

Ultrasound Molecular Imaging for Early Breast Cancer Detection: RPCA-Based Filtering for Enhanced Separation of Free and Bound Targeted Microbubbles

Hoda Hashemi

Mentors: Jeremy Dahl, Ramasamy Paulmurugan, Steven Poplack.

Departments of Radiology, Stanford University

July 2024

Some of the pictures are from Dr. Dahl's RAD 235 course slides.

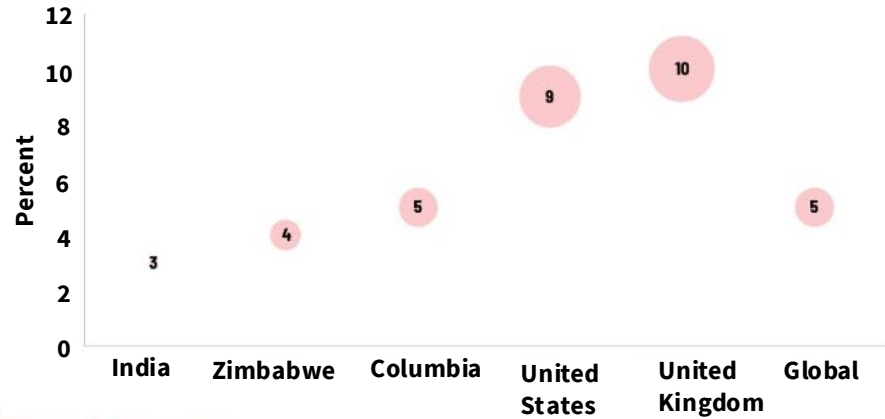
Outline

- Overview:
 - Ultrasound molecular imaging
 - Ultrasound harmonic imaging
- Preclinical method: DTE
- Proposed method: RPCA
- Simulation results
- In vivo experiment

Background

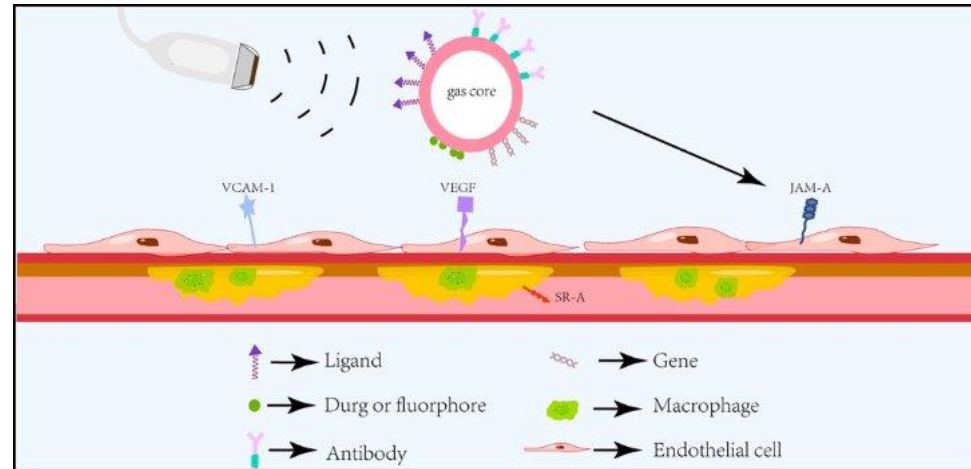
- Breast most frequently diagnosed cancer in women.
- Countries where breast cancer is the most frequently diagnosed cancer in women in 2018:

Cumulative risk of being diagnosed with female breast cancer by age 75 years, globally and in selected countries:



