

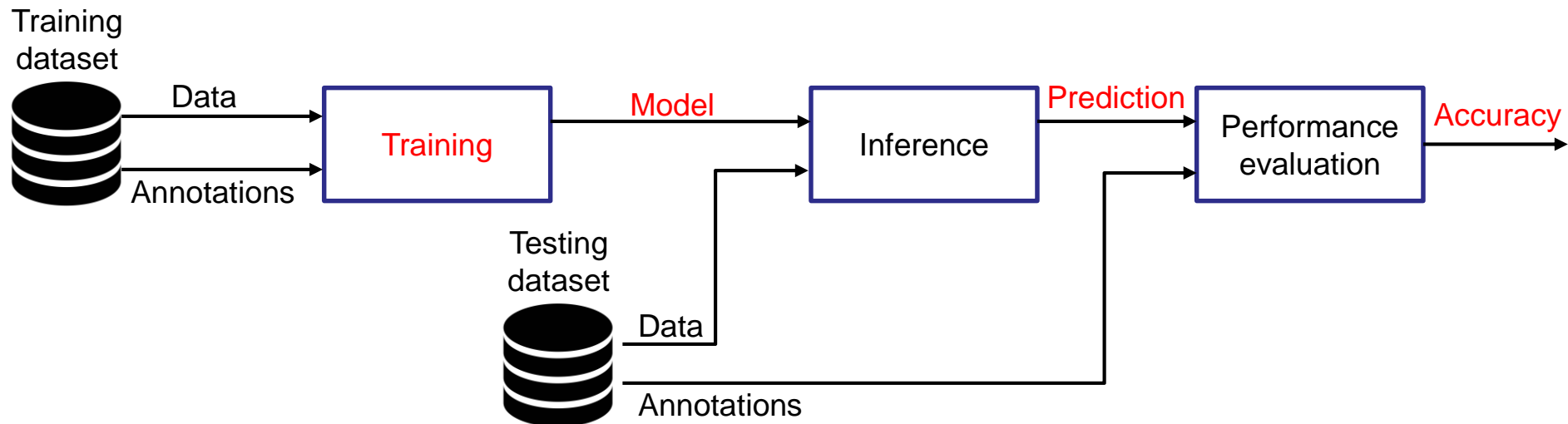
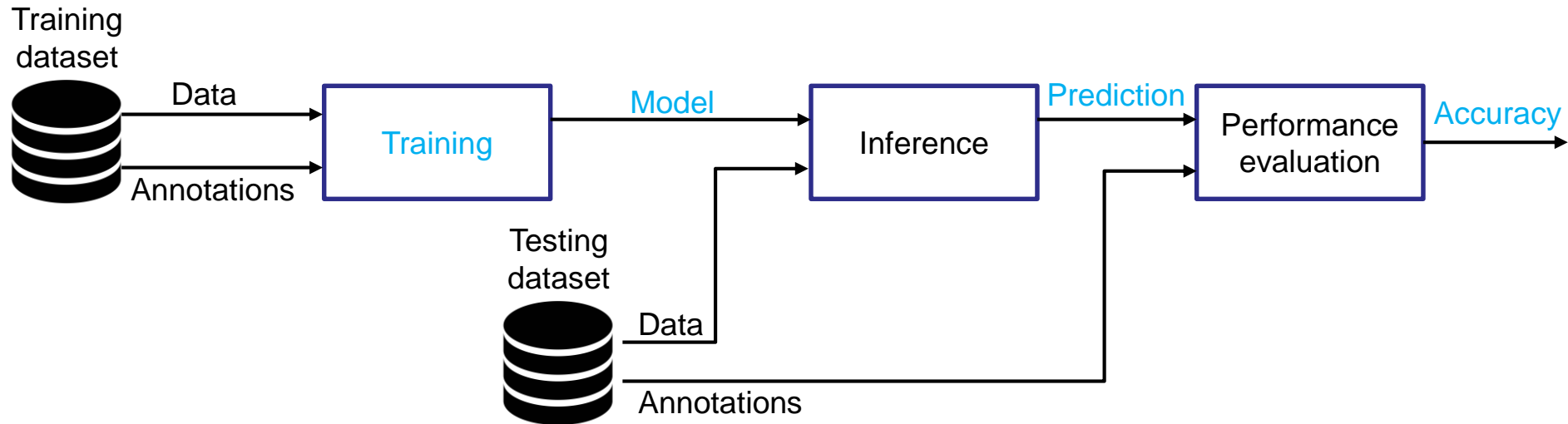
Confidence intervals for tracking performance scores

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Performance evaluation (ranking)



Annotations

[1] Vondrick, Patterson and Ramanan, "Efficiently scaling up crowdsourced video annotations", IJCV 2013

