

dnoiseNET: Deep CNN for image denoising

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Declaration of Financial Interests or Relationships

Speaker Name: Hung Do

Company Name: Canon Medical Systems USA, Inc. (formerly Toshiba Medical)

Type of Relationship: Employee

SNR in MRI

Shorter acquisition time

- Shorter breath-hold
- Less sensitive to motions



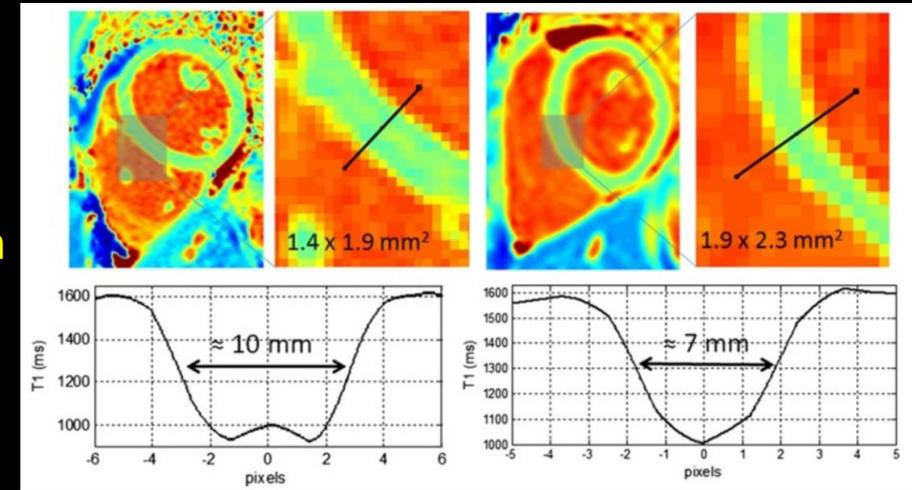
$$\text{Uncertainty} \sim \frac{1}{\text{SNR}}$$

Higher SNR

- Image quality
- Visualization
- Down-stream post-processing

Higher resolution

- Less partial volume effects

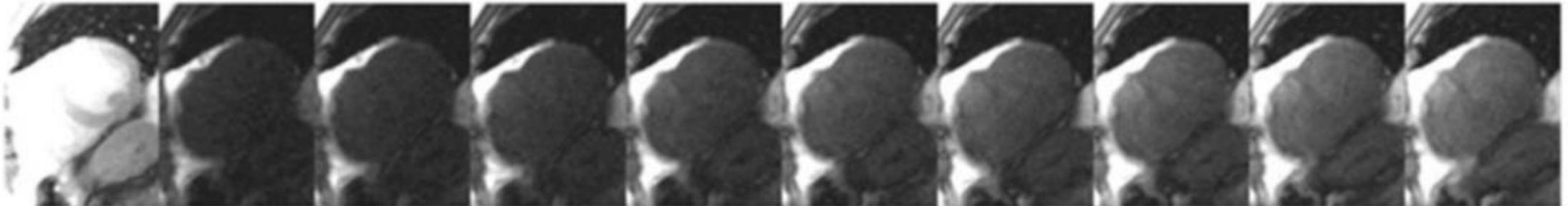
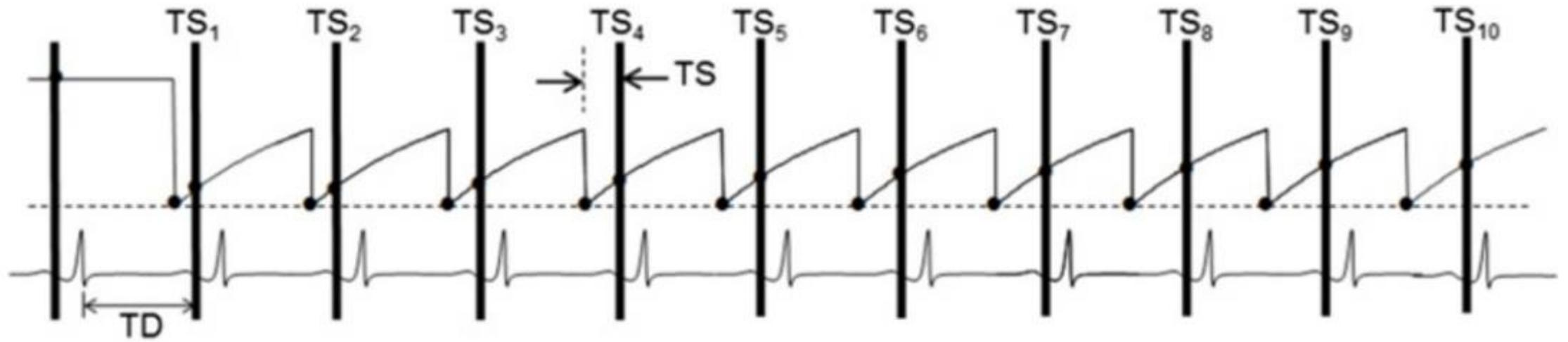


Quantitative CMR

- Relaxometry: T1, T2, T2*, T1rho
- Diffusion: DWI, DTI, IVIM
- Perfusion: ASL
- Water-Fat: PDFF

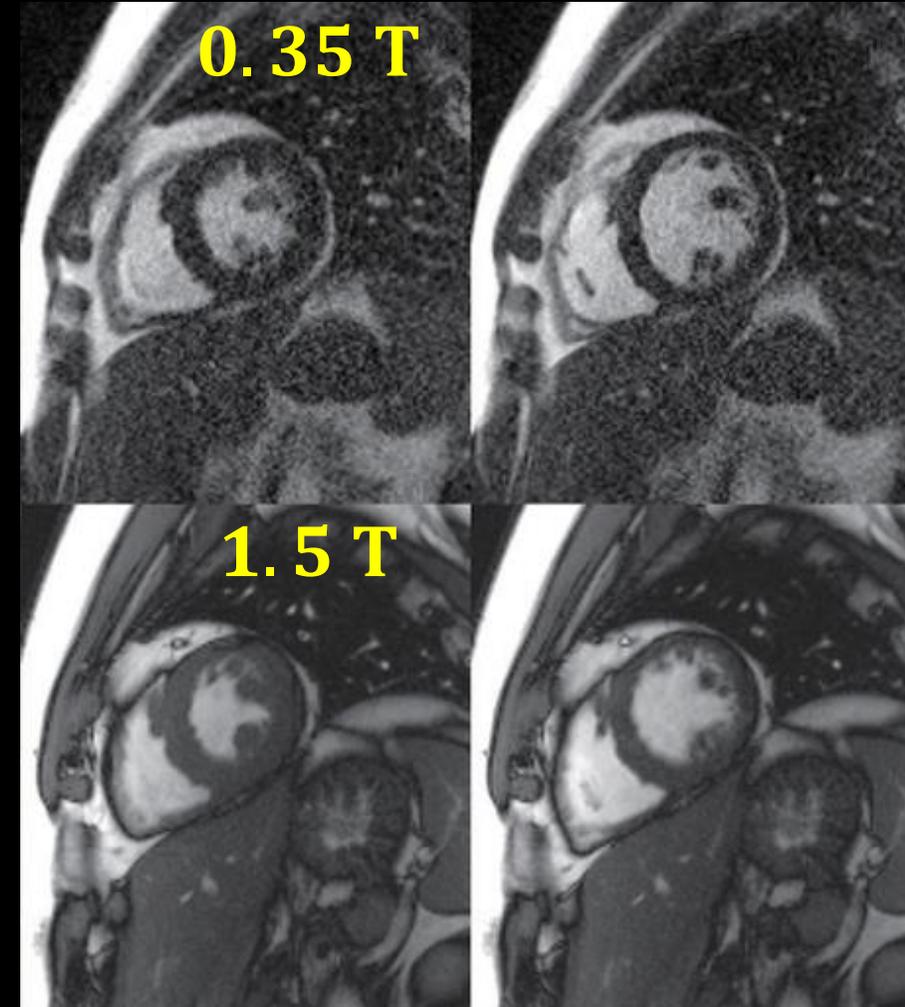
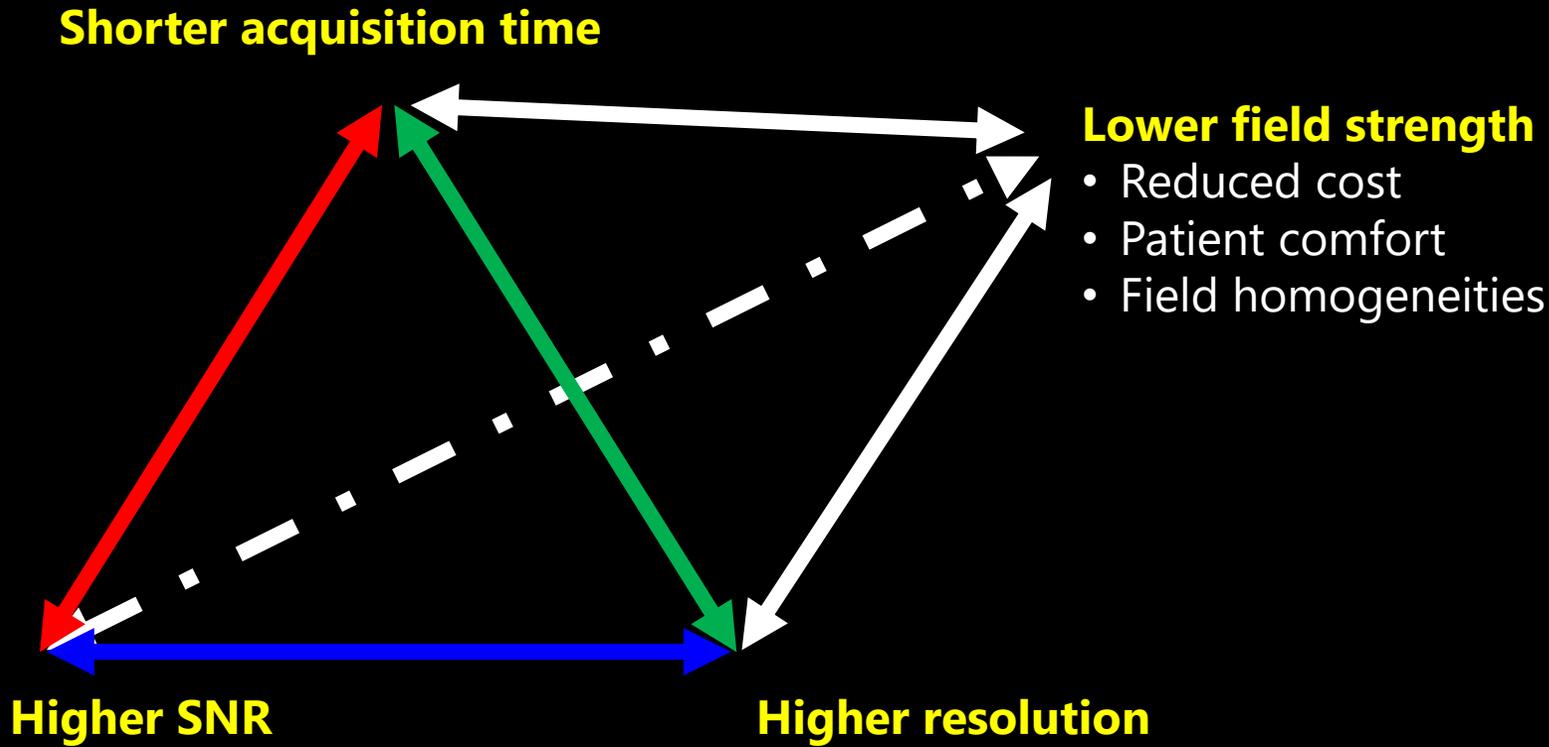
Improved SNR → lower uncertainty