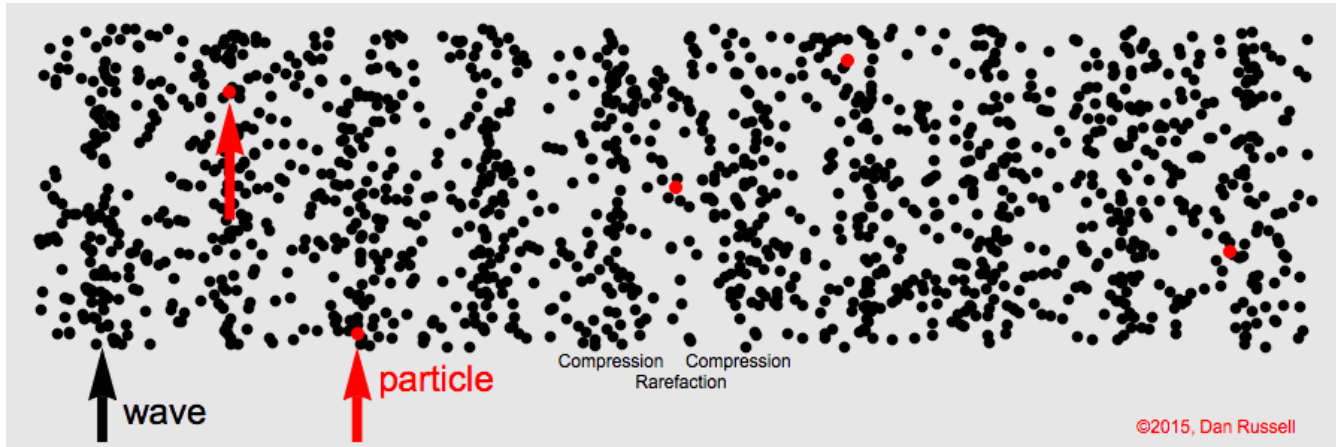
Motivation: Ultrasound Molecular Imaging

- An alternative to conventional ultrasound for the **early detection of breast cancer**,
- By introducing **contrast agents**, such as microbubbles.
- **Microbubbles**: tiny air bubbles that can produce a **large reflection** when injected into the body, improving the **contrast** of the image.
- These air bubbles can be **chemically modified on their surface** to attract to and bind with cancer cells (**Targeted microbubbles**).
- **Goal**: to detect bound microbubbles with ultrasound to identify the **cancer cells**.
- **Challenge**: Free microbubbles



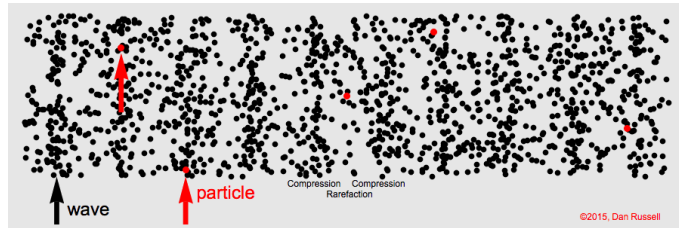
Ultrasound Wave Propagation

- Ultrasound is a longitudinal wave.
- Considered as a sinusoidal wave.
- There are regions of compression and rarefaction on sinusoidal wave.

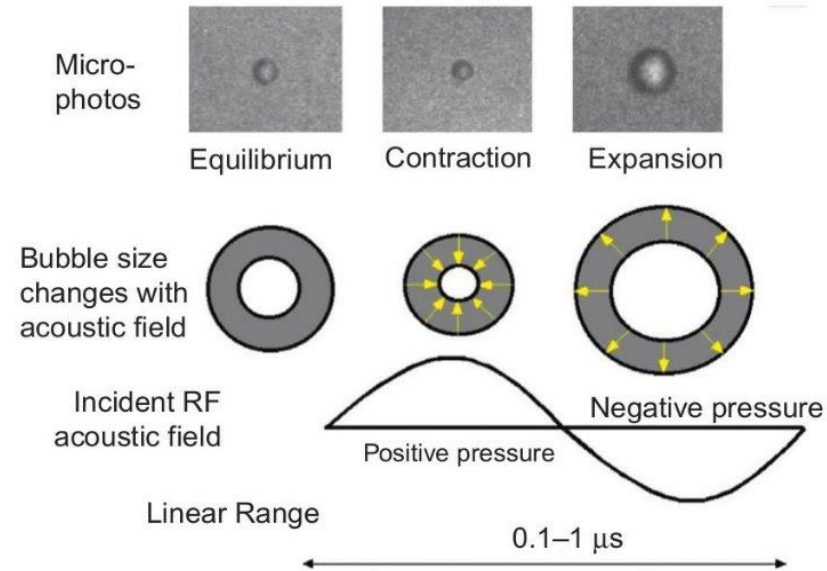


Microbubble Oscillation

- When an ultrasound wave interacts with a microbubble,
- And the ultrasound pressure is very small:
 - Based on these regions of compression and rarefaction,



- The bubble linearly contracts and expands
- Causes the bubble to resonate at ultrasound fundamental frequency (f_0)
- Sends the echos back to transducer (f_0).