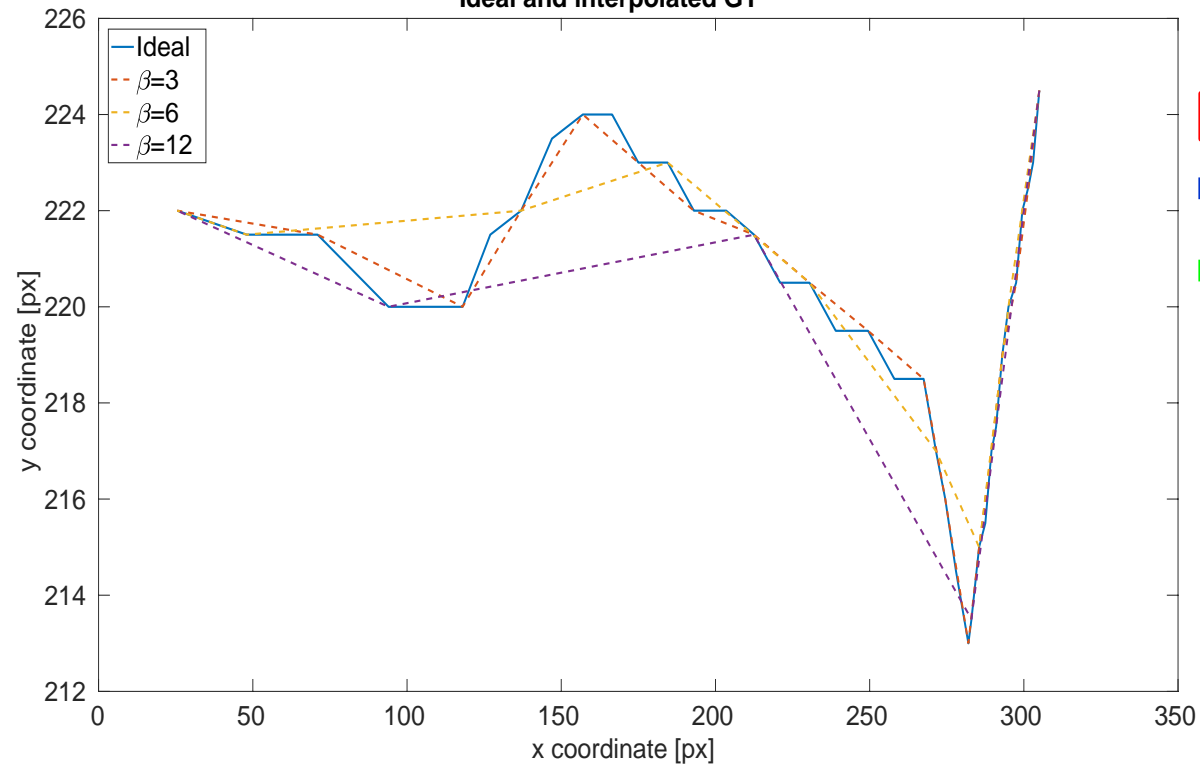
- Performance evaluation requires manual annotations
- Manual annotations are [1]:
 - Tedious
 - Expensive (time and economically wise)
 - (Potentially) inaccurate
 - (Potentially) unfeasible to retrieve
- Inaccurate annotations produce inaccurate evaluations






Tracking: Interpolated annotations



Ideal and interpolated GT



-  Manual (ideal) annotation
-  Interpolated annotation
-  Publicly available annotation (uses linear interpolation)