SASHA T1 Mapping

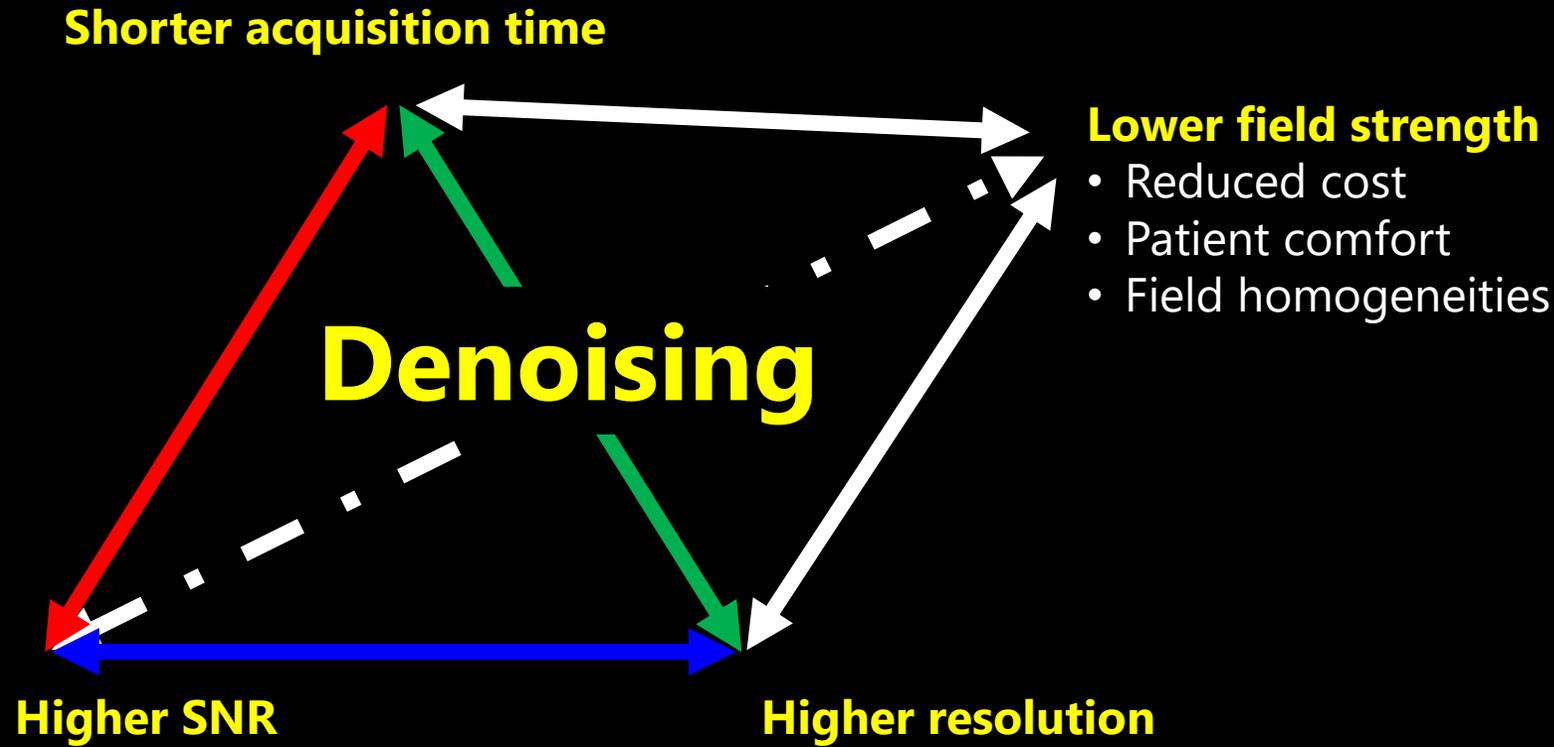


Low-field MRI

$$\text{SNR} \sim B_0^{\frac{3}{2}}$$

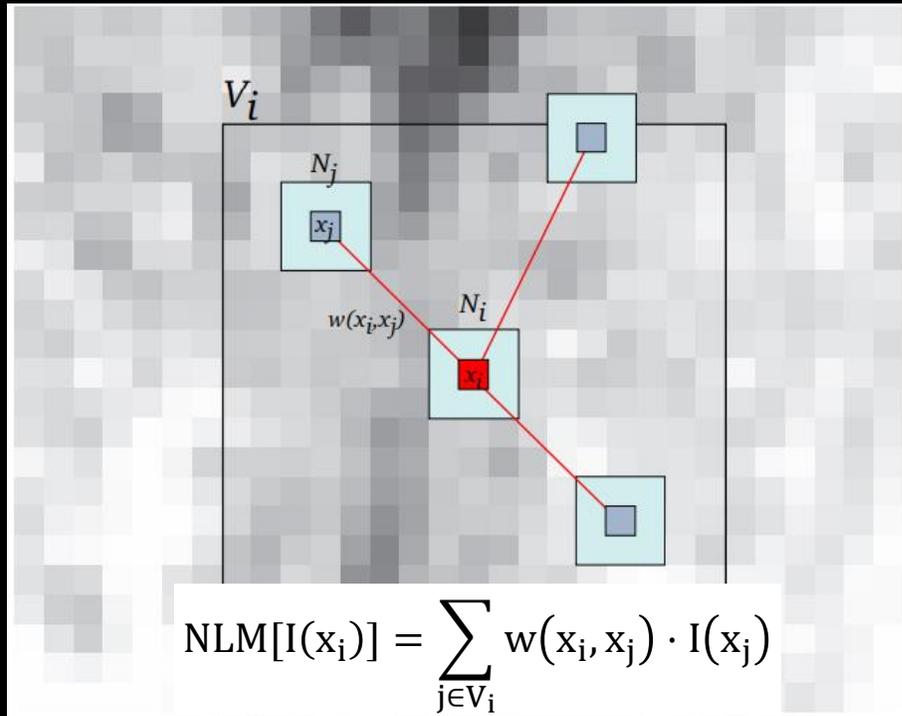


Motivation



1. Shams Rashid, et al. "Cardiac bSSFP MRI at 0.35 T." *Quant Imaging Med Surg* 2018;8(7):627-636
2. Jose Marques, et al., "Low-field MRI: An MR Physics Perspective." *Journal of Magnetic Resonance Imaging* 2019

