# Are Actor and Action Semantics Retained in Video Supervoxel Segmentation?

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# Video Understanding; What?

• Example human synopsis: "A person is climbing a rock-wall."



Applications

- Real-time / Interactive
  - Human computer interaction and entertainment.
  - Healthcare monitoring and surveillance.
- Off-line
  - Video indexing and search.
  - Video to language.
  - Sports analysis.

#### Video Understanding; What?



Method: Laptev. "On Space-Time Interest Points." IJCV 2005.



Method: Wang et al. "Action Recognition by Dense Trajectories." CVPR 2011.

# The (Very Common) Bag-of-Features Pipeline

Source: materials adapted from Laptev's CVPR 2008 slides.



• Examples include Schüldt et al. ICPR 2004, Niebles et al. IJCV 2008, and many works building on this basic idea.

#### **Supervoxel Segmentation: Toward a Representation with Rich Semantics?**



Note: images here are videos in the original slides. [Xu, Xiong and Corso, ECCV 2012]

# **Study Questions**

- Primary Question:
  - Do the segmentation hierarchies retain enough information for the human perceiver to discriminate
    - Actor? (human or animal)
    - Action? (climbing, crawling, eating, flying, jumping, running, spinning, walking)
- Secondary Questions:
  - How does the semantic retention vary with
    - Density of the supervoxels?
    - Actor (human versus animal)?
    - Background (static versus moving)?
  - How does response time vary with action?



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#### **Video Supervoxel Segmentation**

# **Hierarchical Video Supervoxel Segmentation**

- Basic problem statement:
- Segmentation hierarchy

$$\mathcal{S} \doteq \{S^1, S^2, \dots, S^h\}$$

 $S^i \doteq \{s_1, s_2, \dots\}$  such that  $s_j \subset \Gamma$ ,  $\cup_j s_j = \Gamma$ , and  $s_i \cap s_j = \emptyset$  for pairs i, j

Segmentation

 $|\mathcal{S}^*| = \operatorname{argmin} E(\mathcal{S})$ 

• Use the minimum spanning tree method.

$$E(S^{1}|\mathcal{V}) = \tau \sum_{s \in S^{1}} \sum_{e \in MST(s)} w(e) + \sum_{s,t \in S^{1}} \min_{e \in \langle s,t \rangle} w(e)$$

Stage 2: Plaket addoption set interformed in a set in the set of t



Video Input

#### **Streaming Hierarchical Video Segmentation**



Stream\_Video

Note: images here are xide sin the Arigeral Side CCV 2012]

#### **Main Study**

# Study Setup: Data Set



- Video Time (Action starts immediately after play.)
  - About 4 Seconds / shown at half-frame-rate
- Stratified according to
  - Actors: human or animal
  - Background: static or moving
  - Actions: climbing, crawling, eating, flying, jumping, running, spinning, walking
- 3 Levels of the segmentation hierarchy
  - Fine: 8th level / Medium: 16th level / Coarse: 24th level
  - Q: a best level in the hierarchy?
- In total, we have 96 videos
  - 2 actors \* 2 backgrounds \* 8 acts \* 3 levels: images here are videos in the original slides.

#### **Study Setup: Data Collection**

- Study cohort of 20 college-age participants.
  - No student is studying segmentation.
  - Each participant is shown 32 videos and sees a given (input) video only once (in a single segmentation level).

Segmentation Video HIT

- Participants never see the input RGB videos.



# Select Actor Human Animal Select Act Select Act Climbing Crawling Eating Walking Don't Know Act or Actor Flying Spinning Running Jumping Submit Results Note: images here are videos in the original slides.

#### **Discriminate Actor?** (human or animal)

	un	hu	an
unknown	0	0	0
human	0.11	0.86	0.03
animal	0.17	0.05	0.78

#### **Confusion Matrix**

- Overall actor discrimination rate: 82.4%.
- Unknown was chosen when less confident.
- Suspects
  - Performance is so high due to one dominant actor.
    - Locate by svx motion, then determine by svx shape.
  - Performance on human is better than animal due to more variation of animal location and orientation.

