



# Can Image Enhancement be Beneficial to Find Smoke Images in Laparoscopic Surgery?

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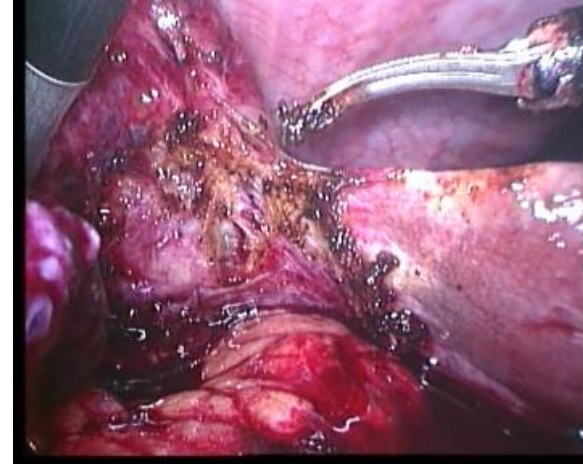
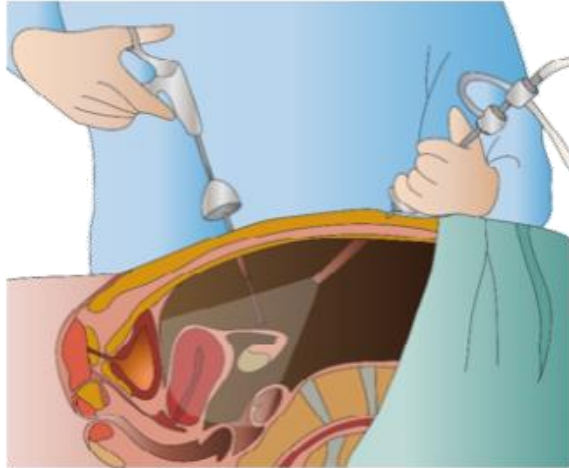
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\*Denotes equal contributions and listed in alphabetical order.

# Motivation

## Laparoscopic Surgery



- Smoke degrades laparoscopic video quality.
  - Influences surgeon's visibility
  - Influences the performance of computer vision based navigation systems
  - May be harmful for surgeons and patients

# Motivation

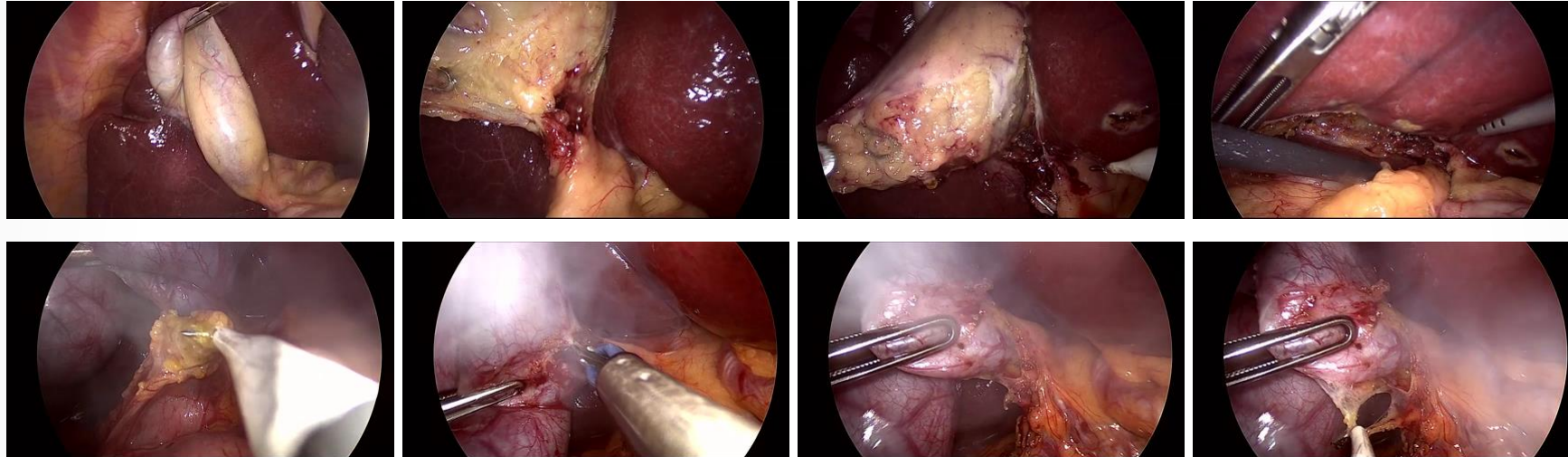
- Desmoking technique
  - Computer vision algorithms
  - Smoke evacuation techniques



