

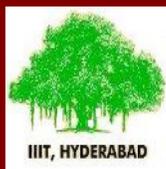
Fine Grain Annotation of Cricket Videos

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<http://cvit.iiit.ac.in/projects/cricketannotation>



Problem: Fine Grain Annotation of Cricket Videos

Can we Detect and Generate detailed commentary from Cricket Videos?



Batsman: Gambhir

Description : "He pulls it off from outside off stump and just manages to clear the deep square leg rope"

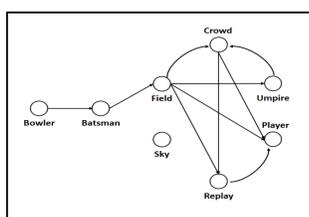
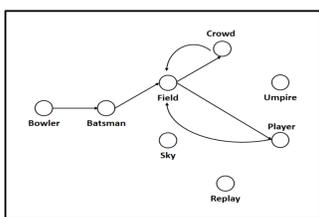
Challenges

Semantic description of actions/activities is difficult using current visual-recognition techniques alone. Typical actions of interest are *tiny* in space-time (90x150 pixels across ~35 frames(1.2 seconds))

Goals

Annotate Cricket Videos with Semantic descriptions at a fine grain spatio-temporal scale. Leverage *parallel* information in the form of online text commentaries, to augment visual recognition. Label actions at shot-level with semantic descriptions of actions and activities.

Scene Segmentation



one-run model includes only a few states

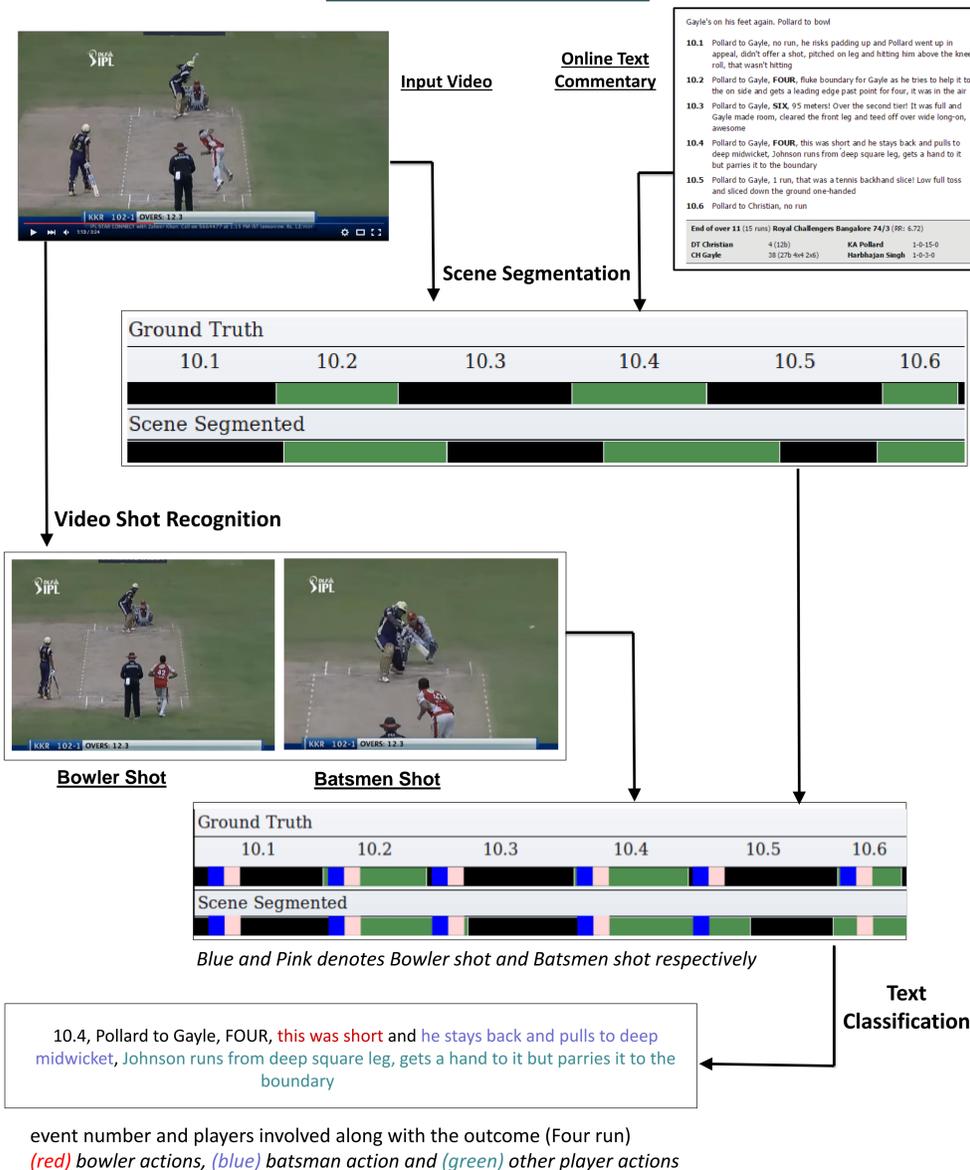
four-run model involves a variety of visuals