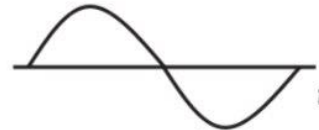


Nonlinear Microbubble Oscillation

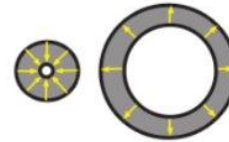
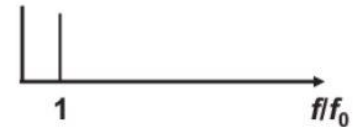
However, within the regular pressure range for the ultrasound imaging:

- Expansion and contraction are not symmetric.
 - Can **expand** with the pressure field
 - A **limit in contraction**
- So **asymmetric** expansion and contraction
- Changes the bubble acoustic response
- Causes the bubble to **resonate at a harmonic frequency** to the incident US wave,
- And emitting harmonic frequency waves back to the US transducer.

Incident RF acoustic field



Incident acoustic spectrum

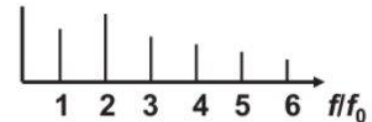


Asymmetrical contraction and expansion produce harmonics

Acoustic bubble response

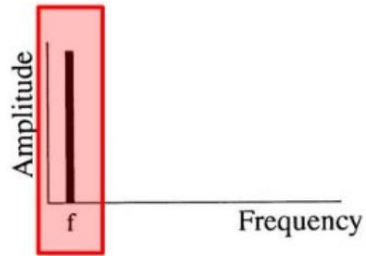
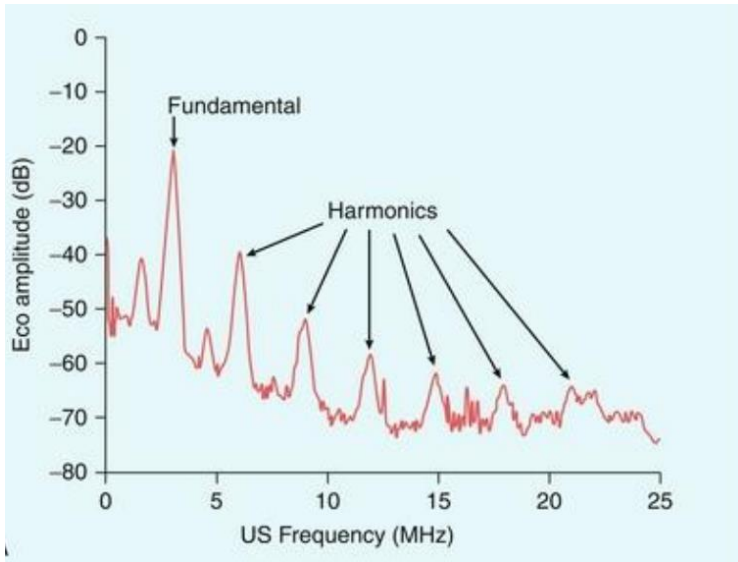


Bubble harmonic response

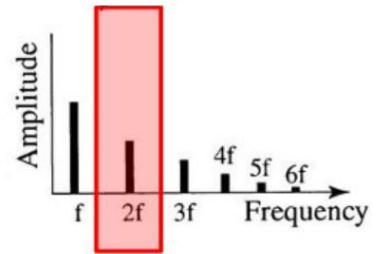


Nonlinear Microbubble Oscillation

Goal: Receive the first harmonic ($2f$) echo instead of the fundamental frequency (f)