NLM and BM3D



Non-local Mean (MLM)

- Average based on self-similarity instead of distance (i.e. "non-local")

Limitations

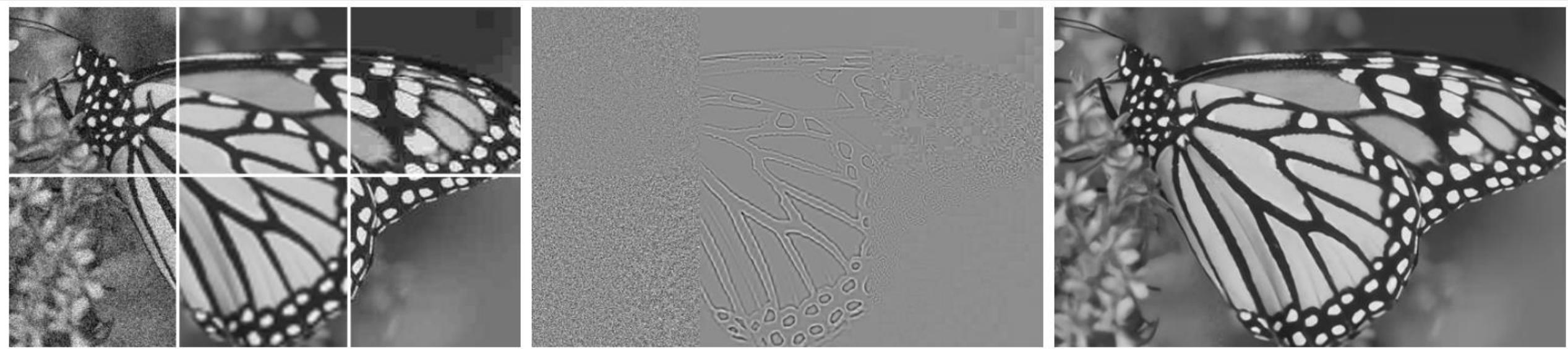
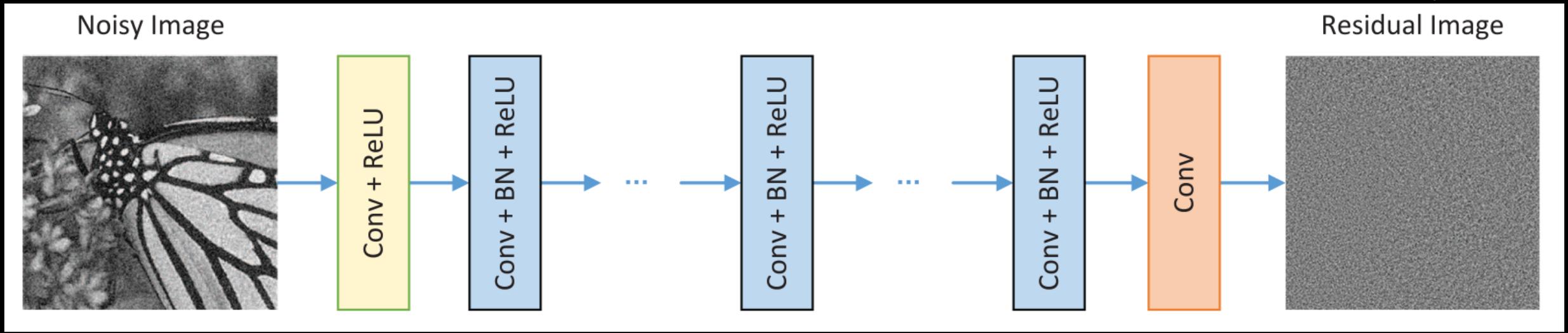
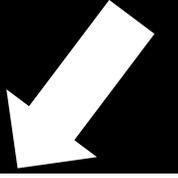
- Slow
- Required human inputs



BM3D algorithm

- Block matching -> 3D stack
- Shrinkage in the sparse 3D transformed domain (Wavelet)

dnCNN (residual learning)

