

BoneMet: An Open Large-Scale Multi-Modal Murine Dataset for Breast Cancer Bone Metastasis Diagnosis and Prognosis



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Overview

We introduce BoneMet: the first Al-ready, large-scale, high-resolution, open-source, and multi-modal murine medical dataset focusing on breast cancer bone metastasis (BCBM) foundation AI models.

- Spanning 5 years (2019–2024).
- Compiling from > 600 murine across 5-week longitudinal scans over the span of breast cancer metastasis from early-stage to failure.
- Containing over 67 terabytes (TB) of spatial-temporal-aligned multi-modal data:
- A. Image modalities:
- 2D Rotation X-Ray (260 projections / mouse)
- ➤ 4 3D CT volumes (Recon-CT, Seg-CT, Regist-CT, and Rol-CT)
- B. Text modalities:
- Medical records (Pre-Op & Post-Op)
- Quantitative analysis
- Supporting at least 6 breast cancer bone metastasis applications.
- Streamlining medical image data preprocessing with 3 open-source APIs for segmentation, registration & Rol cropping.
- Releasing first real-world paired radiograph/CT resource—enables sparse-angle CT reconstruction.

Motivation & Challenges

Motivation

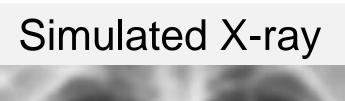
- Breast cancer is the most common non-skin cancers in women.
- Bone metastases dramatically reduce 5-year survival and cause severe pain, immobility and pathologic fractures.

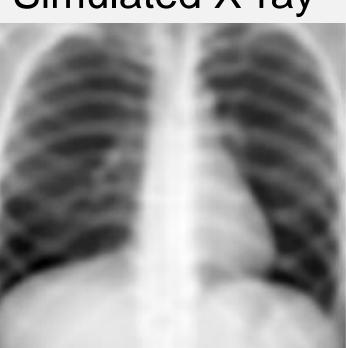
Challenges

- Manual medical image preprocessing and analysis are timeconsuming.
- Existing public X-ray/CT datasets are *low resolution, inadequate* and noisy diagnosis labels, limited of dataset sizes, and lack of diverse modalities.
- No real-world paired radiograph-CT dataset is available to enable sparse-angle CT reconstruction.

Ground Truth X-ray

NIH)



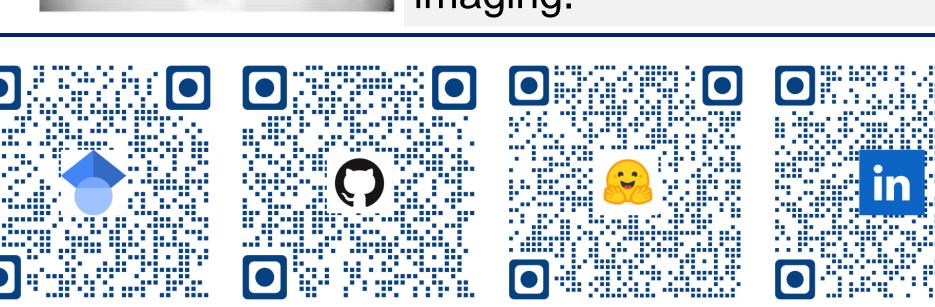


Paper

To date, existing paired Xray/CT datasets are generated through simulation like digitally reconstructed radiographs DRRs), rather than acquired from real, spatially aligned imaging.

Dataset

Contact



Code

Dataset 1. Rotation-X-Ray 5-week longitudinal scans paired 2D/3D modalities 2. Recon-CT 0. In vivo CT scans Seg-API/ **BoneMet Dataviewer** ® Segmentation **Data Collection** (Pre-Op & Post-Op) 3. Seg-CT 6. MiceMediRec **Processing Tool Dataset** Six Sub Datasets **Registration** (to **Quantitative analysis** CTAn®, Abaqus®, **RoI Cropping & RSA-G2 Solids Analyzer** ® **Sequence of scans RoI-API** & 4. Regist-CT

Our dataset is compiled from over 600 murine across 5-week longitudinal scans (just one week shown in the figure above), spanning 5 years with diverse physiological characteristics. It contains 3 components: 1) 67 TB data-collection workflow, 2) 3 APIs as processing tools for workflow automation (e.g., tibiae segmentation, registration (alignment), and region of interest (RoI) cropping), 3) 6 sub datasets with paired 2D/3D modalities: 1 of which is 2D X-ray with 260 rotational projects for each mice, 4 are 3D CT composed by 2D cross-section slices, and 1 is compiled and annotated by human experts serving as the groundtruth labels (e.g., cancer or non-cancer) and biological records (e.g., pre-op records including diagnosis (bone lesions), treatment (chemo, radiation therapy, and placebo control) and post-op quantitative analysis).