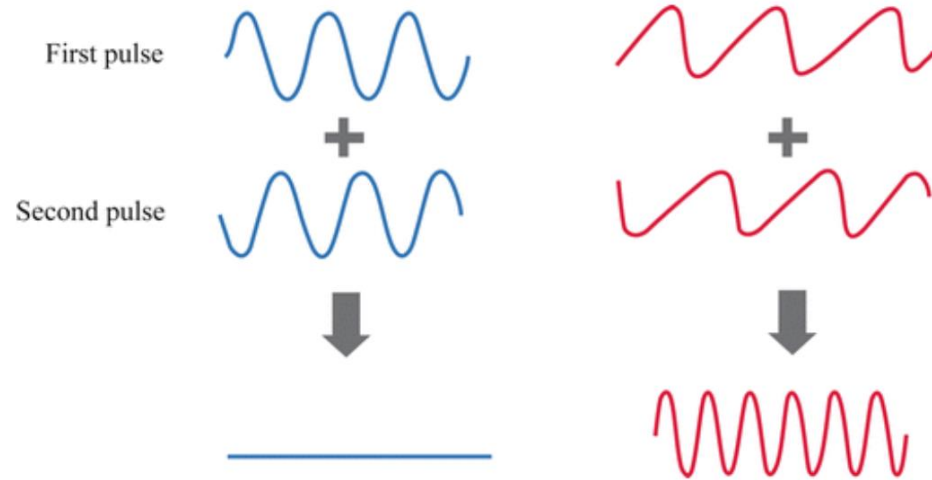
Transmit here:



“Receive” or form image here

Pulse-Inversion Harmonic Imaging

- A pulse is transmitted.
- An **inverted** pulse is transmitted.
- Echoes from the “positive” pulse are **added** to the echoes of the “negative” pulse.
- Eliminates **odd** harmonics and preserves **even** harmonics.
- Preserves **resolution** and increases **SNR**, but loses **frame rate**.

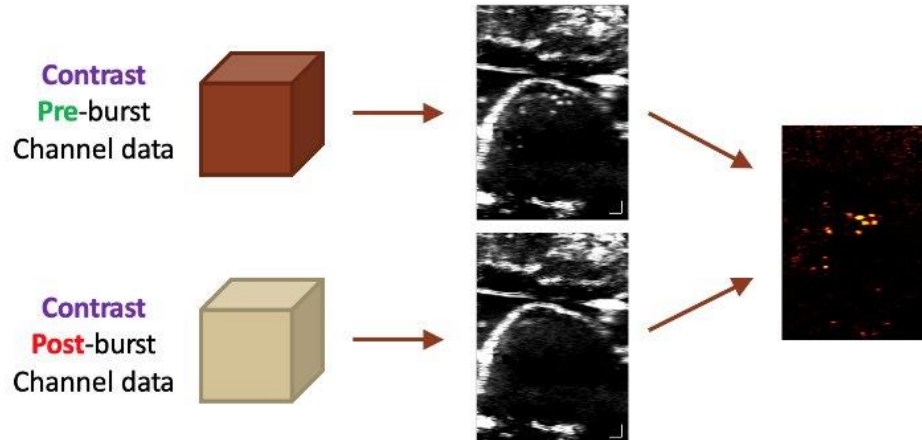


Differential Targeted Enhancement (DTE)

- The preferred preclinical approach to detect microbubbles.
 - A strong acoustic pulse is used to destroy (burst) microbubbles.
 - Images acquired post-burst are subtracted from images acquired pre-burst.
 - Removing the background signals and leaving the targeted microbubbles.

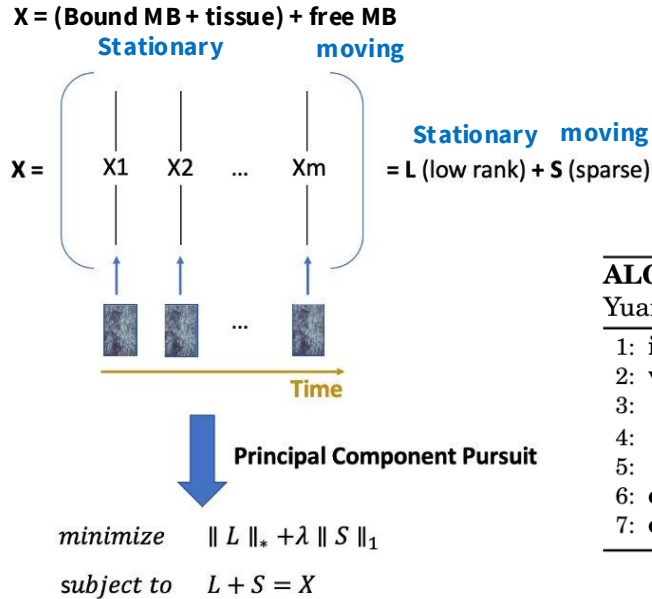
Disadvantage:

- Preclinical method, and can damage the vasculature.
- Still faces challenges in differentiating free and bound microbubbles.



