



Model-Based Real Time Quantitative Ultrasound and Radar Tom Sharon, Yonina C. Eldar

 $\frac{1}{1} \sqrt{2024}$







Problem statements

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Problem statements



Full Waveform Inversion (FWI) algorithm

An iterative optimization method, based on GD, to achieve quantitative imaging

from the signals



Loss [dB]

Advantages

No training Recovers different physical properties

Disadvantages

• Time consuming

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- Tends to converge to a local minima
- Initial guess should be close to true solution

Full Waveform Inversion (FWI) algorithm

An iterative optimization method, based on GD, to achieve quantitative imaging from the signals

$$L\left(\left\{\theta_{j}\right\}_{j=1}^{n_{m}}\right) = \frac{1}{2}\left\|M - F\left(\left\{\theta_{j}\right\}_{j=1}^{n_{m}}\right)\right\|_{2}^{2} \qquad \text{signals}$$

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The update of the physical property in the i+1 iteration:

$$\theta_j^{i+1} = \theta_j^i - \eta \frac{\partial L(\{\theta_j^i\}_{j=1}^{n_m})}{\partial \theta_j}$$

 $\theta_j \in \mathbb{R}^{n_x X n_z}$ is the j'th physical property out of n_m properties M is the measured signals F() is the known wave propagation equation

Data creation

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Model-based deep learning

Better than **pure analytical** solution

Improved inference resultsImproved inference time

Better than standard deep NN

✓ Requires less training data✓ Better generalization

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Our method: MB-QRUS

- Unfolding of FWI with learned gradients (G = $\frac{\partial L}{\partial \theta_i}$) according to a U-Net based block
- SoS density Network input: ullet• M Measured signals Initial properties US • $\{\theta_j^0\}_{j=1}^{n_m}$ US Model F: **Physical properties** Trained Output: ulletWave reconstruction network • $\{\theta_{i}^{L}\}_{i=1}^{n_{m}}$ n_p equation n_{t} Radar permittivity conductivity n_c Initial properties radar







	<u>MNIST '0'</u>		<u>Noisy Medium</u>		<u>Nois</u>	<u>Noisy input</u>		<u>Two Objects</u>		<u>r Probe</u>		
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	Sos	S	-5	6.33%	1.93%	6 8	8.15%			<u> </u>		
	dens	ity	-5	5.43%	13.15%	% θ	5.10%	<	s i second	> 3.	z minutes	



Realistic liver shapes



Property	NRMSE	PSNR	SSIM	
SoS	-62.87%	1.7%	20.72%	
density	-62.35%	53.37%	33.63%	

OURS	FWI
< 1 second	> 32 minutes





Property	NRMSE	PSNR	SSIM
conductivity	-83.53%	24.61%	1150.35%
permittivity	-79.72%	3.91%	467.33%





Phantom1 Real Data Phantom2







Contribution summary

SC

Quantitative results in real time

- Multiple physical properties reconstruction
- Works on realistic data with high accuracy results + real recorded data
- Using data from only 8 elements Suitable for diverse transmission setups including linear probe















