, respectively.

Table 2: The classifier architecture for evaluating the representations on the action classification tasks. Note that “FC-num_class” denotes the output dimension of the fully-connected layer (i.e., the number of action classes).

<table>
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<th>Stage</th>
<th>Detail</th>
<th>Output size: T×H(W)×C</th>
</tr>
</thead>
<tbody>
<tr>
<td>S3D</td>
<td>followed by average pooling</td>
<td>1 × 1² × 1024</td>
</tr>
<tr>
<td>Linear layer</td>
<td>one layer: FC-num_class</td>
<td>1 × 1² × num_class</td>
</tr>
</tbody>
</table>

C Qualitative Results for Video Retrieval

Here we demonstrate the qualitative results for video retrieval. Figure 1 visualizes query video clips and their Top-3 Nearest Neighbors from the UCF101 training set using the concatenation of MVCGC two stream embeddings. As shown in the figure, the representations learned by our MVCGC are able to retrieve videos belonging to the same semantic categories.

References


Checklist

1. For all authors...
   (a) Do the main claims made in the abstract and introduction accurately reflect the paper’s contributions and scope? [Yes]
   (b) Did you describe the limitations of your work? [No]
   (c) Did you discuss any potential negative societal impacts of your work? [No]
   (d) Have you read the ethics review guidelines and ensured that your paper conforms to them? [Yes]

2. If you are including theoretical results...
   (a) Did you state the full set of assumptions of all theoretical results? [No] There are no theoretical result
   (b) Did you include complete proofs of all theoretical results? [No] There are no theoretical result

3. If you ran experiments...
   (a) Did you include the code, data, and instructions needed to reproduce the main experimental results (either in the supplemental material or as a URL)? [No] The code are proprietary now.
   (b) Did you specify all the training details (e.g., data splits, hyperparameters, how they were chosen)? [Yes]
   (c) Did you report error bars (e.g., with respect to the random seed after running experiments multiple times)? [No]
   (d) Did you include the total amount of compute and the type of resources used (e.g., type of GPUs, internal cluster, or cloud provider)? [Yes]

4. If you are using existing assets (e.g., code, data, models) or curating/releasing new assets...
   (a) If your work uses existing assets, did you cite the creators? [Yes]
   (b) Did you mention the license of the assets? [Yes]
   (c) Did you include any new assets either in the supplemental material or as a URL? [No]
   (d) Did you discuss whether and how consent was obtained from people whose data you’re using/curating? [No]
   (e) Did you discuss whether the data you are using/curating contains personally identifiable information or offensive content? [No]

5. If you used crowdsourcing or conducted research with human subjects...
   (a) Did you include the full text of instructions given to participants and screenshots, if applicable? [No]
   (b) Did you describe any potential participant risks, with links to Institutional Review Board (IRB) approvals, if applicable? [No]
   (c) Did you include the estimated hourly wage paid to participants and the total amount spent on participant compensation? [No]