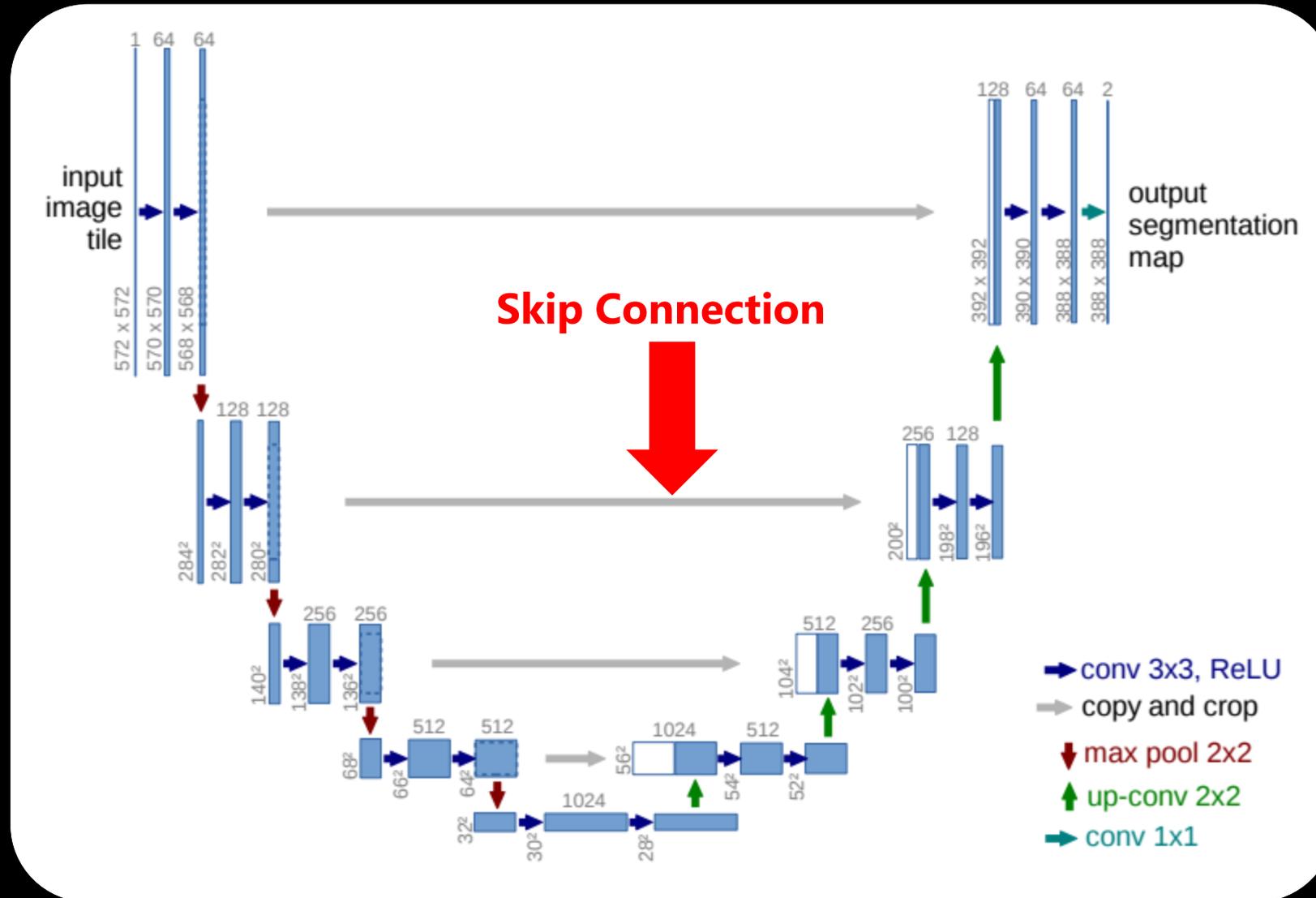


(a) Input Image

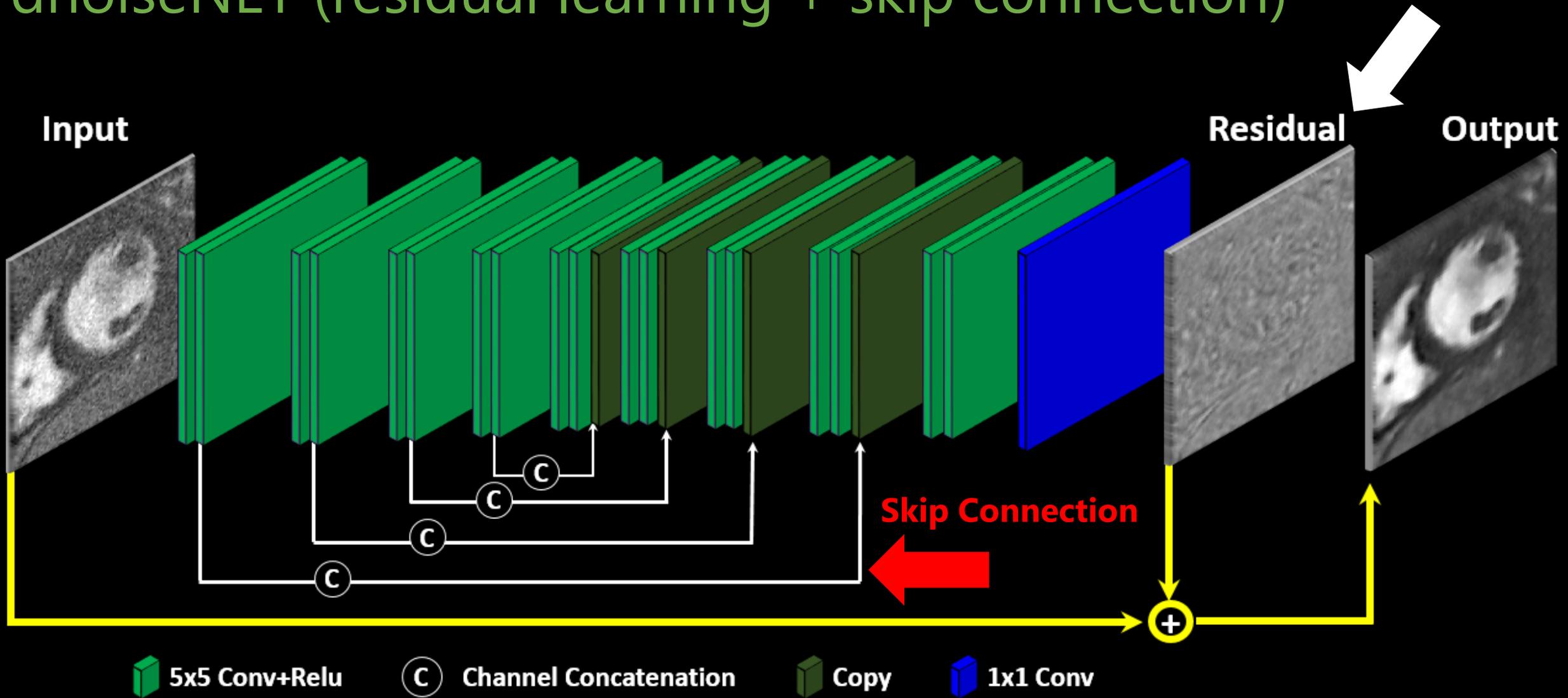
(b) Output Residual Image

(c) Restored Image

U-NET (skip connection)



dnoiseNET (residual learning + skip connection)



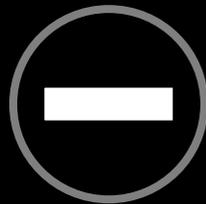
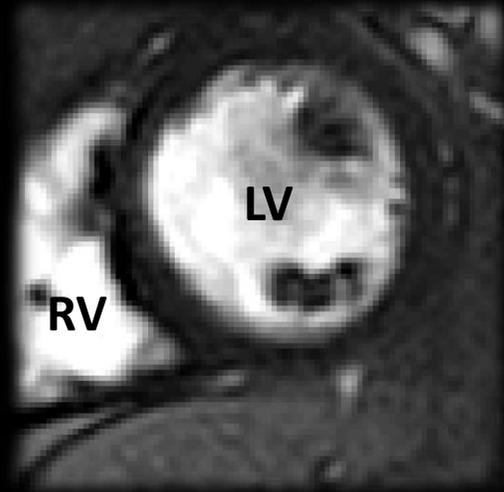
1. Kai Zhang, et al. "dnCNN." *IEEE Transactions on Image Processing* 2017;26(7):3142-3155.

2. Olaf Ronneberger, et al. "U-NET." *MICCAI*, Springer, Cham, 2015;p234-241.

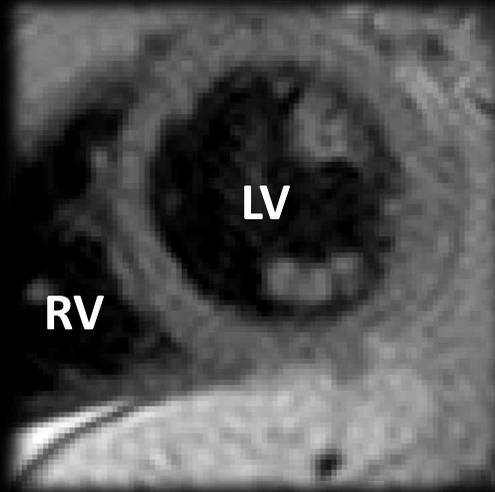
Myocardial ASL Data

- Training and validation data¹:
 - From **22 subjects**: 438/40 images for training/validation
- Test data¹:
 - From **6 heart transplant patients**: 144 images for testing
- i.i.d Gaussian noise was added to magnitude images

Control Image



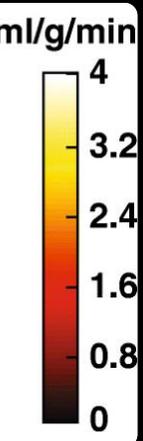
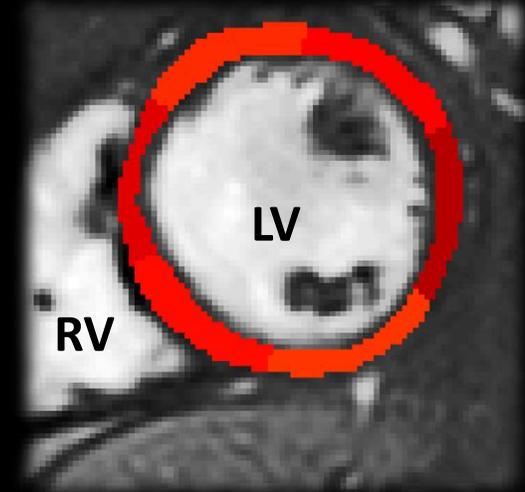
Labeled Image



Quantification



Myocardial Blood Flow (MBF)