Our Proposed Approach: RPCA

- We propose a **non-invasive** optimization approach using the robust principal component analysis (RPCA) to differentiate between bound and free-floating microbubbles.
- Spatial-temporal data matrix X :



Augmented Lagrangian multiplier:

$$l(L, S, Y) = \|L\|_* + \lambda \|S\|_1 + \langle Y, M - L - S \rangle + \frac{\mu}{2} \|M - L - S\|_F^2.$$

Alternating Directions Method of Multipliers (ADMM)

$$\arg \min_S l(L, S, Y) = \mathcal{S}_{\lambda/\mu}(M - L + \mu^{-1}Y). \quad \mathcal{S}_\tau[x] = \text{sgn}(x) \max(|x| - \tau, 0)$$

$$\arg \min_L l(L, S, Y) = \mathcal{D}_{1/\mu}(M - S + \mu^{-1}Y). \quad \mathcal{D}_\tau(X) = US_\tau(\Sigma)V^*$$

ALGORITHM 1: (Principal Component Pursuit by Alternating Directions [Lin et al. 2009a; Yuan and Yang 2009])

- initialize:** $S_0 = Y_0 = 0, \mu > 0.$
 - while** not converged **do**
 - compute $L_{k+1} = \mathcal{D}_{1/\mu}(M - S_k + \mu^{-1}Y_k);$
 - compute $S_{k+1} = \mathcal{S}_{\lambda/\mu}(M - L_{k+1} + \mu^{-1}Y_k);$
 - compute $Y_{k+1} = Y_k + \mu(M - L_{k+1} - S_{k+1});$
 - end while**
 - output:** $L, S.$
-

- The motion of bound microbubbles, appears as a high spatiotemporal coherence \rightarrow Low-rank matrix L
- The flow of free microbubbles, has low coherence \rightarrow Sparse matrix S

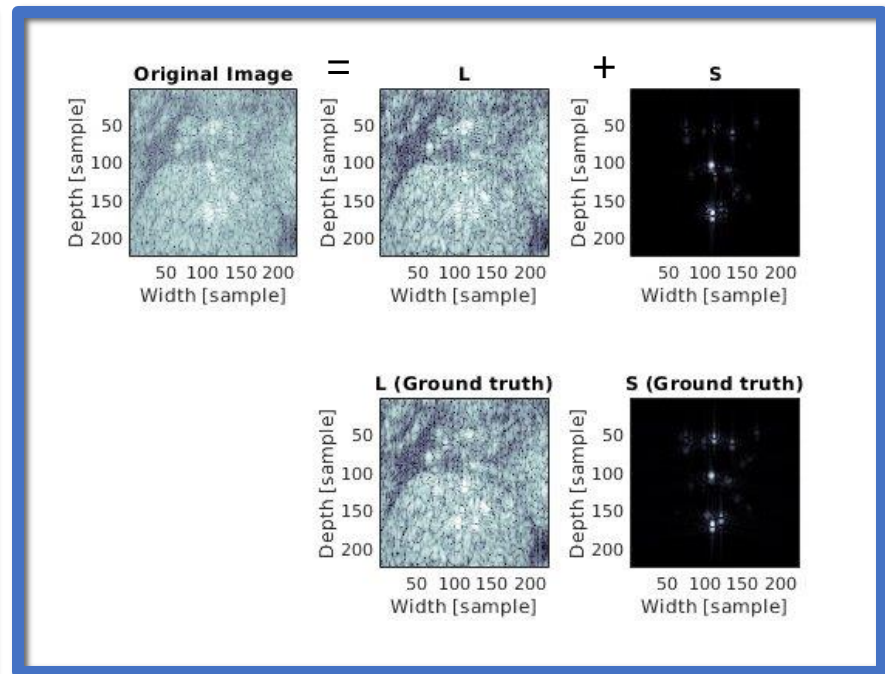
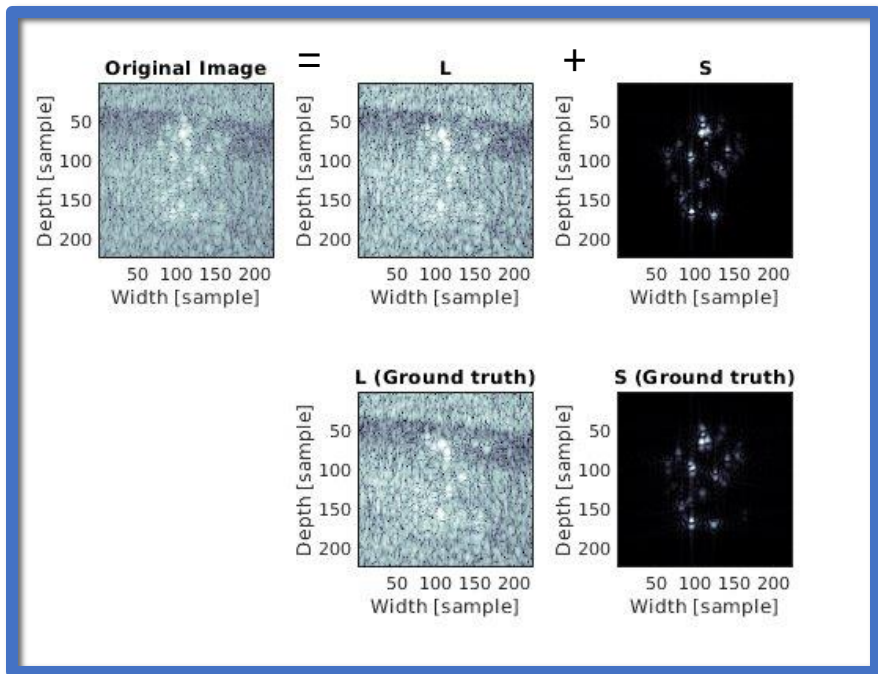
Simulation Data