- **When to start to remove smoke?**

# Motivation

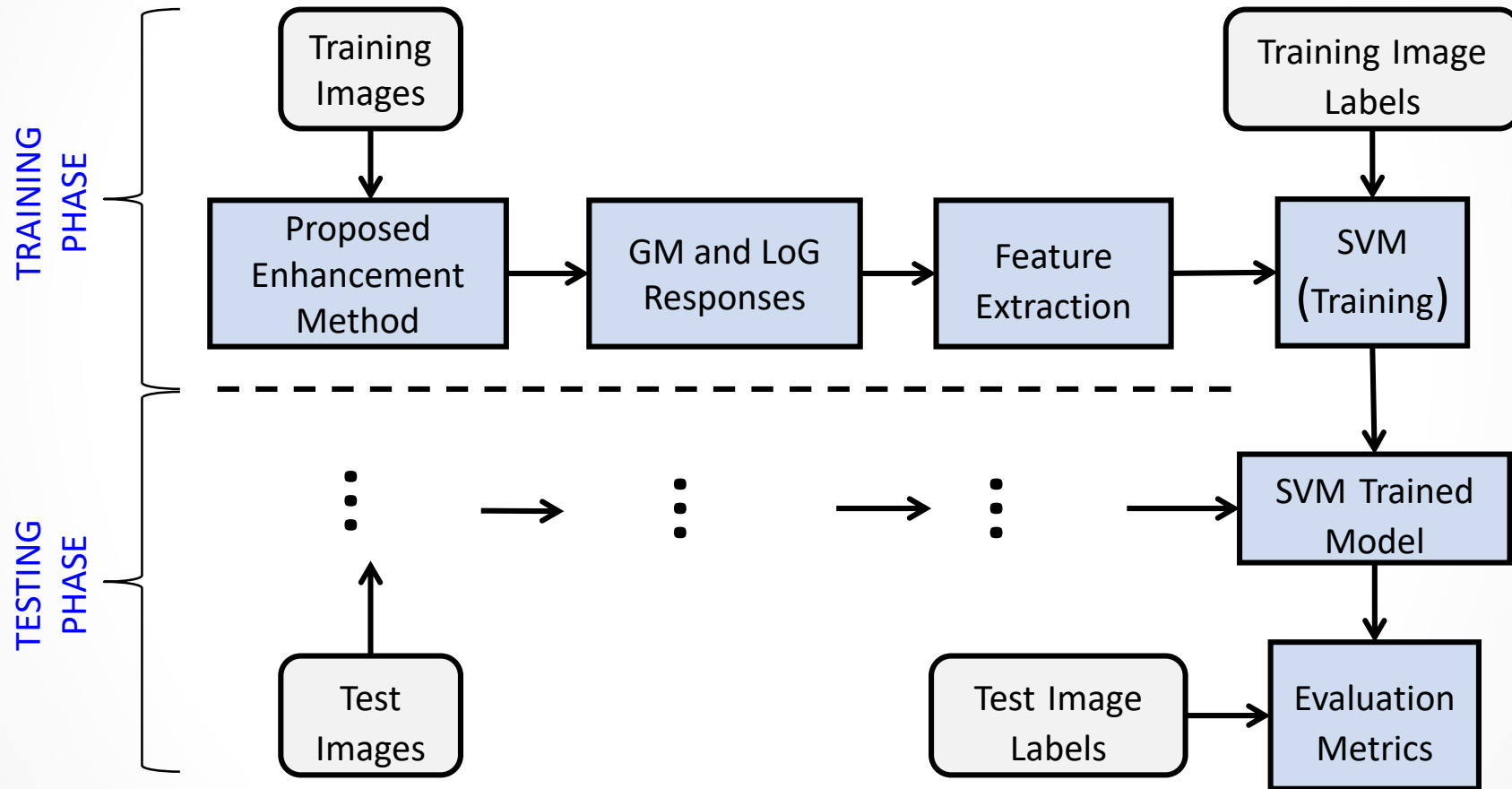
- Smoke/non smoke images classification



- Goal:
  - Enhance the images for improved classification

# Main idea

- Pipeline



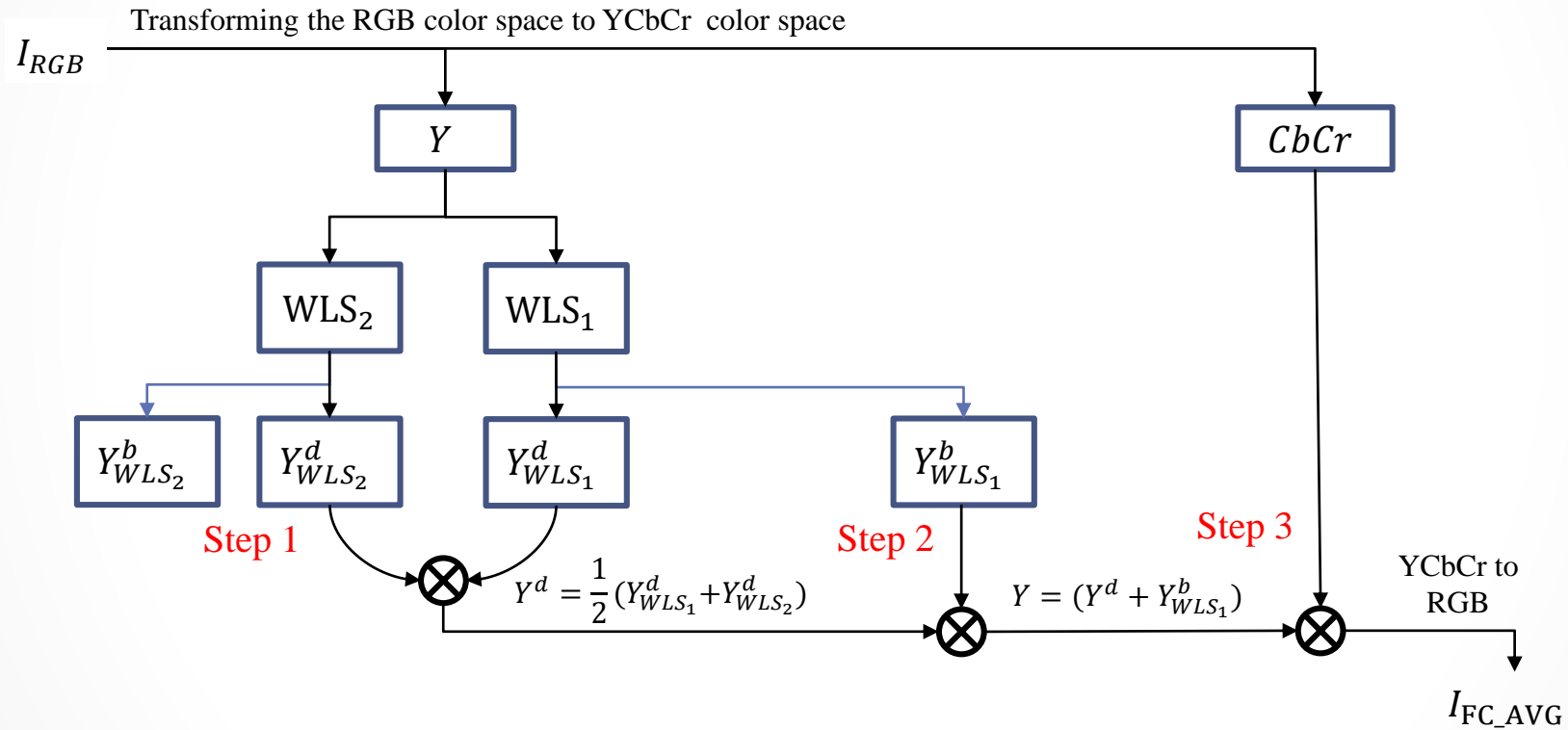
# Proposed enhancement method

- Weighted least squares optimization framework (WLS)

$$g^{Filtered} = F_{\lambda}(g) = (I + \lambda L_g)^{-1} g$$

- Decomposition of a base-layer and detail-layers.
  - Base layer =  $g^{Filtered}$
  - Detail layer =  $g - g^{Filtered}$