Quality assessment: MSE, SSIM, and PSNR

Reference

Noisy

NLM

U-NET

dnCNN

dnoiseNET

Difference Image

Mean \pm SD from 144 images in the test set

MSE = $10^{-3} \times$

29.4 \pm 40.0

4.8 \pm 5.7

3.8 \pm 4.3

4.2 \pm 4.9

2.7 \pm 2.6

SSIM =

0.40 \pm 0.15

0.62 \pm 0.10

0.67 \pm 0.12

0.66 \pm 0.12

0.72 \pm 0.10

PSNR =

18.6 \pm 5.5

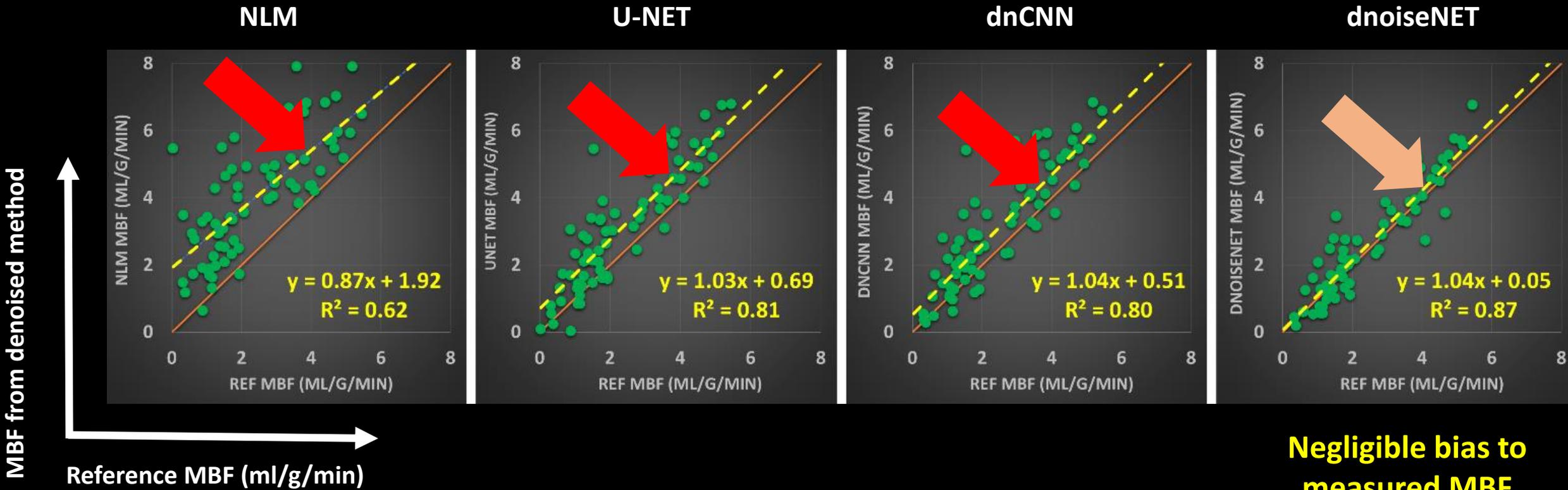
25.2 \pm 4.1

26.4 \pm 4.3

26.0 \pm 4.5

27.6 \pm 4.1

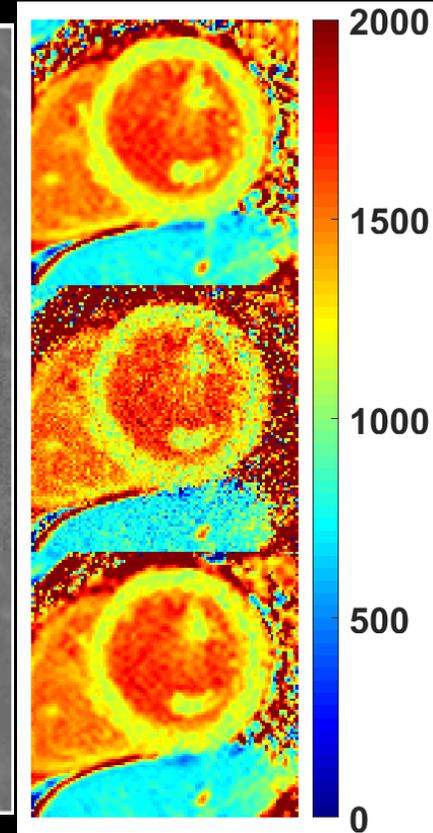
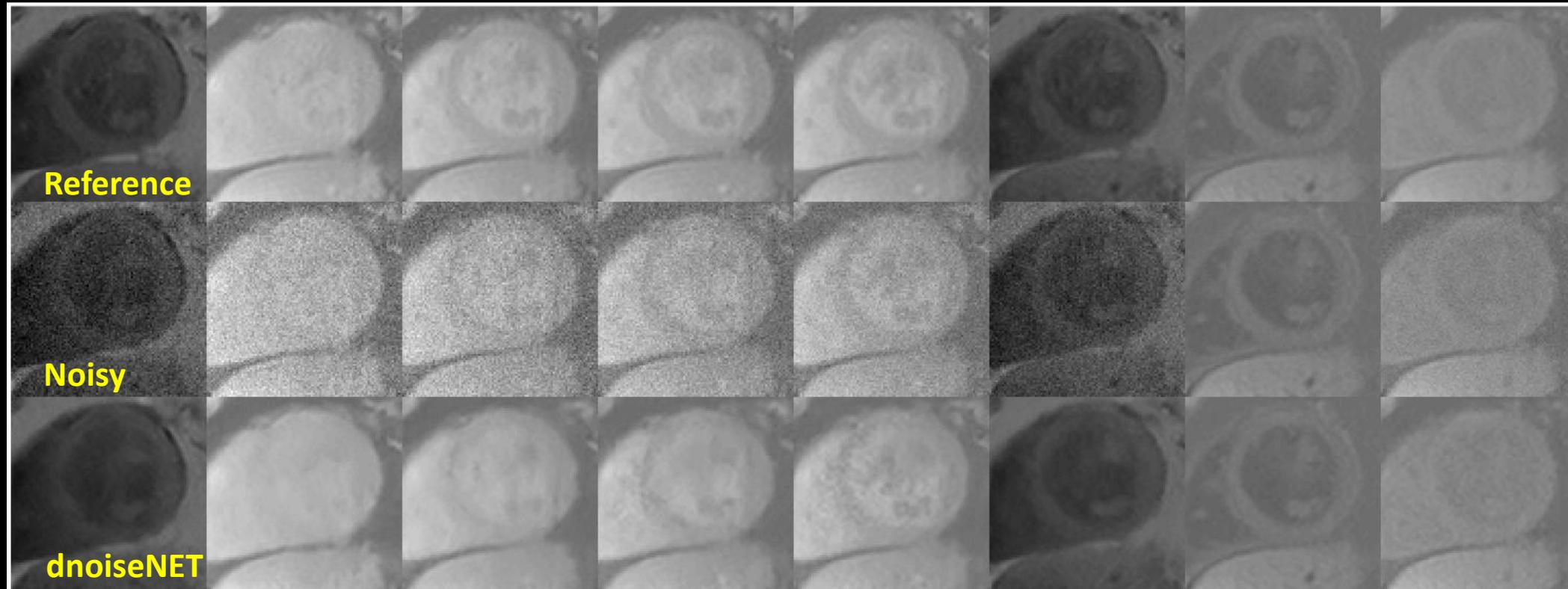
Task-specific quality assessment



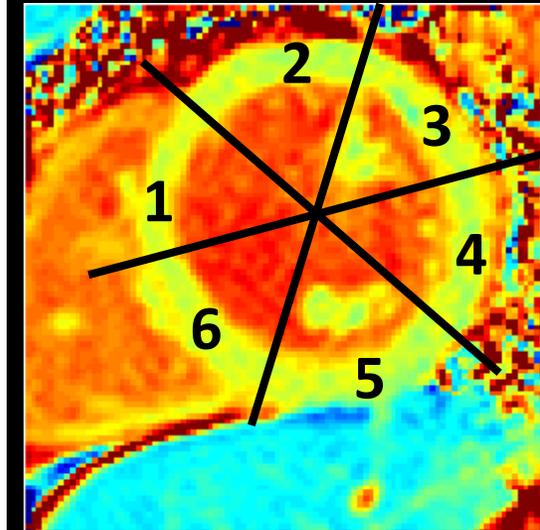
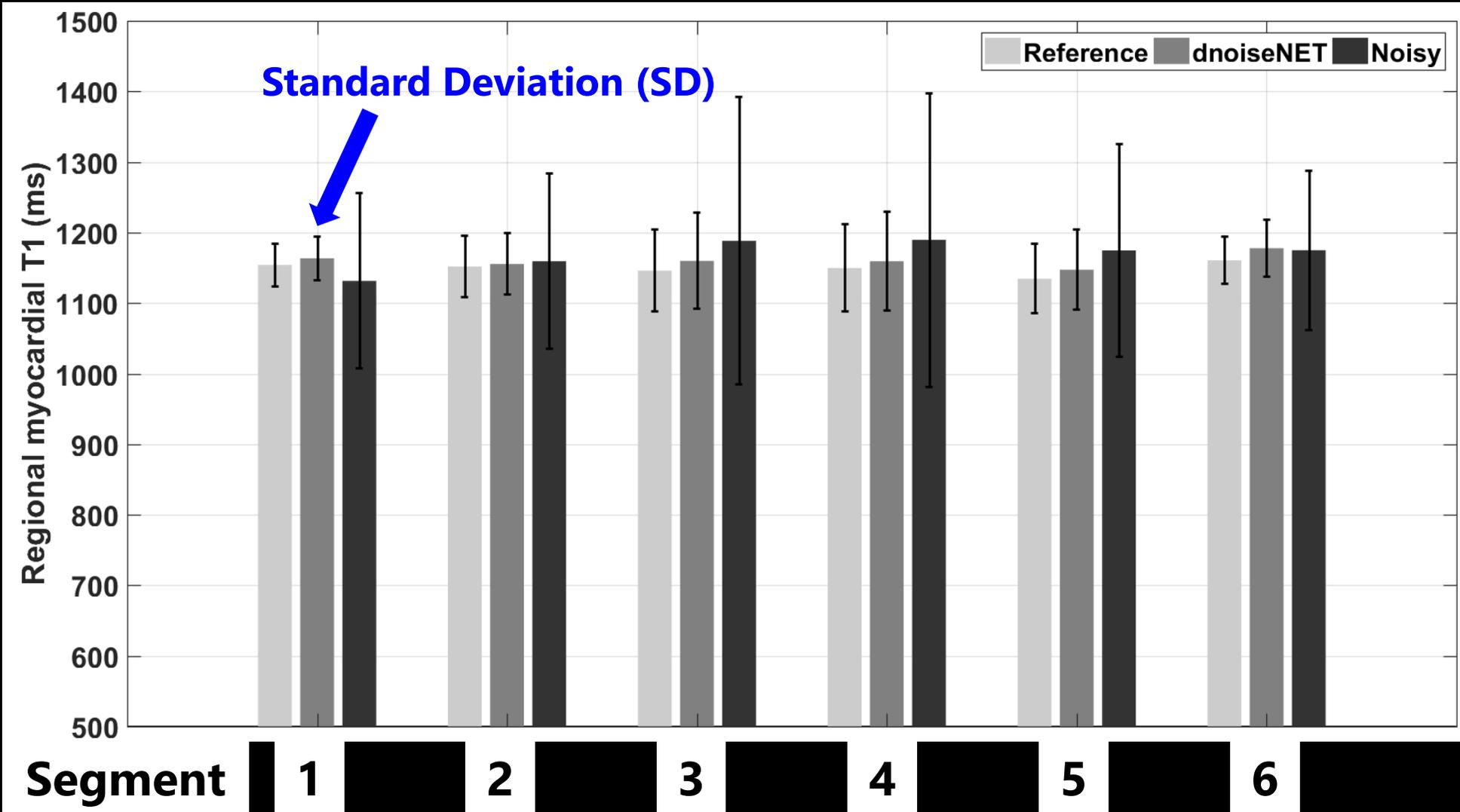
T1 mapping