CTAn®

5. RoI-CT

Experiments & Results

I. Spatial (260 projections)-temporal (5 weekly scans) aligned (STA) Rotation-X-Ray dataset for BCBM early diagnosis

Table 1:BCBM diagnosis on the Rotation-X-Ray dataset.

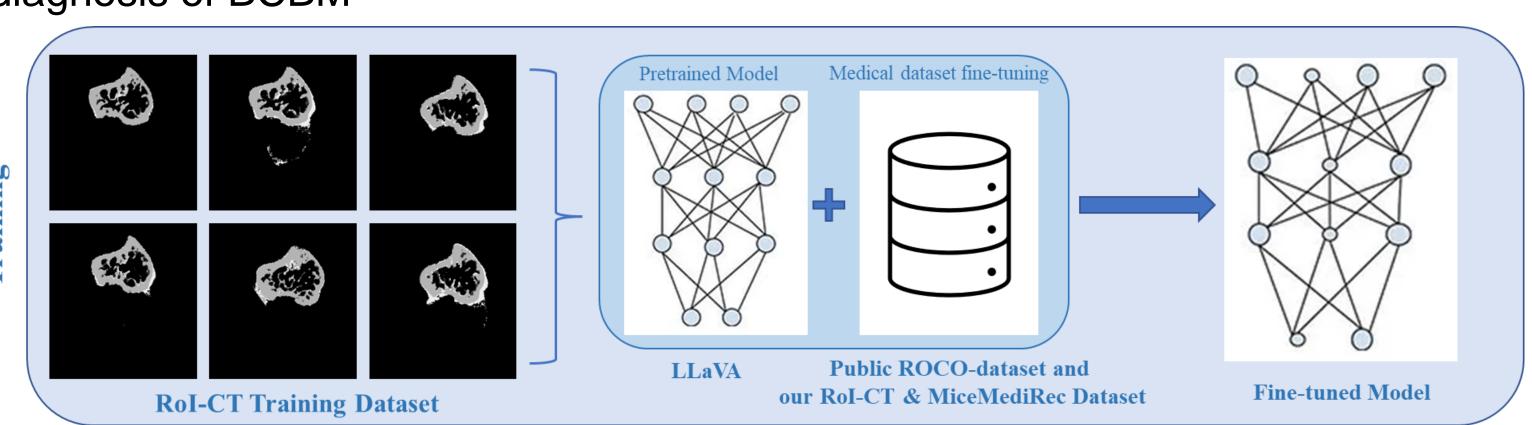
Methods	Training				Test			
	Precision	Recall	F1-Score	Accuracy	Precision	Recall	F1-Score	Accuracy
ViT (w/o STA) ViT (w/ STA)	99.6 95.4	99.7 96.9	99.7 96.1	99.5 95.0	92.3 92.1	80.5 90.6	86.0 91.3	79.1 89.0

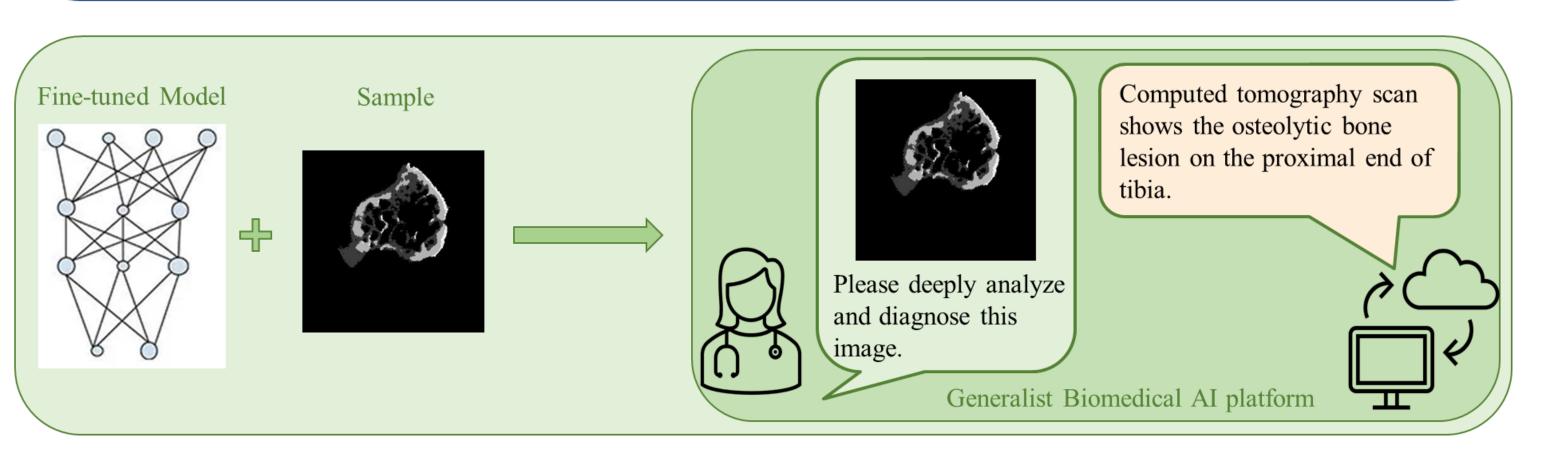
Observation:

- Dataset effectiveness has been validated through both self-supervised pretraining and supervised finetuning.
- Spatial-temporal alignment mitigates overfitting and enhances model generalizability.

