Video summarization via kcenter clustering

Filippo Geraci

May 7, 2019

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- You want to know the content without read it ?

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William J. Drake is the President of Computer Professionals for Social Responsibility, a global civil society organization, and an ICT policy consultant based in Geneva, writerland.

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Our goal

• We attempt to produce abstracts for video



Motivations for video summarization

- Discover quickly if the video content is of interest without watching it
- Allow comfortable browsing experience in video databases

Example

• Find a particular episode in a TV series



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Video summary taxonomy



Summary computation time

- .On-line: for web applications, allow user to personalize it (i.e choose summary size)
- .Off-line: must to be computed in advance. (It is not acceptable to wait for a time comparable with the video length)



Video Summarization Pipeline

- Input: scenes/frames. 256-dimensional HSV vectors, extracted only once (on video submission)
- **Clustering**: determine the scenes to show. Customizations:
 - length of the storyboard,
 - clustering running time.
- **Post-processing**: filter out near-duplicates and extract the selected frames/scenes from the video.



- Scene: bounded set of consecutive frames,
 - represented by the mean of the HSV vectors of the frames
- a movie is a partition of scenes,
- a frame fall in just a single scene,
- audio and video must be combined to determine scene,
 - video cut alone can be only a camera change,
 - silence is frequent in dialogs among people.

A metric space for video (1)



- Convert each pixel in its HSV representation
- Build a histogram of the frequences for each component

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Generalized Jaccard distance (GJC)

• Given two vectors $H_1 = h_{1,1}, \ldots, h_{1,256}$ and $H_2 = h_{1,1}, \ldots, h_{1,256}$ representing two frames, the Generalized Jaccard distance is defined as:

$$d(H_1.H_2) = 1 - \frac{\sum_{i=1}^{256} \min(h_{1,i}, h_{1,i})}{\sum_{i=1}^{256} \max(h_{1,i}, h_{1,i})}$$

- The vectors H_i endowed with the Generalized Jaccard distance define a metric space,
- the GJC exploit better the distance between two frames with respect to other distance functions used in the literature,
- GJC is fast and simple to compute.

• Partition data in homogeneous groups;

- Data independent procedure;
 - Available also when no a-priori knowledge about the data domain;
 - No problem dependent optimizations allowed;
- Many fields of application: IR, bioinformatics.

Cluster hypothesis

If two objects are closely related and the former is also related to a third object, then more likely also the latter has a similar relation.

K-center problem



• *K*-center minimize the largest cluster diameter

```
\min\max_{j}\max_{x\in C_j}M_v(x,\mu_j)
```

Data: Let *O* be the input set, *k* the number of clusters **Result**: *C*, *k*-partition of *O* C = x such that *x* is an arbitrary element of *O*; **for** $i = 0; i < k; i + + \mathbf{do}$ | Pick the element *x* of $O \setminus C$ furthest from the closest element in *C*; $C_i = C_i = x;$ **end forall** $x \in O \setminus C$ **do** | Let *i* such that $d(c_i, x) < d(c_j, x), \forall j \neq i \ C_i.$ append (*x*);

end

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Based on two cases

- There is one center of FPF for each center of the optimal clustering
- 2 There are two centers of FPF for each center of the optimal clustering

Unproved here: there are no other possible cases







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Exploit triangular inequality



• Most of the time spent to find the closest center

• In some cases can avoid to check all points of a cluster

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Drawbacks in FPF



- Outlayers became centers with high probability
- Poor clusters and affect the choice of k
- Centers are not so representative

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Data: Let O be the input set, k the number of desired clusters **Result**: \mathcal{C} : a k-partition of O Initialize R with a random sample of size $\sqrt{|O|k}$ elements of O; $\mathcal{C} = \mathbf{FPF}(R, k);$ forall $C_i \in \mathcal{C}$ do $\mu_i = \text{getCenter}(C_i);$ end forall p in $O \setminus R$ do assign p to cluster C_i such that $d(p, \mu_i) < d(p, \mu_i), \forall j \neq i$; end



- Given a set of points C, let $a, b \in C$ be two diametral points of C. The **medoid** of C is the point $x \in C$ that minimize:
- M(x) = |D(a,x) D(b,x)| + |D(a,x) + D(b,x)|,

- Select a random point R
- find the farthest point A from R
- find the farthest point B from A
- Return (A, B) as the diametral pair

Data: Let O be the input set, k the number of desired clusters **Result**: \mathcal{C} : a k-partition of O Initialize R with a random sample of size $\sqrt{|O|k}$ elements of O; $\mathcal{C} = \mathbf{FPF}(R, k);$ forall $C_i \in \mathcal{C}$ do $t_i = \text{getRandomPoint} (C_i);$ $a_i = c_i$ such that max $d(c_i, t_i)$ for each $c_i \in C_i$; $b_i = c_i$ such that max $d(c_i, a_i)$ for each $c_i \in C_i$; $m_i = c_i$ such that $\min |d(c_i, a_i) - d(c_i, b_i)| + |d(c_i, a_i) + d(c_i, b_i) - d(a_i, b_i)|;$ end forall p in $O \setminus R$ do assign p to cluster C_i such that $d(p, m_i) < d(p, m_i), \forall j \neq i$; if $d(p,b_i) > d(a_i,b_i)$ then $a_i = p$; if $d(p, a_i) > d(a_i, b_i)$ then $b_i = p$; if $d(p, b_i) > d(a_i, b_i)$ or $d(p, a_i) > d(a_i, b_i)$ then $|m_i = c_i$ such that $\min |d(c_i, a_i) - d(c_i, b_i)| + |d(c_i, a_i) + d(c_i, b_i) - d(a_i, b_i)|;$

 \mathbf{end}

end

How many clusters?



- Problem dependent question
- Few clusters are often not homogeneous
- Too many break homogeneous clusters

.. and so?

- ad-hoc solutions
 - may be not feasible
- theoretical approachs (i.e stability):
 - always available
 - not related to the problem

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Approximations for video summaries



Adding a new point to cluster

- Most of the time is spent to decide for each point to which cluster it should be assigned.
- If the new point is "close" to the previous one, it is inserted in the same cluster.

Distance between two consecutive frames





- Given a set of points C, let $a, b \in C$ be two diametral points of C. The **medoid** of C is the point $x \in C$ that minimize:
- M(x) = |D(a, x) D(b, x)| + |D(a, x) + D(b, x)|,



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Video summary using the approximated FPF algorithm



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• Python module for clustering and other machine learing tasks: http://scikit-learn.org/stable/modules/ clustering.html#clustering