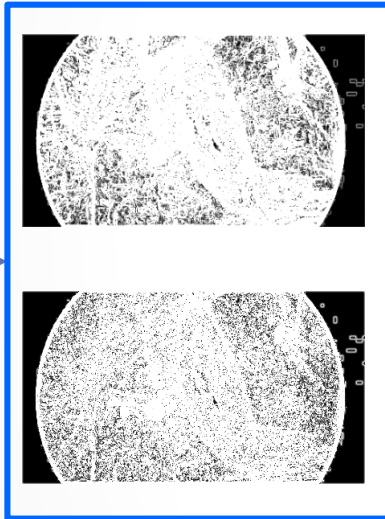
# Proposed enhancement method



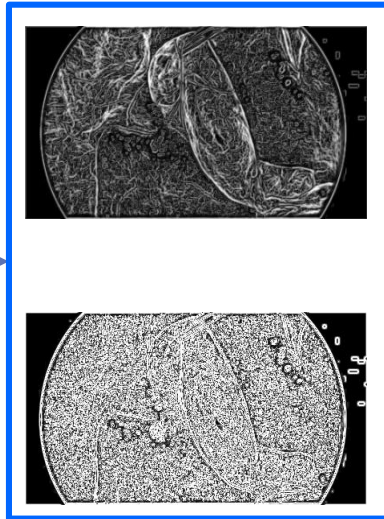
# Feature extraction

- Gradient Magnitude (GM) features
- Laplacian of Gaussian (LoG) features

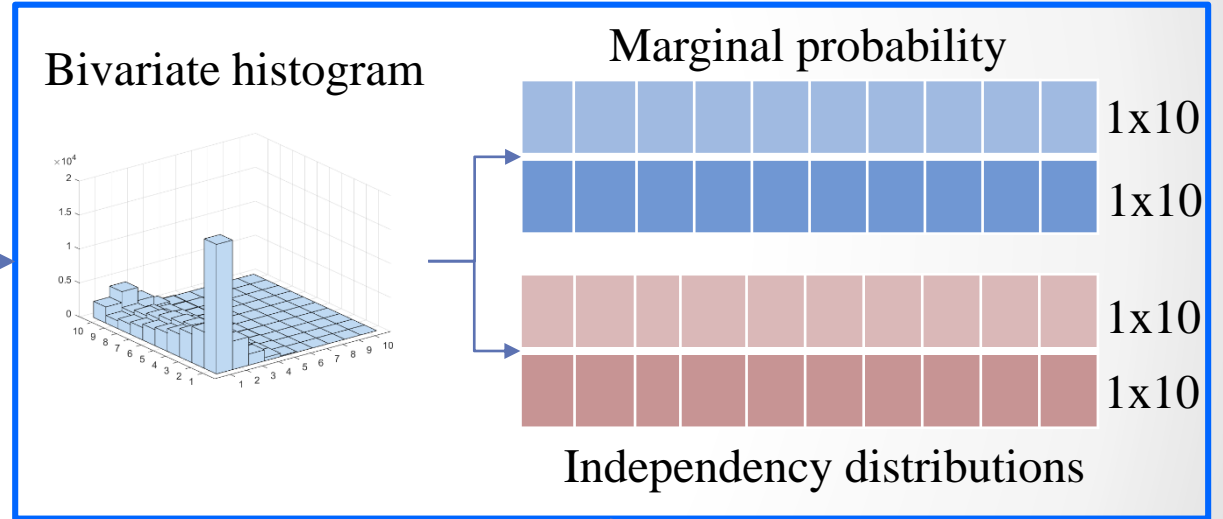
