Performance analysis of interest point detection/matching on shiny and non-textured surfaces

A case study on aircraft engine borescope inspection videos

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Background

- > Automated damage assessment in borescope inspection videos (See Figure 1)
- > 3D models rely on interest point detection \rightarrow good interest point detection / matching likely results in good 3D models
- Related work analyses algorithms in different environments \rightarrow still unclear how these perform in these environments



Figure 1: A frame from one of the borescope inspection videos



What interest point detection / matching algorithm performs best on shiny and non-textured surfaces as found in borescope inspection videos?



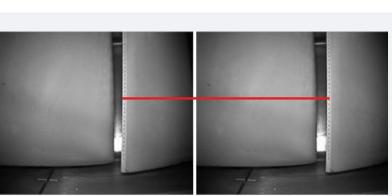


Figure 2a: Matches between two consecutive frames as computed by SIFT.

Figure 2b: An example of a relevant match



Figure 2c: An example of a irrelevant match

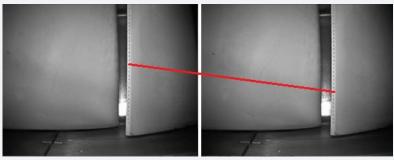


Figure 2d: An example of an incorrect match

Figure 2: Assessing individual matches on relevance and corectness

Method

- \succ Compare both recent neural network and traditional approaches \rightarrow SIFT, SuperGlue and LoFTR
- > Evaluated on video segments of various borescope inspection videos
- > Algorithms are assessed quantitatively by various metrics
 - Manual
 - Automated using RANSAC
- > Qualitative assessment using SfM

ults	Algorithm	Avg # matches	% of irrelevant matches	% of correct matches	% of incorrect matches	Table 1: results of manual assessment on the video found in figure 1	
Res	SIFT	66.05	45	45	10		
Ð	SuperGlue	555.89	37	62	1.0		
	LoFTR	5205.60	51	49	0.0		
	Algorithm	Avg # matches	Avg #	% of	% of	Table 2: results of	
		materies	correct matches	correct matches	incorrect matches	automated assessment on the	•
	SIFT	63.40					•
	SIFT SuperGlue	63.40	matches	matches	matches	assessment on the	•
		materies				assessment on the	-





Figure 3: 3D model reconstructed with SfM using SIFT

Figure 4: 3D model reconstructed with SfM using SuperGlue

- > LoFTR detects more matches than SIFT and SuperGlue
- > LoFTR has the lowest number of incorrect matches
- > Qualitative results show that SuperGlue performs best with SfM

Conclusion

- > Neural network based approaches outperform traditional
- > SuperGlue performing best in practice (SfM), LoFTR best according to metrics
- > Test influence of different parameters