

## REVIEWER'S REPORT

Manuscript No.: IJAR-53538

Date: 27/08/25

**Title:** Visual Correspondence-Based Explanations Improve Convolutional Neural Networks For Classification of Mammograms

### Recommendation:

Accept as it is .....

**Accept after minor revision...yes.....**

Accept after major revision .....

Do not accept (*Reasons below*) .....

Rating	Excel.	Good	Fair	Poor
Originality		yes		
Techn. Quality		yes		
Clarity		yes		
Significance		yes		

Reviewer Name: Dr. Shaweta Sachdeva

**Date:** 27/08/25

### Reviewer's Comment for Publication. Accepted with some minor revisions

*(To be published with the manuscript in the journal)*

*The reviewer is requested to provide a brief comment (3-4 lines) highlighting the significance, strengths, or key insights of the manuscript. This comment will be Displayed in the journal publication alongside with the reviewers name.*

### Significance

1. The work addresses an important problem in **medical imaging and breast cancer detection**, where early and accurate diagnosis is crucial for patient survival.
2. It contributes to the growing field of **Explainable Artificial Intelligence (XAI)** in healthcare, where interpretability is essential for clinical adoption.
3. The study leverages **large-scale real-world mammography data** (15,040 images from Vietnam National Cancer Hospital), which increases the reliability and applicability of findings.

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### Strengths

1. **Novelty in approach** – The proposed Visual Correspondence-Based Explanations (EMD-Corr) introduces patch-level interpretability that goes beyond image-level similarity.

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2. **Comparative evaluation** – By benchmarking against both ResNet-18 and kNN, the study clearly demonstrates performance trade-offs while highlighting interpretability gains.
  3. **Clinical relevance** – Unlike purely accuracy-focused works, this study emphasizes **localization of abnormal regions**, which is crucial for radiologists to trust AI outputs.
  4. **Efficiency** – The method improves upon earlier patch-based approaches by limiting analysis to the top-5 patch pairs, balancing **interpretability with computational efficiency**.
  5. **Strong dataset partitioning** – Clear separation into training, validation, and testing sets ensures reproducibility and robustness of results.
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## Key Insights

1. **kNN slightly outperforms EMD-Corr in classification accuracy**, but EMD-Corr provides superior explainability and localization—highlighting the trade-off between raw performance and interpretability.
2. **Patch-level correspondence captures fine-grained tumor features** that whole-image methods might overlook, aligning AI decisions more closely with radiologists' annotations.
3. The relatively high **tumor finding rate (up to 85.62%)** shows that even with limited patch pairs, the model identifies meaningful regions relevant to diagnosis.
4. The framework demonstrates how **prototype-based and correspondence-based learning** can be integrated into medical AI for more trustworthy and transparent predictions.
5. The research bridges the gap between **black-box CNNs and clinically usable AI systems**, paving the way for broader acceptance of XAI in medical imaging.

## *Detailed Reviewer's Report*

1. Correct “*Clasification*” → “*Classification*” in the title. Ensure uniform spelling of “*ResNet-18*” (sometimes written as *Resnet 18* or *Resnet-18*). Replace “*Beural*” with “*Neural*” in keywords.
2. Make captions more descriptive. For example:  
“*Figure 5: Comparison between patch locations (left) versus doctor's annotation (right)*” →  
“*Figure 5: Visual comparison of EMD-Corr identified patch locations with radiologist annotations of malignant regions.*”
3. Streamline sentences for clarity. Example:  
“*Our visual correspondence model (EMD-Corr) improves kNN and ResNet-18 where EMD-Corr slightly underperforms kNN model*” →  
“*Our EMD-Corr model outperforms ResNet-18 and achieves performance close to kNN.*”

# International Journal of Advanced Research

Publisher's Name: Jana Publication and Research LLP

*www.journalijar.com*

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4. Provide clarity on patch size ( $64 \times 64$ ) earlier in methods section for readers unfamiliar with patch-based approaches. Briefly explain *why* only top-5 patch pairs were chosen (trade-off between efficiency and accuracy).
5. Ensure all equations (cosine similarity, IoU) are numbered consistently (Eq. 1, Eq. 2, etc.). Check subscripts and notation (e.g.,  $f(Q)$  vs  $f(Q)$  spacing).
6. In Table 3, check consistency of totals (e.g., Benign count differs between earlier Table 1 and Table 3). Provide a clarifying note if due to filtering.
7. Highlight that while kNN slightly outperforms EMD-Corr in classification, **EMD-Corr offers interpretability and localization**, which is the unique strength of your method.