Raw MR images from MOLLI 5(3s)3

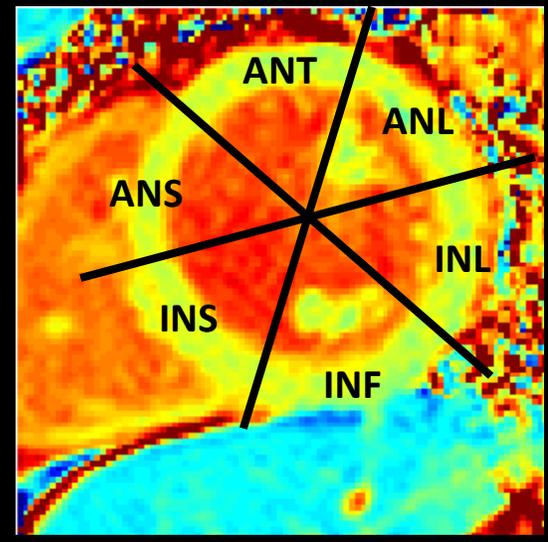
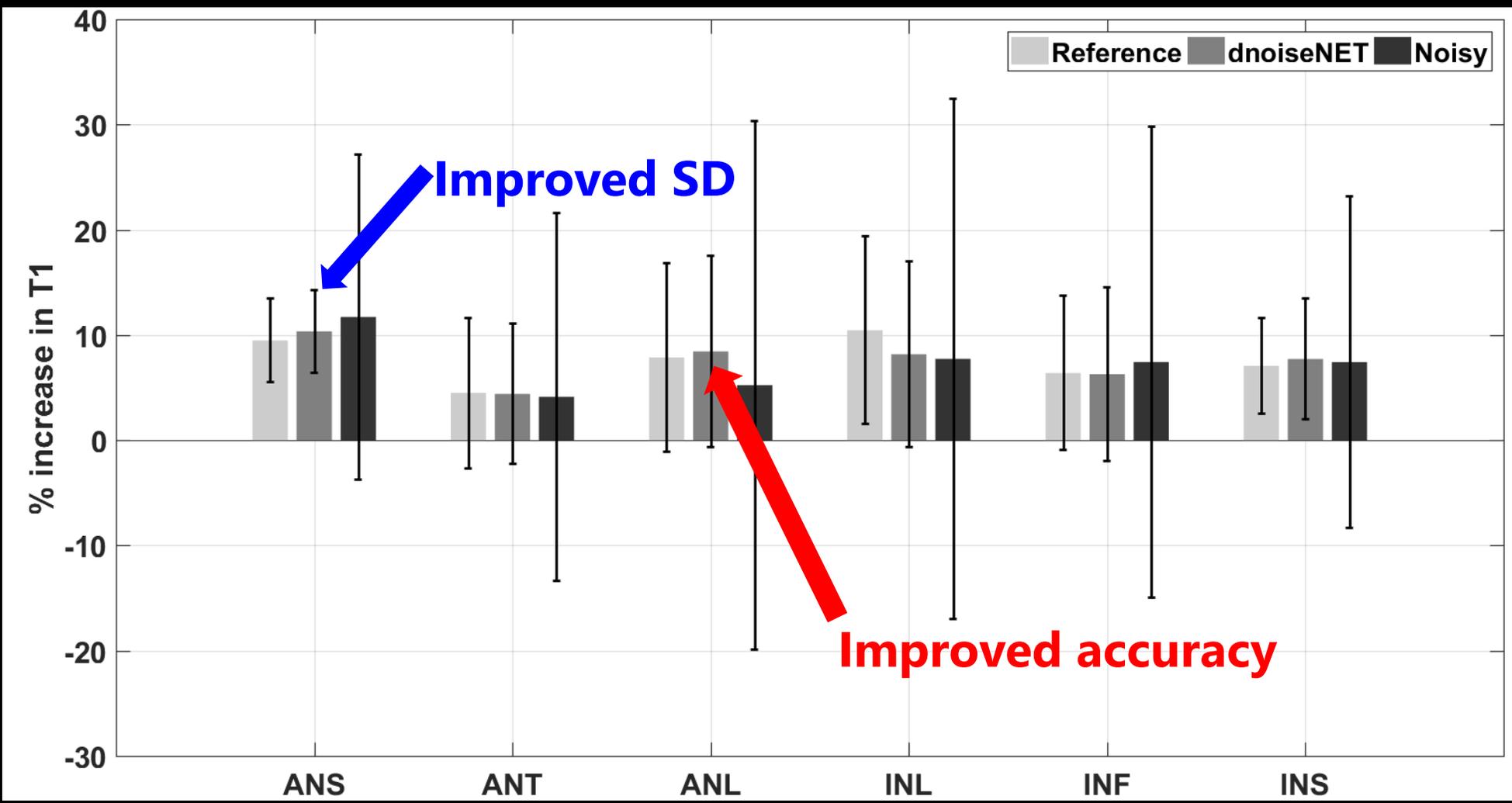
T1 map (ms)