- 20 simulations, each one has 20 frames over time (Field II)
- Modelled bound and free-floating microbubbles.
- Total of 400 frames.
- L12-3v transducer.
- Images are compounded from 25 TX angles.
- 60 scatterers per resolution voxel.

Simulation Results

Original image (X) = Low rank (L) + Sparse (S)

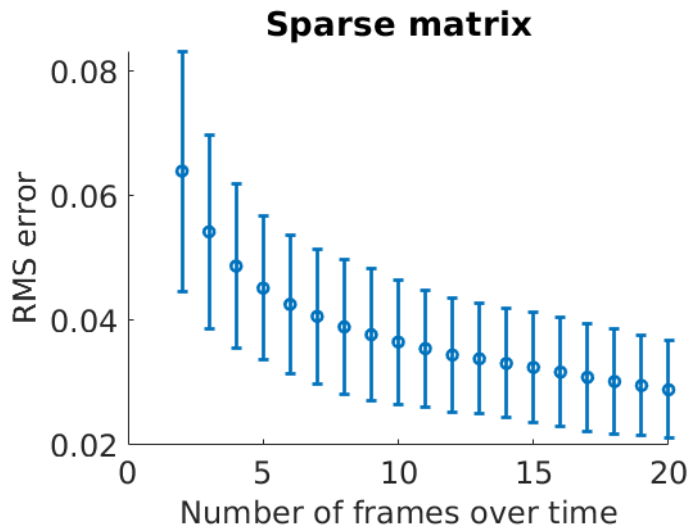
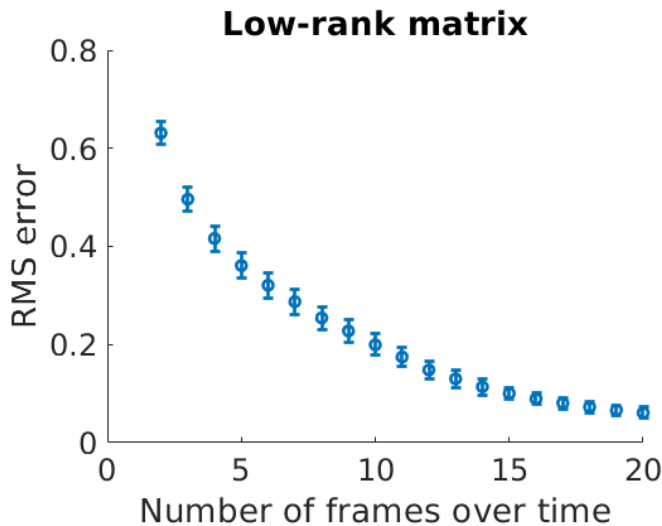
Two examples:



Simulation Results

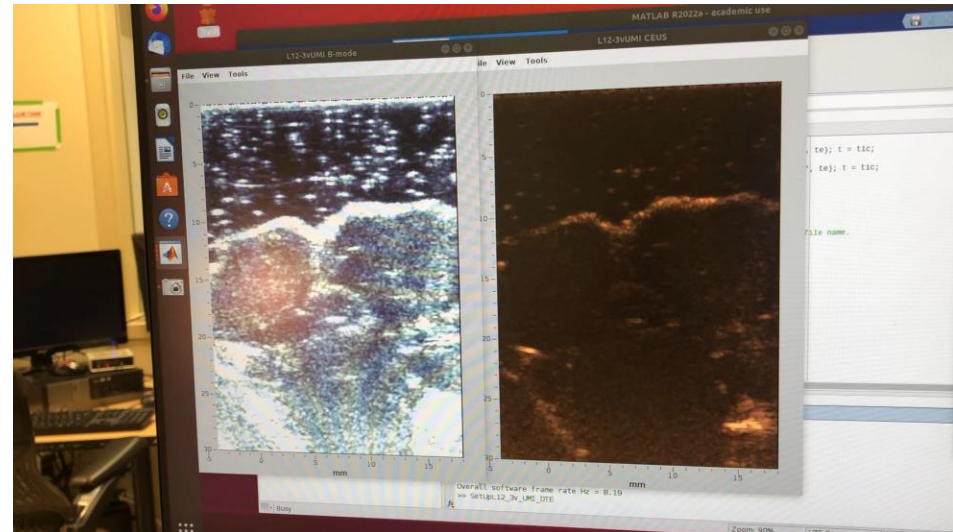
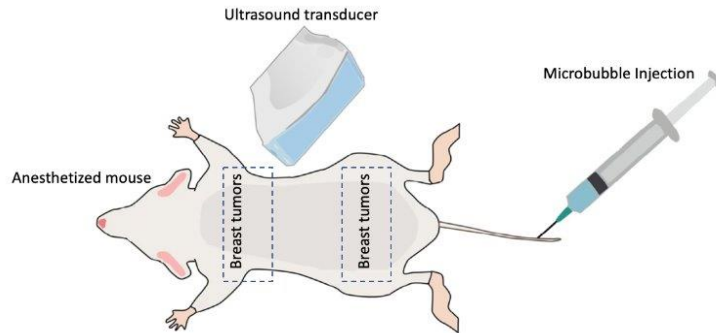
- RMS error = $\sqrt{\frac{(L_e - L_{gt})^2}{m*n}}$
- 20 simulations
- RMS error vs. the number of frames that we used over the time to make the spatial-temporal data matrix X

$$\text{RMS error} = \sqrt{\frac{(S_e - S_{gt})^2}{m*n}}$$



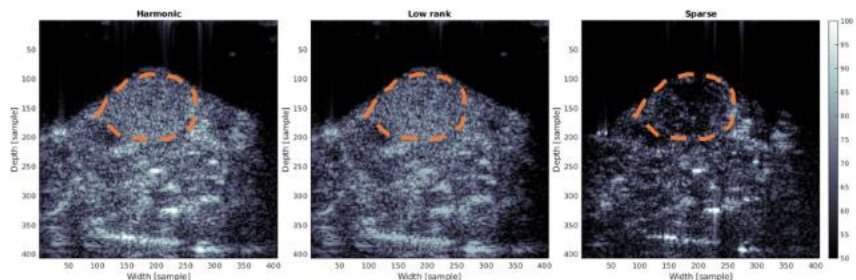
In vivo Data Collection

- Transgenic mouse models of breast cancer development.
- B7-H3- and PDL1-targeted microbubbles injected intravenously.
- Verasonics Vantage 256 research scanner.
- Harmonic imaging (contrast mode).