GM and LoG maps



Joint normalization



Statistical feature description



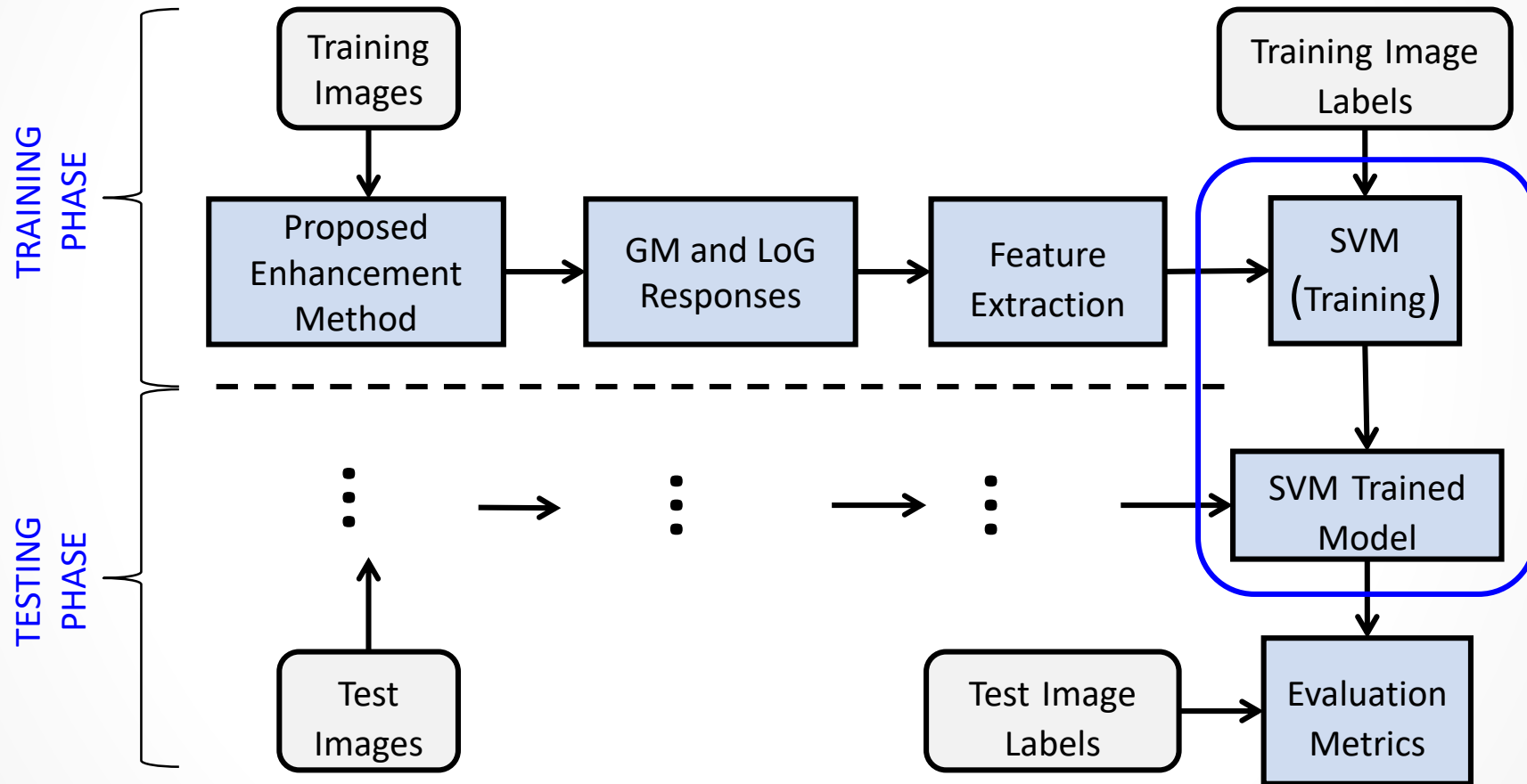
A feature vector of size 1x40





# Classifier

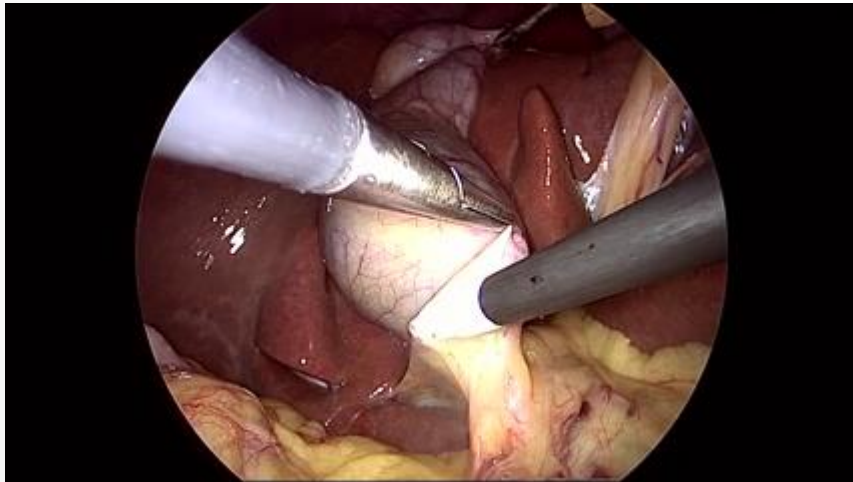
- Pipeline