Experiments & Results (continue)

2. Multi-modal Generative AI (GenAI) and Large Vision Models (LVM) for accurate diagnosis of BCBM





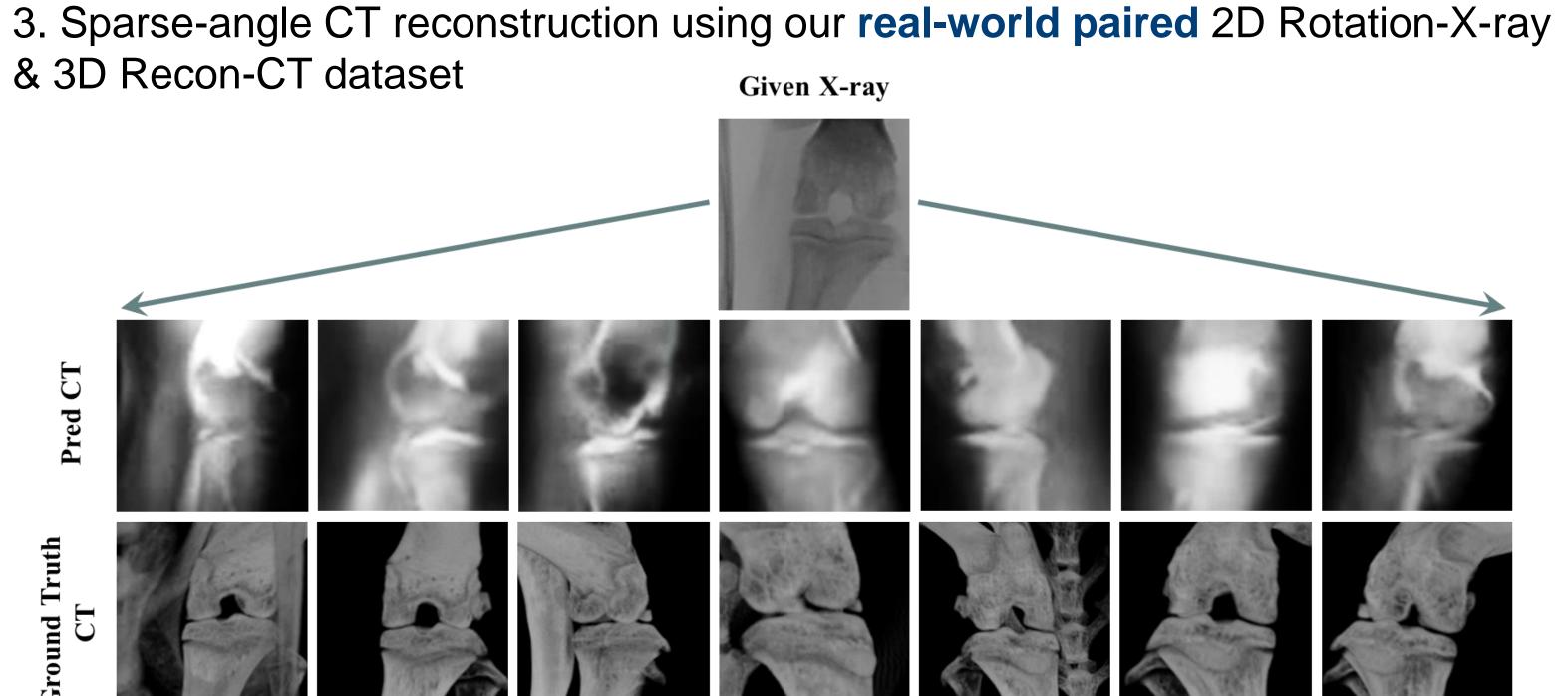


Table 2: Quantitative evaluations of sparse-angle CT reconstruction

Methods	PSNR (†)	SSIM (↑)	FID (↓)	$KID(\downarrow)$
PixelNeRF	20.2	0.740	155.2	0.128
MedNeRF	30.2	0.810	91.5	0.092
		141 1		

	Multi-projections with known angle	Ground Truth	High resolution
Exist paired X-ray/CT datasets	×	Simulated	
Ours		Real	

Clinical Relevance & Broad Impact

- Mimics disease progression. Murine model captures micro- to macro-scale metastatic evolution, filling the gap in human data of longitudinal imaging of cancer progress, and providing a testing benchmark for BCBM disease management.
- Pre-clinical relevance. Murine imaging data is validated as a surrogate for human metastasis, making our BoneMet dataset enabling sparse-angle reconstruction algorithms to convert a single human radiograph into CT for clinical applications.