Improved SNR \rightarrow lower uncertainty

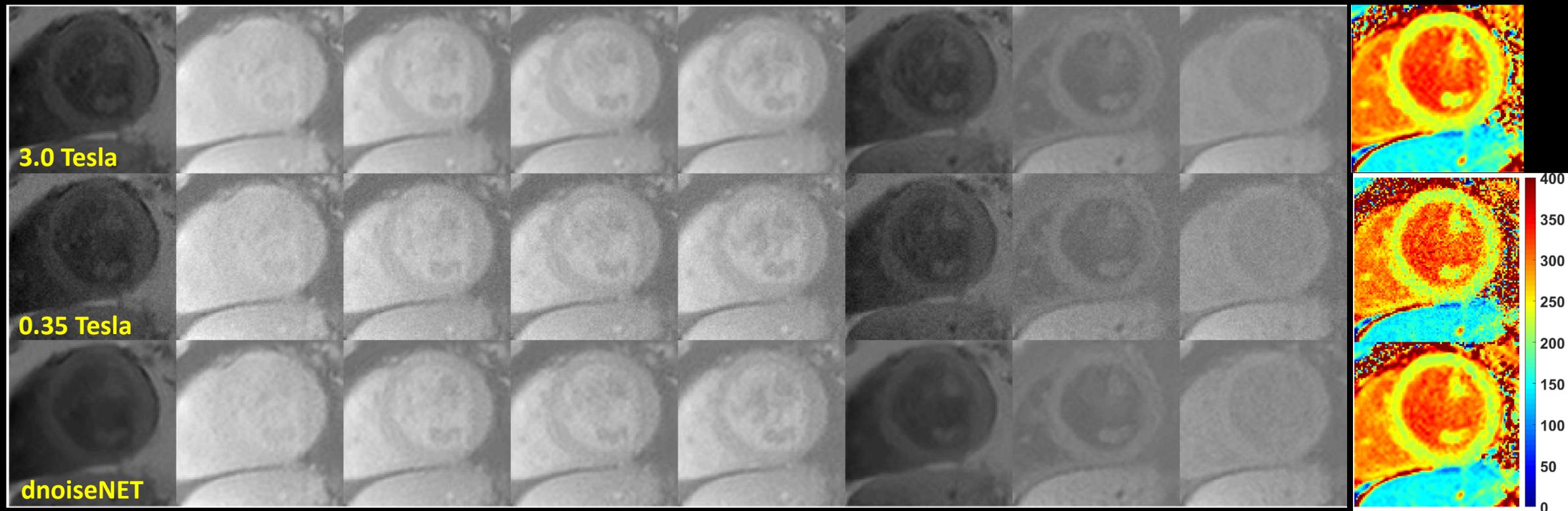


% increase in T1 – rest and stress T1

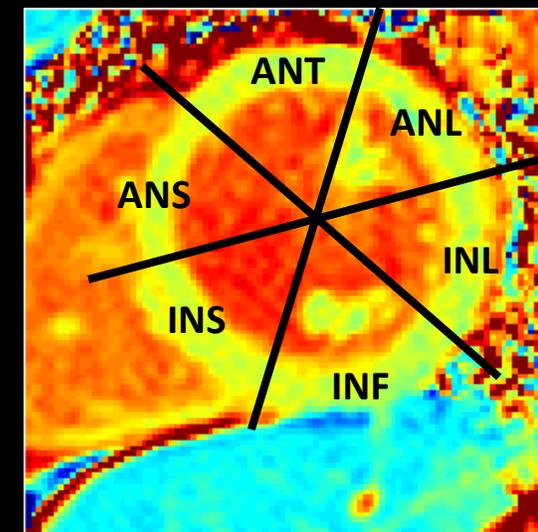
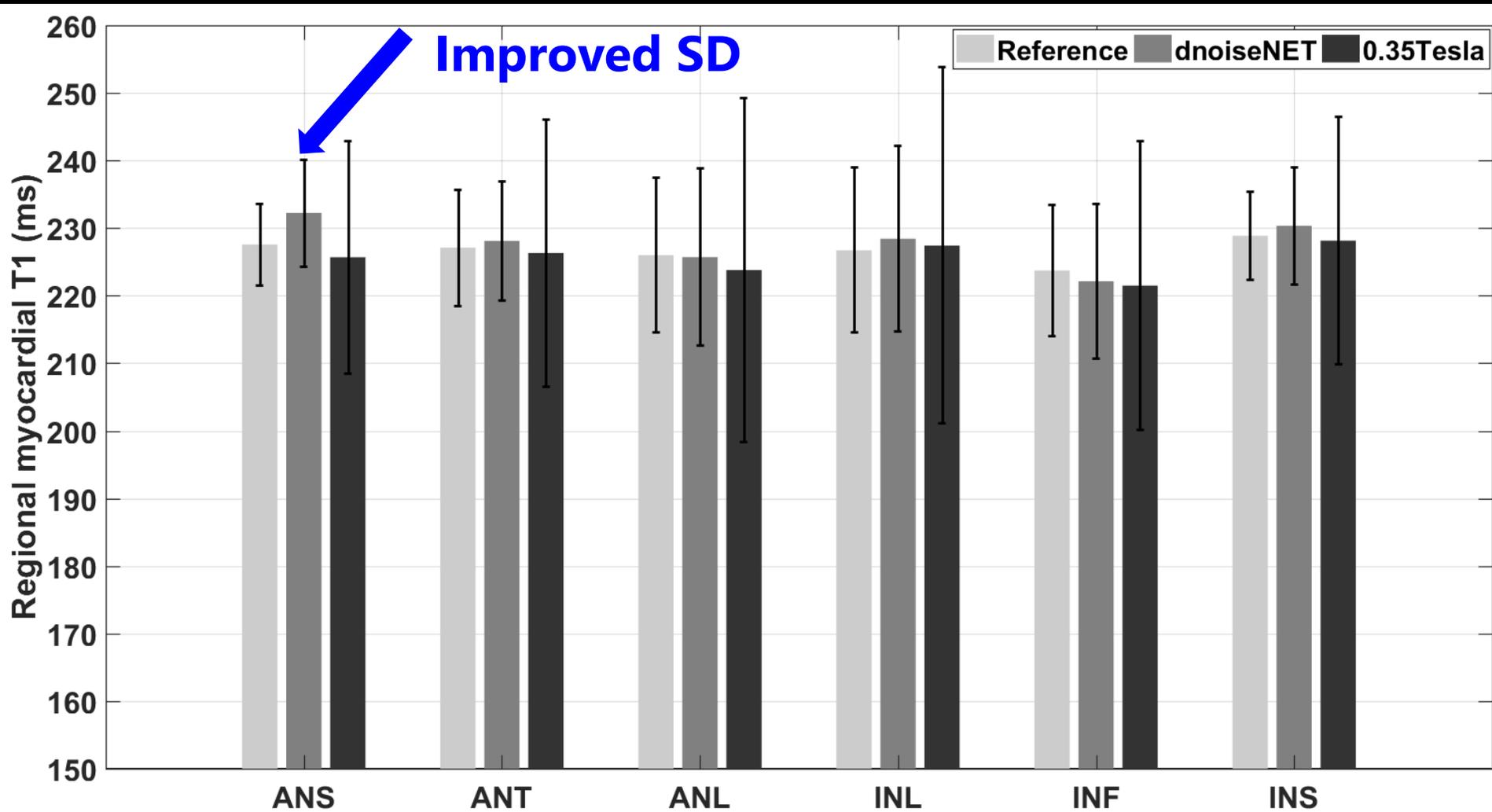


$$\% \text{ increase in } T_1 = 100 \cdot \frac{(T_{1\text{stress}} - T_{1\text{rest}})}{T_{1\text{rest}}}$$

T1 Mapping at simulated 0.35 Tesla



T1 Mapping at simulated 0.35 Tesla



Conclusions

dnoiseNET: Residual learning and skip connections

- Superior performance in term of MSE, PSNR, SSIM
- More importantly, it does not introduce any significant bias to quantitative MBF

Quality Assessment:

- MSE, PSNR, SSIM may not be sufficient for quality assessment
- Task-specific quality assessment is desired (MBF in this work)

Future works:

- Low-field MRI
- Other type of noise distributions (Rician, Parallel Imaging Noise, and residual artifacts from under sampled data, etc.)

Acknowledgements

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 **USC** University of
Southern California

Made For life

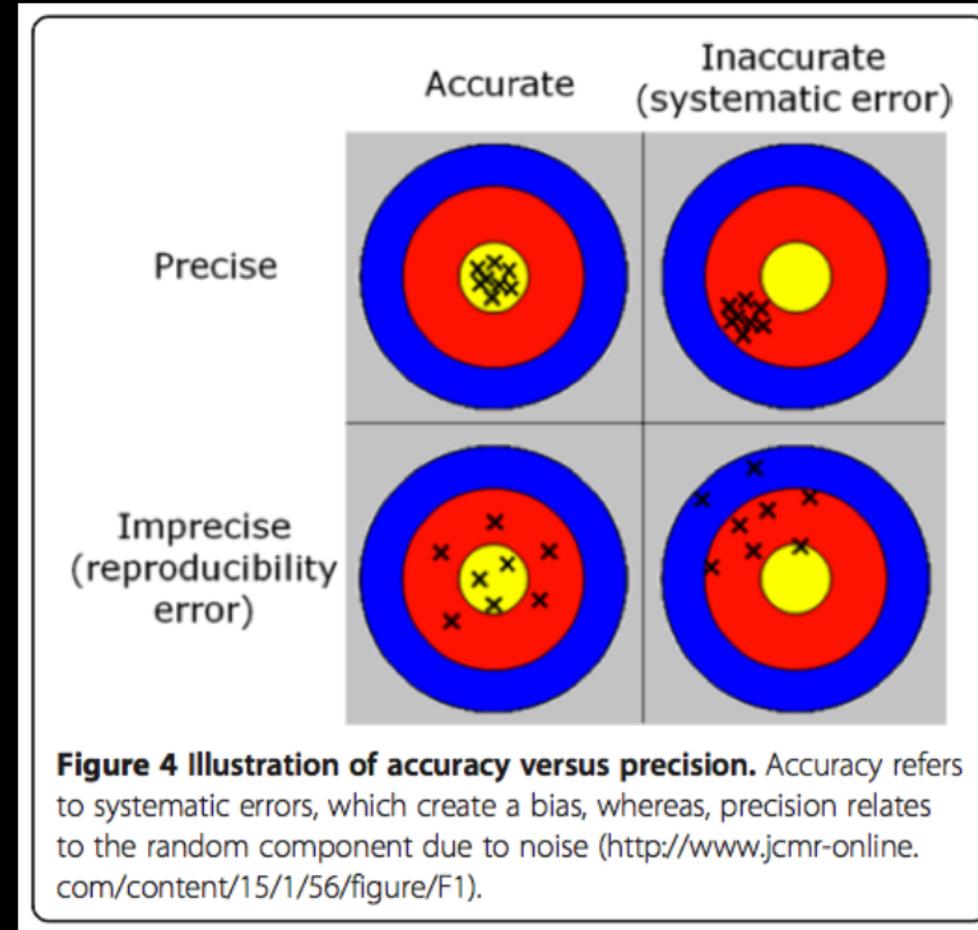
For over 100 years, the Canon Medical Systems `Made for Life' philosophy prevails as our ongoing commitment to humanity. Generations of inherited passion creates a legacy of medical innovation and service that continues to evolve as we do. By engaging the brilliant minds of many, we continue to set the benchmark, because we believe quality of life should be a given, not the exception.