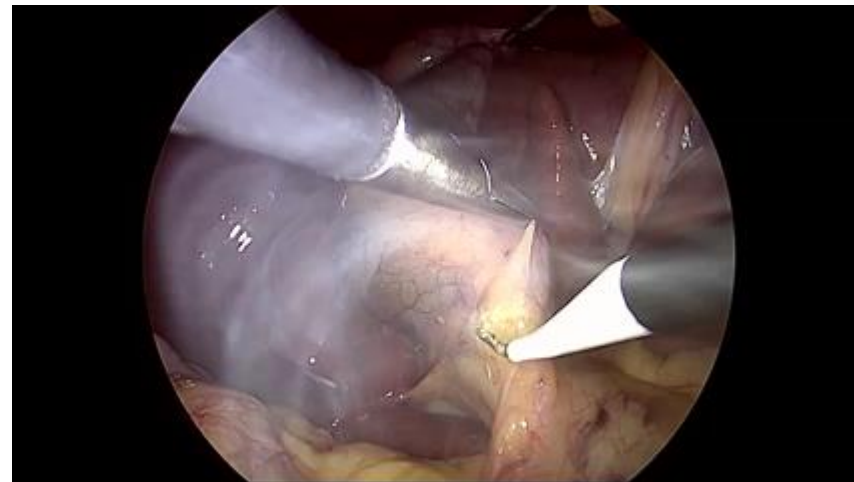
# Experiments

- Cholec80 dataset: cholecystectomy surgeries manually labeled with smoke/non-smoke image sequence
  - Training: 4,381 images obtained from three videos
  - Testing: 10,653 images obtained from nine videos