#### **Discriminate Action? (one of eight)**

• Overall action discrimination rate: 70.4%.

	un	wl	sp	rn	jm	ea	cl	cr	fl	
unknown	0	0	0	0	0	0	0	0	0	
walking	0.11	0.57	0.12	0.12	0	0.01	0.01	0.04	0	
spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06	
running	0.01	0.07	0.07	0.79	0.04	0	0	0.01	0	
jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09	
eating	0.19	0	0	0	0	0.76	0.04	0	0.01	
climbing	0.06	0.01	0	0	0.03	0	0.90	0	0	
crawling	0.20	0.03	0	0.06	0.01	0	0.01	0.69	0	
flying	0.19	0.03	0.01	0	0.016t	e:OnQalge	ടാറിള്ള പ	ar <del>o</del> .Wdeo	oDi7 <mark>fe</mark>	original sl

• Dominant unidirectional motion.

	un	wl	sp	rn	jm	ea	cl	cr	fl
unknown	0	0	0	0	0	0	0	0	0
walking	0.11	0.57	0.12	0.12	0	0.01	0.01	0.04	0
spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06
running	0.01	0.07	0.07	0.79	().04	0	0	0.01	0
jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09
eating	0.19	0	0	0	0	0.76	0.04	0	0.01
climbing	0.06	0.01	0	0	0.03	0	0.90	0	0
crawling	0.20	0.03	0	0.06	0.01	0	0.01	0.69	0
flying	0.19	0.03	0.01	0	0.01	0.01	0.03	0.03	0.70



Human\_Running



Human\_Climbing





Animal\_Running Animal\_Climbing Note: images here are videos in the original slides.

• Dominant unidirectional motion.

	un	wl	sp	rn	jm	ea	cl	cr	fl
unknown	0	0	0	0	0	0	0	0	0
walking	0.11	0.57	0.12	0.12	0	0.01	0.01	0.04	0
spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06
running	0.01	0.07	0.07	0.79	().04	0	0	0.01	0
jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09
eating	0.19	0	0	0	0	0.76	0.04	0	0.01
climbing	0.06	0.01	0	0	0.03	0	0.90	0	0
crawling	0.20	0.03	0	0.06	0.01	0	0.01	0.69	0
flying	0.19	0.03	0.01	0	0.01	0.01	0.03	0.03	0.70



Human\_Running



Human\_Climbing



Animal\_Running Animal\_Climbing Note: images here are videos in the original slides.

• Semantic ambiguity in videos.

	un	wl	sp	rn	jm	ea	cl	cr	fl	
unknown	0	0	0	0	0	0	0	0	0	
walking	0.1	0.57	0.12	0.12	0	0.01	0.01	0.04	0	
spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06	
running	0.01	0.07	0.07	0.79	0.04	0	0 0.01		0	
jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09	
eating	0.19	0	0	0	0	0.76	0.04	0	0.01	
climbing	0.06	0.01	0	0	0.03	0	0.90	0	0	
crawling	0.20	0.03	0	0.06	0.01	0 0.01 0.0		0.69	0	
flying	0.19	0.03	0.01	0	0.01	0.01	0.03	0.03	0.70	



Human\_Walking



Human\_Jumping





Animal\_Walking Animal Jumping Note: images here are videos in the original slides.

• Semantic ambiguity in videos.

	un	wl	sp	rn	jm	ea	cl	cr	fl	
unknown	0	0	0	0	0	0	0	0	0	
walking	0.1	0.57	0.12	0.12	0	0.01	0.01	0.04	0	
spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06	
running	0.01	0.07	0.07	0.79	0.04	0	0	0.01	0	
jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09	
eating	0.19	0	0	0	0	0.76	0.04	0	0.01	
climbing	0.06	0.01	0	0	0.03	0	0.90	0	0	
crawling	0.20	0.03	0	0.06	0.01	0	0 0.01 0.69		0	
flying	0.19	0.03	0.01	0	0.01	0.01	0.03	0.03	0.70	



Human\_Walking



Human\_Jumping



Animal\_Walking Animal Jumping Note: images here are videos in the original slides.