β : interpolation factor

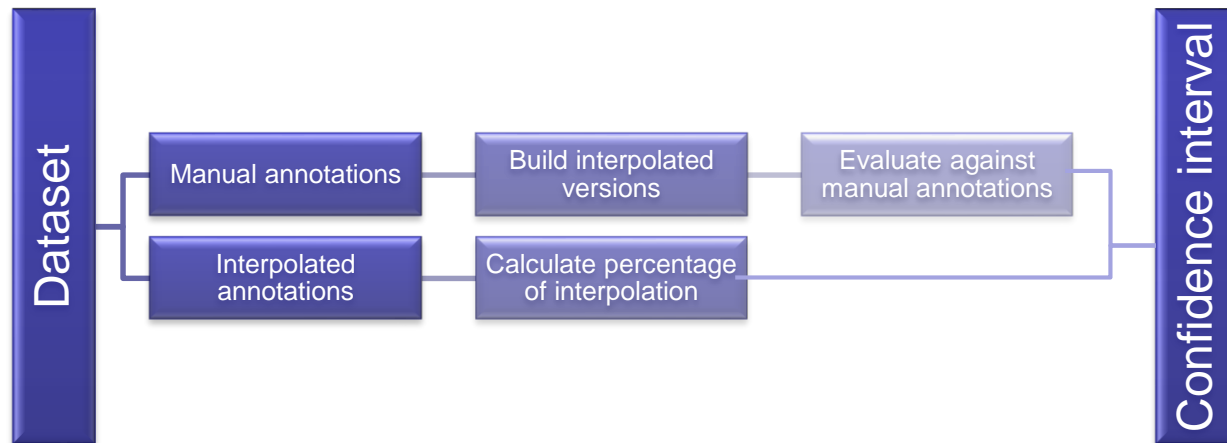
Tracking: Datasets

- Most public datasets use linear interpolation for their annotations

Dataset	Tool	Linear Interpolation
CAVIAR	CaviarGui	
TUD	NA	✓
ETH	NA	✓
PETS09	NA	✓
KITTI	MT	NA
i-LIDS	ViPER	✓
MOTB15	VATIC	✓
MOTB16	NA	✓

Confidence intervals

- Annotation inaccuracies should be taken into account
- Estimate uncertainty in annotations for a given dataset:
 - Unknown interpolation
 - Without doing further annotations
 - Estimate confidence interval



Separating manual/interpolated annotations

$$\mathbb{Z} = \{\mathbf{z}_k^\lambda : \lambda = 1 \dots \Lambda; k = 0 \dots K_\lambda - 1\}$$

$$\mathbf{z}_k^\lambda = (u, v, w, h)$$

λ : target identity

k : time

$$\mathbb{Z} = \tilde{\mathbb{Z}} \cup \hat{\mathbb{Z}}, \quad \tilde{\mathbb{Z}} \cap \hat{\mathbb{Z}} = \emptyset$$

\mathbb{Z} : public dataset

$\tilde{\mathbb{Z}}$: manually annotated subset

$\hat{\mathbb{Z}}$: linearly interpolated subset ← Second derivatives equal to zero



- Percentage of interpolation in the dataset: $\frac{|\hat{\mathbb{Z}}|}{|\mathbb{Z}|}$

Building interpolated versions

- Generate interpolated versions of the manual subset through a decimation-interpolation procedure

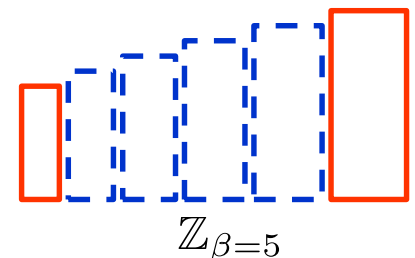
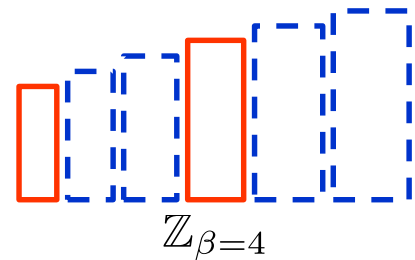
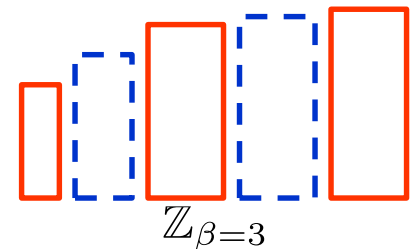
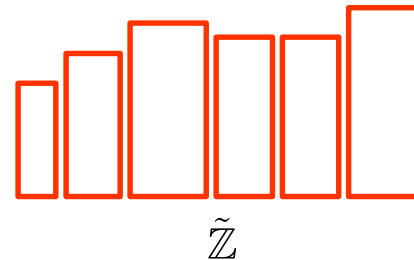
$$\mathbb{Z} \longrightarrow \begin{cases} \tilde{\mathbb{Z}} \\ \hat{\mathbb{Z}} \end{cases} \longrightarrow \mathbb{Z}_\beta$$

$\beta \in \{3, 6, 9, 12\}$: interpolation factor

\mathbb{Z} : public dataset

$\tilde{\mathbb{Z}}$: manually annotated subset

$\hat{\mathbb{Z}}$: linearly interpolated subset



Confidence intervals

- Compare interpolated version against manual version

$$\alpha_{s,\beta} = s(\tilde{\mathbb{Z}}, \mathbb{Z}_\beta)$$

$$s(\cdot, \cdot) = 100 - MOTA$$

$$s(\cdot, \cdot) = 100 - MOTP$$

$$MOTA = 1 - \frac{1}{N} \sum_{\lambda=0}^{\Lambda} \sum_{k=0}^{\tilde{K}_\lambda - 1} (FN_k^\lambda + FP_k^\lambda + IDSW_k^\lambda)$$

$$MOTP = \frac{1}{N} \sum_{\lambda=0}^{\Lambda} \sum_{k=0}^{\tilde{K}'_\lambda - 1} \frac{\tilde{\mathbf{z}}_k^\lambda \cap \mathbf{z}_{k,\beta}^\lambda}{\tilde{\mathbf{z}}_k^\lambda \cup \mathbf{z}_{k,\beta}^\lambda}$$