0 - -non smoke



1 - -smoke



# Experiments

- How to evaluate the classification result ?
  - Accuracy
    - The higher the better

$$\begin{aligned} \text{Accuracy} &= \frac{TP + TN}{TP + TN + FP + FN} \\ &= \frac{\text{The number of correct classified smoke / non smoke images}}{\text{Total number of testing images}} \end{aligned}$$

- F1-Score
  - The higher the better

$$F1 - \text{Score} = 2 \cdot \frac{\text{precision} \cdot \text{recall}}{\text{precision} + \text{recall}}$$

# Experiments

- Comparison with other enhancement methods

Method	Accuracy	F1-Score
RGB	0.60	0.60
IMSHARP	0.58	0.58
BF	0.60	0.59
GF	0.60	0.59
WLS	0.60	0.59
BFWLS_AVG	0.57	0.56
FC_MAX( <b>Ours</b> )	0.60	0.59
FC_AVG( <b>Ours</b> )	<b>0.64</b>	<b>0.64</b>

Tab. 1: Comparison with the baseline RGB images and other enhancement methods

# Experiments

- Comparison with other enhancement methods

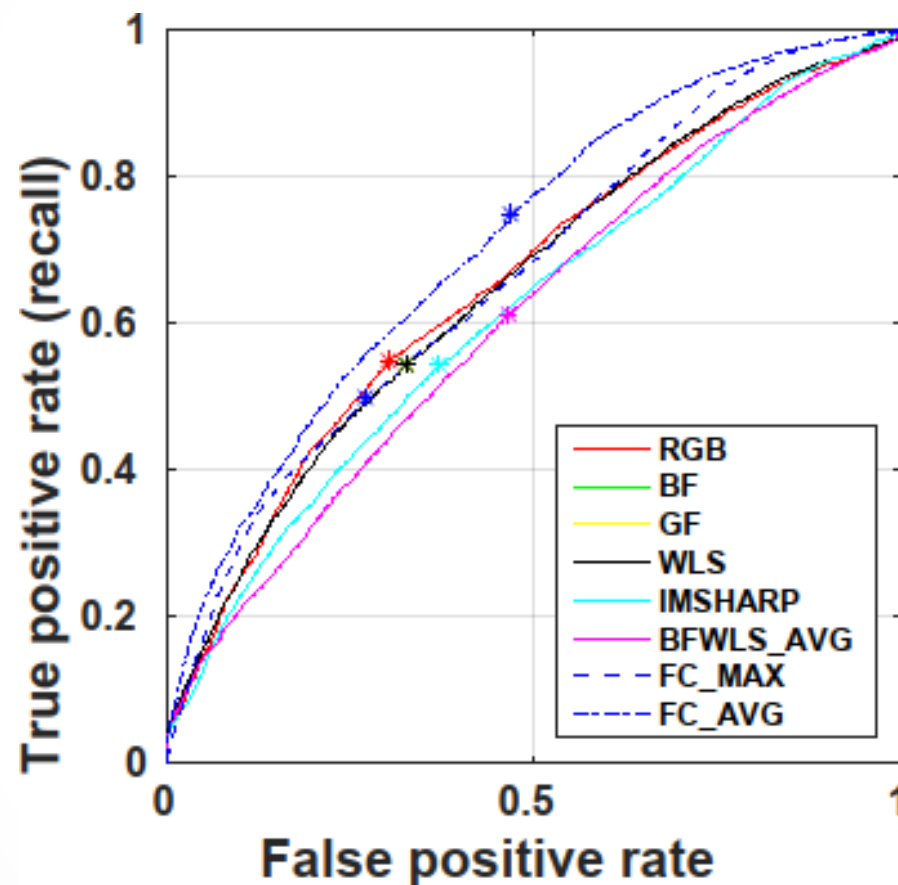


Fig. 1: The ROC curves for smoke/non-smoke classification task. \* denotes the EER when the false accept rate is equal to the false reject rate.