State transition diagrams for two scene categories: (left) One Run and (right) Four-Runs

Video Shot Recognition



Solution Overview



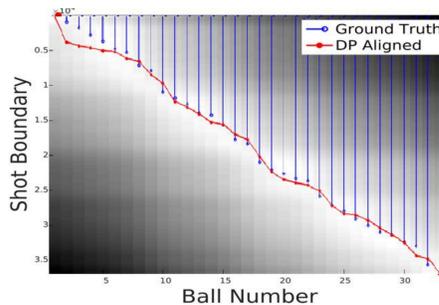
Experimental Results

Datasets

| Name | Matches | No. of Phrases | Role |
|-------------------|-------------------|---|------------------------|
| IPL Video Dataset | 4 Matches (20Hrs) | 960 Phrases | Video Shot Recognition |
| CricInfo Dataset | 300 Matches | 1500 Bowler Phrases 6000 Batsmen Phrases | Text Classification |



Dataset is collected from the YouTube channel of the Indian Premier League (IPL) Tournament. Commentary Dataset is collected from CricInfo.com



Results of Scene Segmentation depicted over five-overs. The background of the image is the cost function of the scene segmentation.

Red Line - Optimal backtrack path
 Red Circles - Inferred scene boundaries
 Blue Lines - GT

Phrase Prediction

| R | Bowler Shot | Batsman Shot |
|----|--------------|--------------|
| 2 | 22.15 | 39.4 |
| 4 | 43.37 | 47.6 |
| 6 | 69.09 | 69.6 |
| 8 | 79.94 | 80.8 |
| 10 | 87.87 | 88.95 |

R is Neighborhood to search for correct Shot

Video Shot recognition

| Kernel + CRF | Vocab: 300 | Vocab: 1000 |
|--------------|------------|--------------|
| Linear | 78.02 | 82.25 |
| Polynomial | 80.15 | 81.16 |
| RBF | 81 | 82.15 |
| Sigmoid | 77.88 | 80.53 |

Text Classifier

| Method | SVM |
|--------------------|-------|
| Bag of Words (SVM) | 89.09 |

Calculated using 2 Fold cross Validation



Summary and Conclusion

- Our solution enables rich semantic annotation of Cricket Videos at a fine temporal scale.
- Circumvention of technical challenges in visual recognition by utilizing online text-commentaries.
- We obtain high annotation accuracy as evaluated over a large video collection.
- In future, our labelled dataset could be used to learn classifiers for fine-grain activity recognition.

Other Applications



Highlight Generation
Automatic Highlight generation from a Match



Sports
Event Detection in Tennis Videos



Activity learning
Learning classifier for Cover drive and so on.



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