<u>What is Surgical Phase Recognition?</u> Predict what surgical phase is occurring at each frame in the surgical videos. t+1 <u>Why Multi-Stage Architecture?</u> The imperfect predictions can be further refined. Surgical video contents contain rich temporal patterns. Ground Ground Truth y Truth y Videos Visual Predictor Refinement $\longrightarrow \hat{y}_r$ Stage Features Stage Videos

End-to-End not Work in Multi-Stage

1. The inputs of the refinement stage during training and inference are different. 2. The limited size of current datasets.

Our Solution

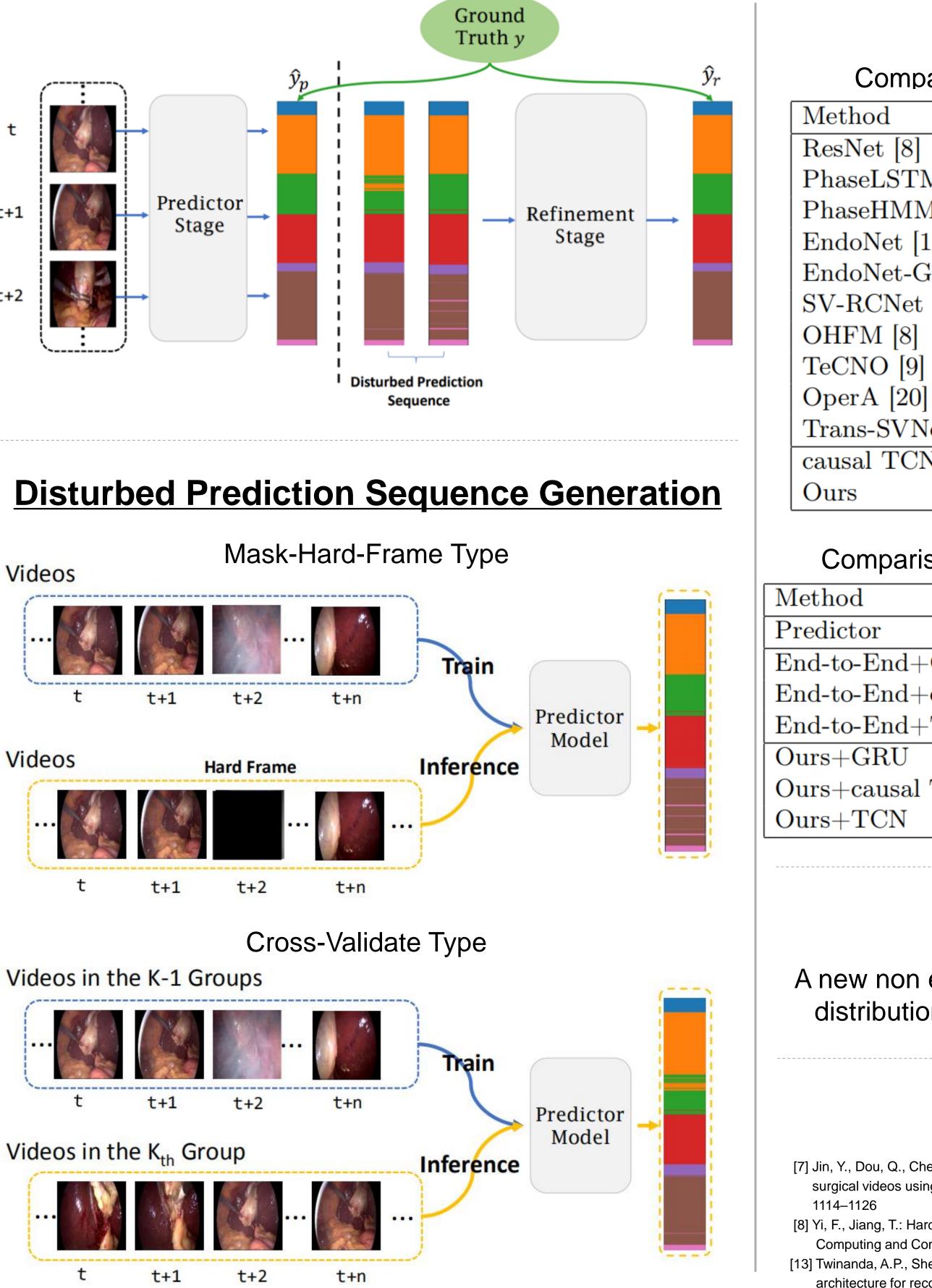
Train predictor stage and refinement stage separately. Design two types of training sequences to simulate the real output of the predictor during inference.



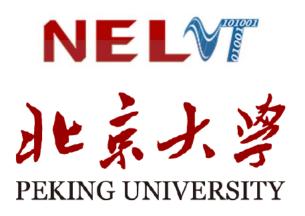
Not End-to-End: Explore Multi-Stage Architecture for **Online Surgical Phase Recognition**

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Experiments

Comparison with SOTA on Cholec80 dataset

	Acc	JACC	Rec
	78.3 ± 7.7	52.2 ± 15.0	-
M [25]	80.7 ± 12.9	$64.4{\pm}10.0$	-
M [25]	71.1 ± 20.3	$62.4{\pm}10.4$	-
13]	81.7 ± 4.2	-	79.6 ± 7.9
GTbin [13]	81.9 ± 4.4	-	80.0 ± 6.7
[7]	85.3 ± 7.3	-	83.5 ± 7.5
	$87.0 {\pm} 6.3$	66.7 ± 12.8	-
	$88.6 {\pm} 2.7$	-	$85.2{\pm}10.6$
]	$85.8 {\pm} 1.0$	-	87.7 ± 0.7
Vet [21]	90.3 ± 7.1	$\textbf{79.3}{\pm\textbf{6.6}}$	$88.8{\pm}7.4$
N	88.8 ± 6.3	73.2 ± 9.8	84.9 ± 7.2
	$\textbf{92.0}{\pm\textbf{5.3}}$	77.1 ± 11.5	$87.0{\pm}7.3$

Comparison with End-to-End on Cholec80 dataset

	Acc	JACC	Rec
	88.8 ± 6.3	73.2 ± 9.8	84.9 ± 7.2
-GRU	87.1 ± 7.8	69.7 ± 12.6	83.2 ± 9.4
-causal TCN	87.7 ± 6.3	77.7 ± 11.2	84.3 ± 6.3
-TCN	$89.8 {\pm} 6.6$	$75.8 {\pm} 8.4$	87.4 ± 7.5
	$90.8 {\pm} 7.0$	75.5 ± 11.1	85.6 ± 10.0
TCN	91.0 ± 5.2	74.2 ± 11.8	84.1 ± 9.6
	$\textbf{92.8}{\pm\textbf{5.0}}$	$\textbf{78.7}{\pm\textbf{9.4}}$	$\textbf{87.5}{\pm}\textbf{8.3}$

Conclusion

A new non end-to-end training strategy to minimize the distribution gap between the training and inference.

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