# How does the performance vary with density of the supervoxels?



#### **Study Results: Performance by Level**

- Bar figures are the response time.
  - X-axis: Time at the half-frame-rate.
  - Y-axis: density of responses.
  - Blue bars: simple histogram.
  - Red curve: Gaussian kernel density estimate.



#### **Study Results: Performance by Level**

- Correct action matches:
  - Response distributions are early equivalent.
  - Heavily weighted toward the shorter end of X-axis.
- If the participant knows the answer then typically knows it quickly.



#### **Study Results: Performance by Level**

- Incorrect action matches:
  - Different patterns.
  - Fine videos peaked at about eight seconds.
- Participant watched the whole video and still got the wrong action perception.



#### Study <u>Results: Performance by Level</u>

 Information in finer details are unlikely to be needed when performing the task.





Seconds





Fine





# How does the performance vary with actor (human versus animal)?

## **Study Results: Performance by Actor**



- In general, human action has better match.
- For speed (one peak vs. multiple peaks)
  - Greater variation in appearance of animals.
- Human activity is easier to perceive than animal.
  - A correlation between knowing the actor and recognizing the action correctly.

#### **Study Results: Performance by Actor**







#### **Study Results: Performance by Actor**







# How does the performance vary with background (static versus moving)?

#### **Study Results: Performance by Background**



- Static Background:
  - The dominant actor is more easily picked out.
- Moving Background:
  - The flat curve suggests the response time for a single video highly depends on the specific background within that video.

#### **Study Results: Performance by Background**









#### **Study Results: Performance by Background**









#### How does response time vary with action?

#### **Study Results: Speed by Action**

	Orașelia a		un	wl	sp	rn	jm	ea	cl	cr	fl
		unknown	0	0	0	0	0	0	0	0	0
		walking	0.11	0.57	0.12	0.12	0	0.01	0.01	0.04	0
		spinning	0.15	0.06	0.65	0.03	0	0	0.01	0.04	0.06
2 4 6 8 10		running	0.01	0.07	0.07	0.79	().04	0	0	0.01	0
Jumping	Running	jumping	0.19	0.01	0.04	0.09	0.57	0	0	0.01	0.09
		eating	0.19	0	0	0	0	0.76	0.04	0	0.01
		climbing	0.06	0.01	0	0	0.03	0	0.90	0	0
		crawling	0.20	0.03	0	0.06	0.01	0	0.01	0.69	0
	2 4 6 8 10 Coconde	flying	0.19	0.03	0.01	0	0.01	0.01	0.03	0.03	0.70

#### **Action Discrimination**

 Actions whose semantics have been strongly retained are generally responded to more quickly.



#### **Study Results: Speed by Action**

- Actions whose semantics have been strongly retained are generally responded to more quickly.
- Unusual actions take more time to get a response.

## **Summary of Study**

- Segmentation hierarchies generate rich decompositions of the video content.
- They compress the signal significantly, but does enough semantic information retained to discriminate actor and action?
- Yes! 82% accuracy on actor and 70% on act.
- Performance increases with coarseness of the signal.
- Performance for human actors is better than animals.
- Performance for a static background is better than a moving background.
- Future Work:
  - Semantic ambiguity and large study cohort.
  - More Supervoxel Algorithms: SWA etc.
  - Action recognition based on Supervoxels.

#### Video Understanding; What?



Method: Laptev. "On Space-Time Interest Points." IJCV 2005.



Method: Wang et al. "Action Recognition by Dense Trajectories." CVPR 2011.

#### Alas, what makes such a good representation?



Method: Supervoxel segment boundaries. Xu and Corso CVPR 2012.

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#### **Segmentation: Toward a Representation with Rich Semantics?**









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## Thank you!