\mathbb{Z} : public dataset

$\tilde{\mathbb{Z}}$: manually annotated subset

$\hat{\mathbb{Z}}$: linearly interpolated subset

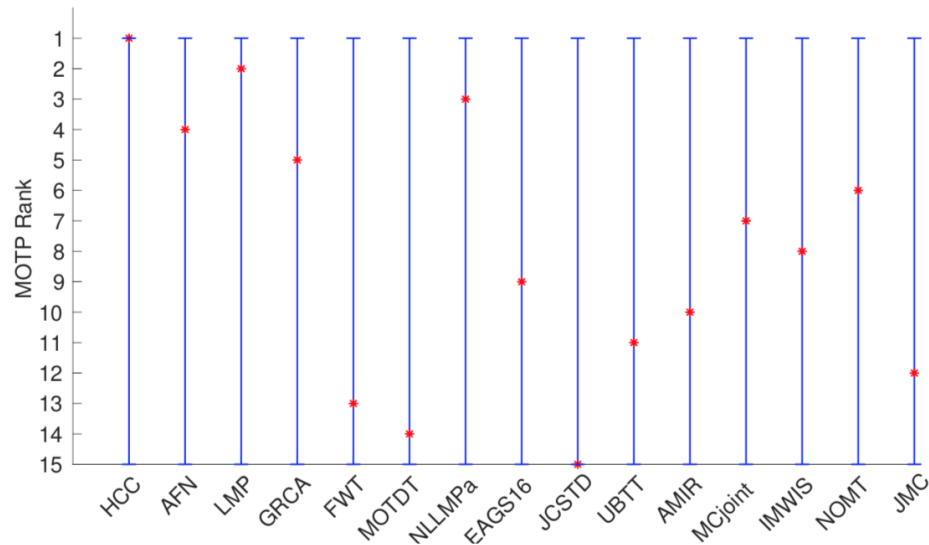
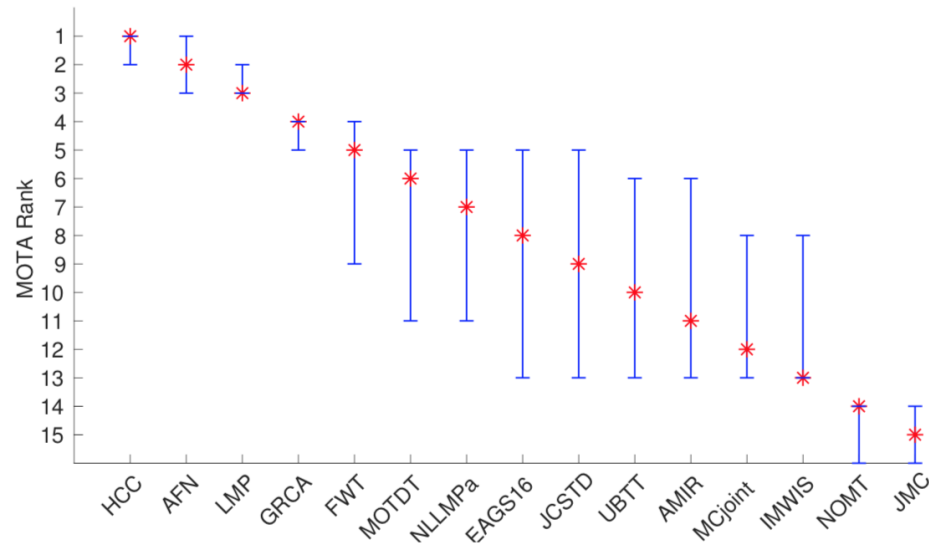
N : # annotations

β : interpolation factor

Results on MOTB16

- 39.7% annotations are generated through linear interpolation
- MOTA confidence: 0.22
- MOTP confidence 3.14

Impact on MOTB16



Conclusion

- Interpolation:
 - Makes possible to annotate large scale datasets
 - Makes possible to annotate not visible objects
 - Introduces inaccuracies
- Confidence interval:
 - Allows to take into account annotation inaccuracies
 - From an already annotated dataset
 - With unknown annotation policy
 - Without the need of doing further annotations
- Ranking of methods should be redesign considering confidence intervals

Q&A

