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# Laser Damage Image Pre-processing

## Based on Total Variation

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# Outline

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-  Background
-  Motivation
-  Damage Image Pre-processing
-  Experiments
-  Conclusion



# 1 Background

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## Laser Damage

- Because of reasons such as self-focusing, laser-induced damages are likely to happen to the optics.
- Damages need to be inspected and tracked upon being initially produced, otherwise the optics would be damaged heavily and become non-repairable.

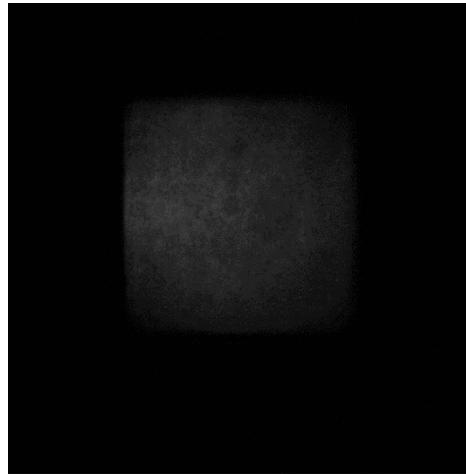


# 1 Background

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## Laser Optics Damage Inspection

- Laser-induced defects or flaws on the optics are presented in images acquired by specific charge coupled devices (CCDs).



Example image acquired by the CCD

- Laser optics damage inspection relies heavily on the image analysis to identify the damage defects or flaws.

# 1 Background

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## Related Researches

- To develop the optic illumination techniques.
- To improve the damage image analysis and defect identification technology.

## Challenges of Damage Defect Identification

- The size of defects is quite tiny compared to the image.
- The grey value of different image areas is different because of the uneven distribution of illumination.
- The low-light-level property of damage images acquired makes defects blurred into the backgrounds.



## 2 Motivation

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- Due to the challenges listed above, the accuracy of damage defect identification is not satisfying: **high false alarm** rates and **high missing** rates.
- One question: ***Can we develop algorithms to pre-process these laser damage images acquired by CCDs, and to improve the identification of defect points?***



# 3 Damage Image Pre-processing

## Problem Description

- Given the unknown pollutions  
image pre-processing → image restoration
- Assume that:  $f = K\bar{x} + \omega$ 
  - $\bar{x} \in \mathbb{R}^{n^2}$  : the original  $n \times n$  image
  - $K \in \mathbb{R}^{n^2 \times n^2}$  : the blurring operator
  - $\omega \in \mathbb{R}^{n^2}$  : the additive noise
  - $f \in \mathbb{R}^{n^2}$  : the observation
- **Our objective:** to recover  $\bar{x}$  from  $f$



# 3 Damage Image Pre-processing

## Total Variation (TV) Based Model

- Combining TV regularization with  $l_2$  norm fidelity, we get the TV based image reconstruction model as following:

$$\min_x \sum_{i=1}^{n^2} \|D_i x\|_2 + \frac{\mu}{2} \|Kx - f\|_2^2$$

$\downarrow$  TV regularization      $\downarrow$   $l_2$  norm fidelity

- Equivalent constrained formulation:

$$\min_{x,y} \sum_i \|y_i\| + \frac{\mu}{2} \|Kx - f\|_2^2$$

s.t.  $y_i = D_i x, \quad i = 1, \dots, n^2$



# 3 Damage Image Pre-processing

## Applying ADMM to TV Based Model

- the augmented Lagrangian function :

$$\Gamma_A(x, y, \lambda) = \sum_i (\|y_i\| - \lambda_i^T (y_i - D_i x) + \frac{\beta}{2} \|y_i - D_i x\|_2^2) + \frac{\mu}{2} \|Kx - f\|_2^2.$$

- Iterative scheme:

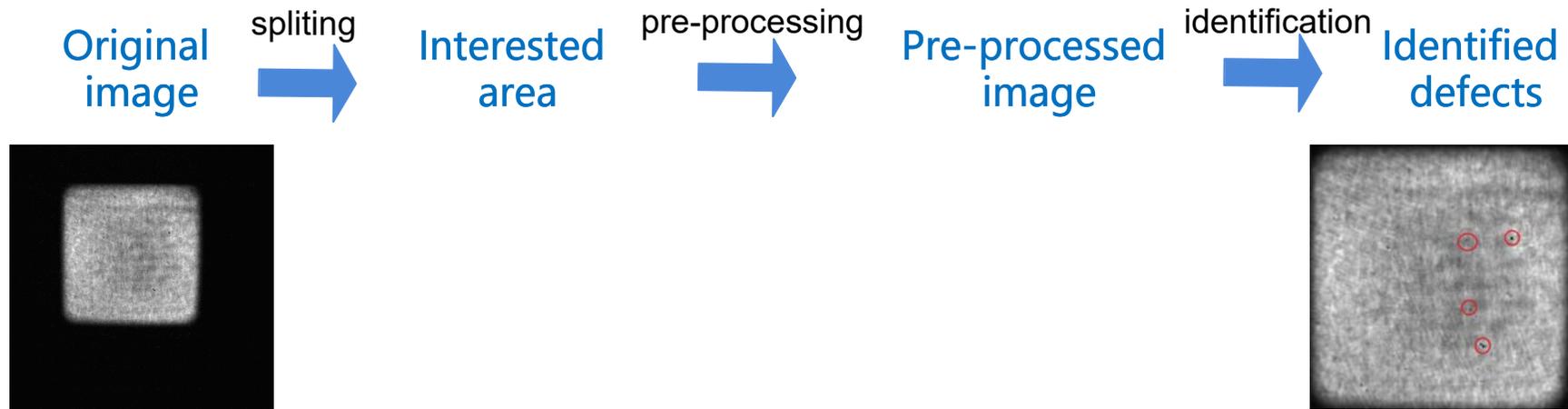
- $y^{k+1} = \arg \min_y \Gamma_A(x^k, y, \lambda^k)$
- $x^{k+1} = \arg \min_x \Gamma_A(x, y^{k+1}, \lambda^k)$
- $\lambda^{k+1} = \lambda^k - \beta(y^{k+1} - Dx^{k+1})$

- Terminate condition:  $\frac{\|x^{k+1} - x^k\|}{\max\{\|x^k\|, 1\}} < \varepsilon$



# 4 Experiments

## Procedure



## The comparison of accuracy

Method	False Alarm Rate	Missing Rate
No pre-processing	30%	15%
With pre-processing	8 %	5%

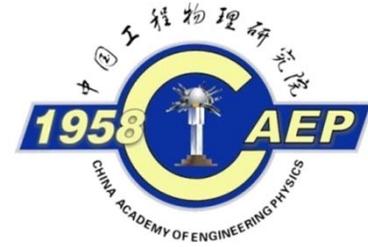


# 5 Conclusion

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- Concluding remarks
  - Optics damage inspection directly with original images shows a high false alarm rate and a high missing rate
  - Pre-processing images through reconstructing them, by utilizing the total variation (TV) based model and an alternating direction method of multipliers (ADMM ) algorithm
  - Preliminary experiments demonstrate the potential of pre-processing method: both the false alarm rate and the missing rate are reduced
- Future research
  - The optimal regularization parameter of this method varies when dealing with different images.
  - The scheme of the optimal regularization parameter selecting.





*Thanks*

*for your attention!*



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