In vivo Results

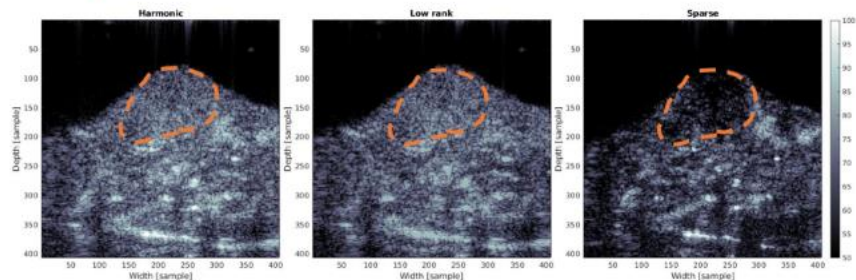
Two Examples of in vivo data:



(i) US image

(j) Low rank

(k) Sparse



(l) US image

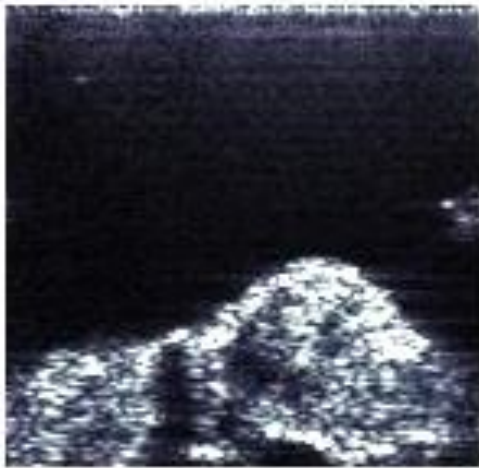
(m) Low rank

(n) Sparse

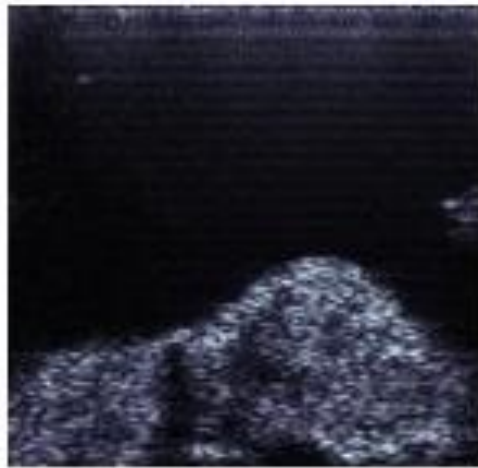
In vivo Results

An Example of in vivo data:

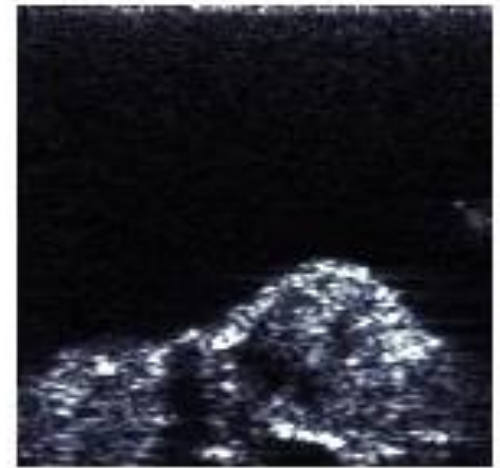
Original harmonic image



Tissue + attached bubbles
(low-rank matrix)



Free-floating bubbles
(sparse matrix)



Summary

- **Ultrasound molecular imaging (UMI)** enables early detection of **breast cancer**.
- Detection of bound microbubbles is a **vital step** in UMI, which is challenging due to the presence of free microbubbles.
- We applied a **novel RPCA-based method** to separate free and bound microbubbles.
- The method was applied to **simulation** and **in vivo data** from transgenic mouse models of **breast cancer** development using targeted microbubbles.
- The stationary output of this method (tissue + bound microbubbles) can be used as input for **neural networks** to improve the differentiation of bound and free-floating microbubbles in nondestructive UMI.

Acknowledgment

- Ultrasound Imaging & Instrumentation Lab
- Stanford Cancer Imaging Training (SCIT) Program
- Radiological Sciences Laboratory (RSL)
- NIH T32 Training Grant
- Canary Center at Stanford
- Dongwoon Hyun
- Jihye Baek
- Arut Natarajan
- Farbod Tabesh
- Ramasamy Paulmurugan
- Jeremy Dahl



CANARY CENTER
AT STANFORD