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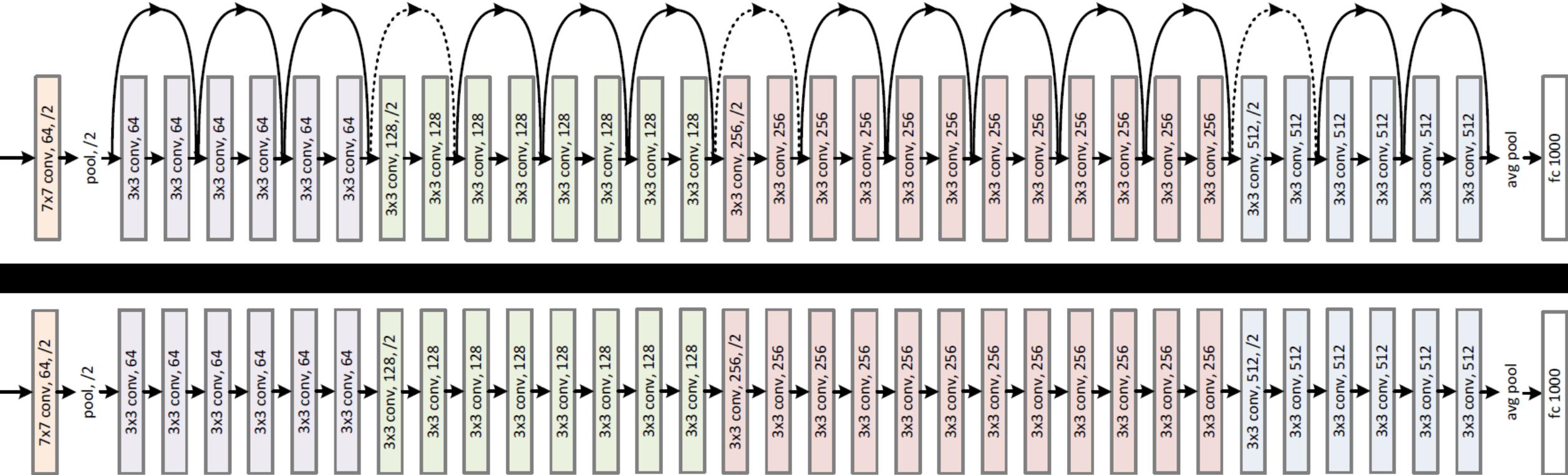
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Accuracy vs. Precision

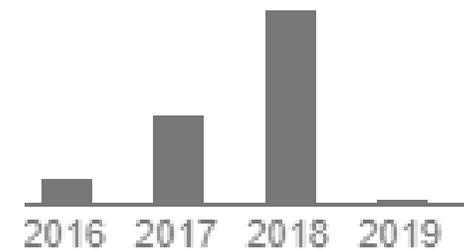
1. Spatial average over ROI $\rightarrow \underline{T_1^{REF}}$
2. Spatial average over ROI $\rightarrow \underline{T_1^{PROP}(t_n)}$
3. Spatial SD over ROI $\rightarrow SD(T_1^{PROP}(t_n))$
4. Accuracy = time average of $abs(\underline{T_1^{REF}} - \underline{T_1^{PROP}(t_n)})$
5. Bias = $\underline{T_1^{REF}} -$ time average of $\underline{T_1^{PROP}(t_n)}$
6. Precision = time average of $SD(T_1^{PROP}(t_n))$
7. Reproducibility = SD of $\underline{T_1^{PROP}(t_n)}$



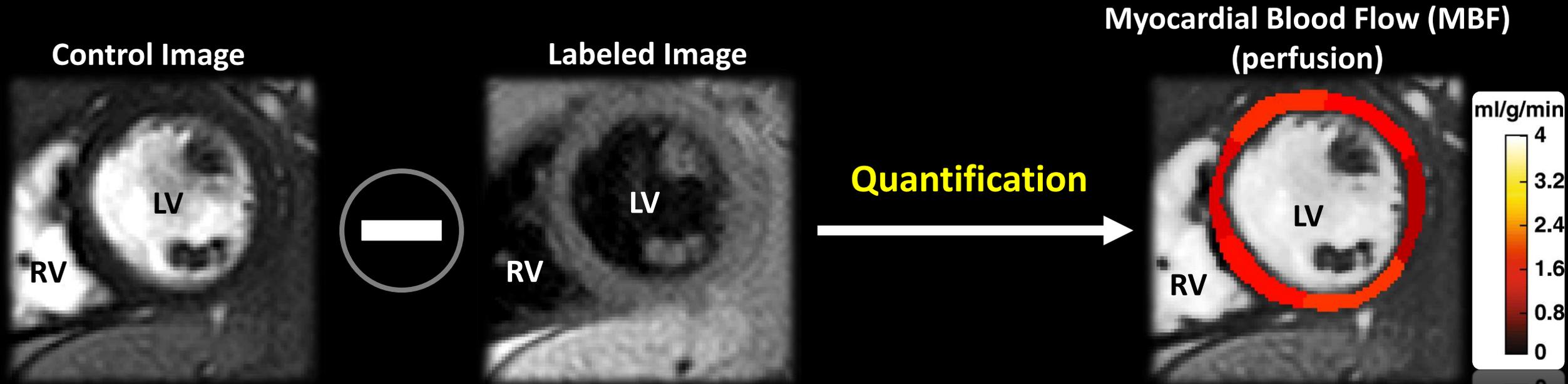
Skip connection



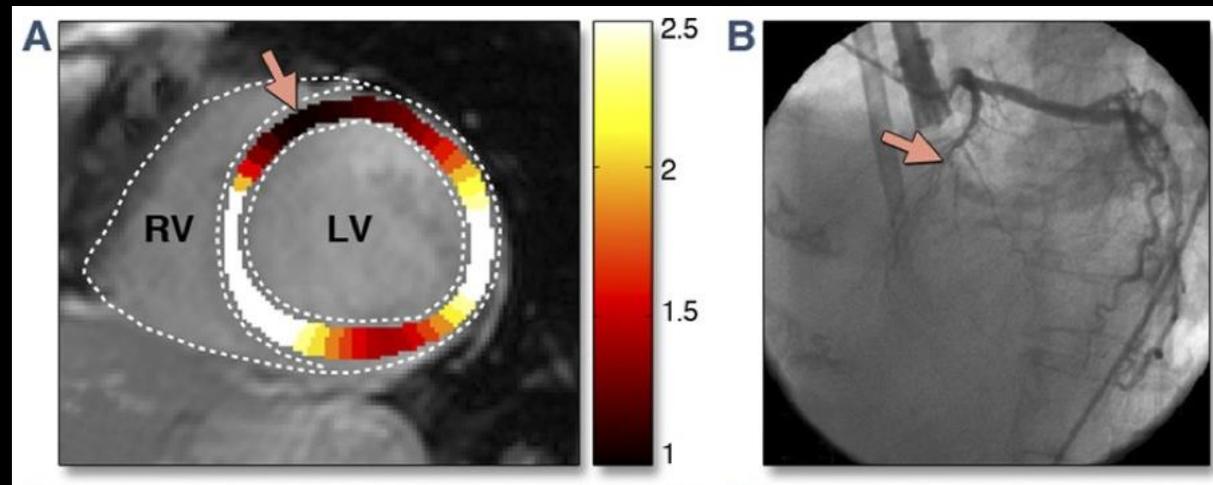
Total citations **Cited by 16660**



Methods(1): Myocardial Arterial Spin Labeling (ASL)



Coronary artery disease (CAD)



1. Kober, Frank et al. "Myocardial arterial spin labeling." *Journal of Cardiovascular Magnetic Resonance* 2016; 18:22.
2. Zun, Zungho et al., "ASL-CMR Detects Clinically Relevant Increases in Myocardial Blood Flow With Vasodilation." *JACC* 2011; 4(12):1253-1261.