# Experiments

- Comparison with the saturation histogram based classification methodologies

Method	Accuracy	F1-Score
SPA	0.63	0.58
SAN	0.63	0.59
FC_AVG(Ours)	0.64	0.64

Tab. 2: Comparison with the saturation histogram based classification methodologies Saturation Analysis (SAN) and Saturation Peak Analysis (SPA)

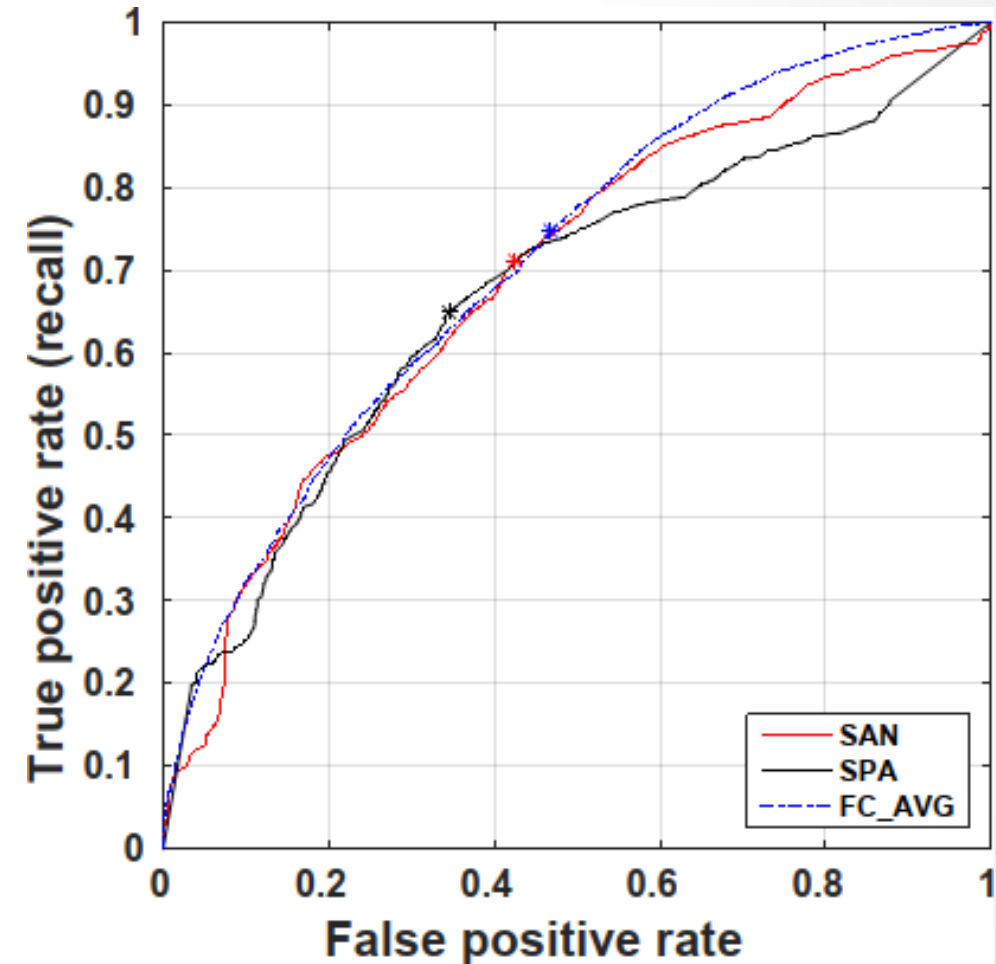


Fig. 2: The ROC curves for the three methods

# Conclusion & Discussion

- Propose a method to enhance the informative features
- Combine the enhancement method with a SVM based classification method
- Improved smoke/non-smoke classification results
  - Better result compare to the baseline RGB images
  - Better result compare to he saturation histogram based classification methodologies.
- Future work
  - Employ CNN architecture for the classification task



# Thank you, any questions ?

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