

# Towards Robust Natural-looking Mammography Lesion Synthesis On Ipsilateral Dual-views Breast Cancer Analysis

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## PROBLEMS

- Multi-view approaches in mammogram-based breast cancer classification face conflict between breast sides on cancerous information, impacting **outcome reliability**.
- Data insufficiency.
- Mammogram synthesis and augmentation techniques like MixUp and CutMix can lead to label conflicts and unreliable samples.

## MAIN CONTRIBUTIONS

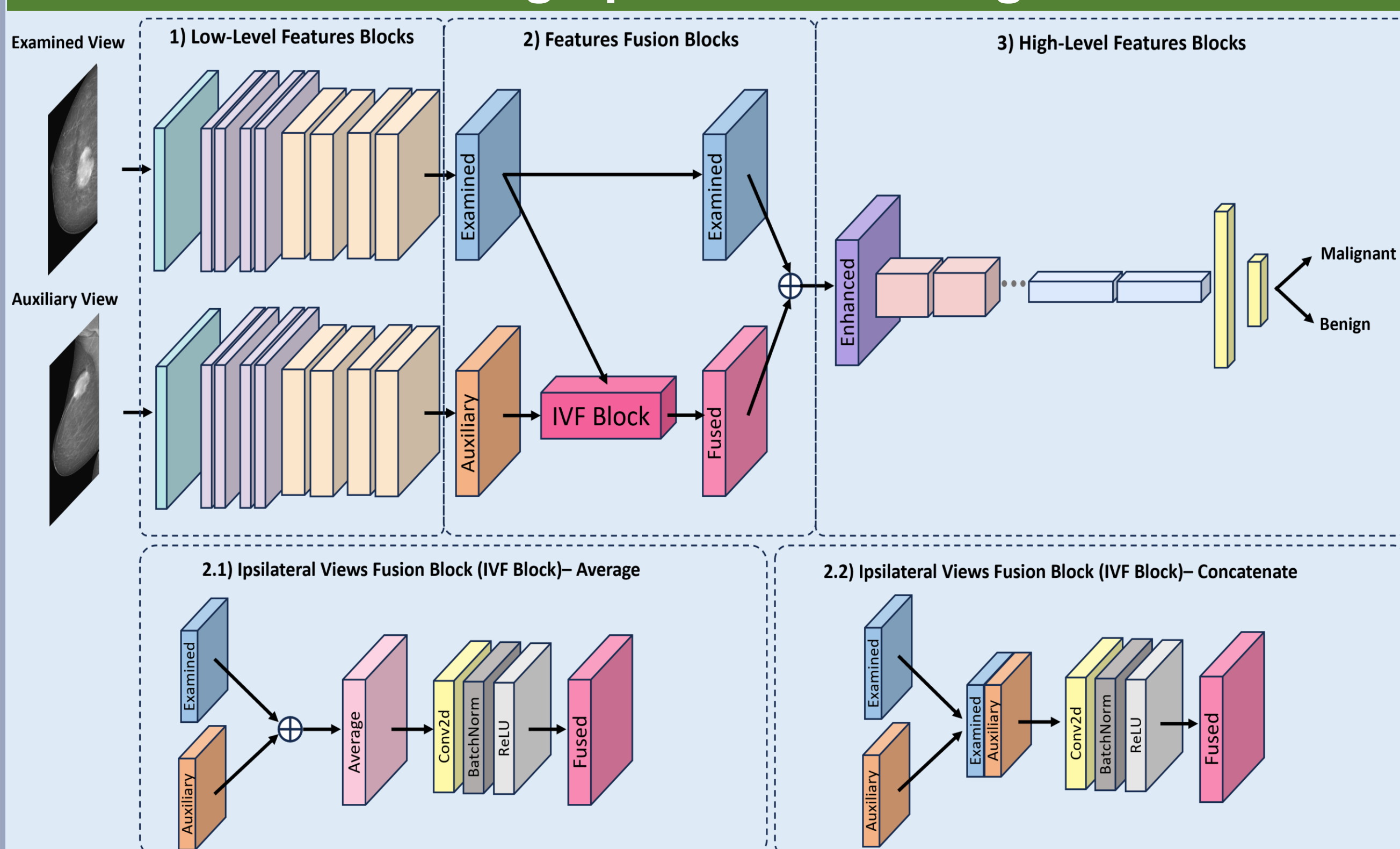
- DIVF-Net**: A multi-view network with **dual fusion operations** leveraging mammogram CC and MLO views for breast cancer classification.
- A new robust mammogram synthesis **framework** replacing benign to malignant regions with smoothed informative patches, employing Fourier Domain Adaptation.

## FUTURE DIRECTIONS

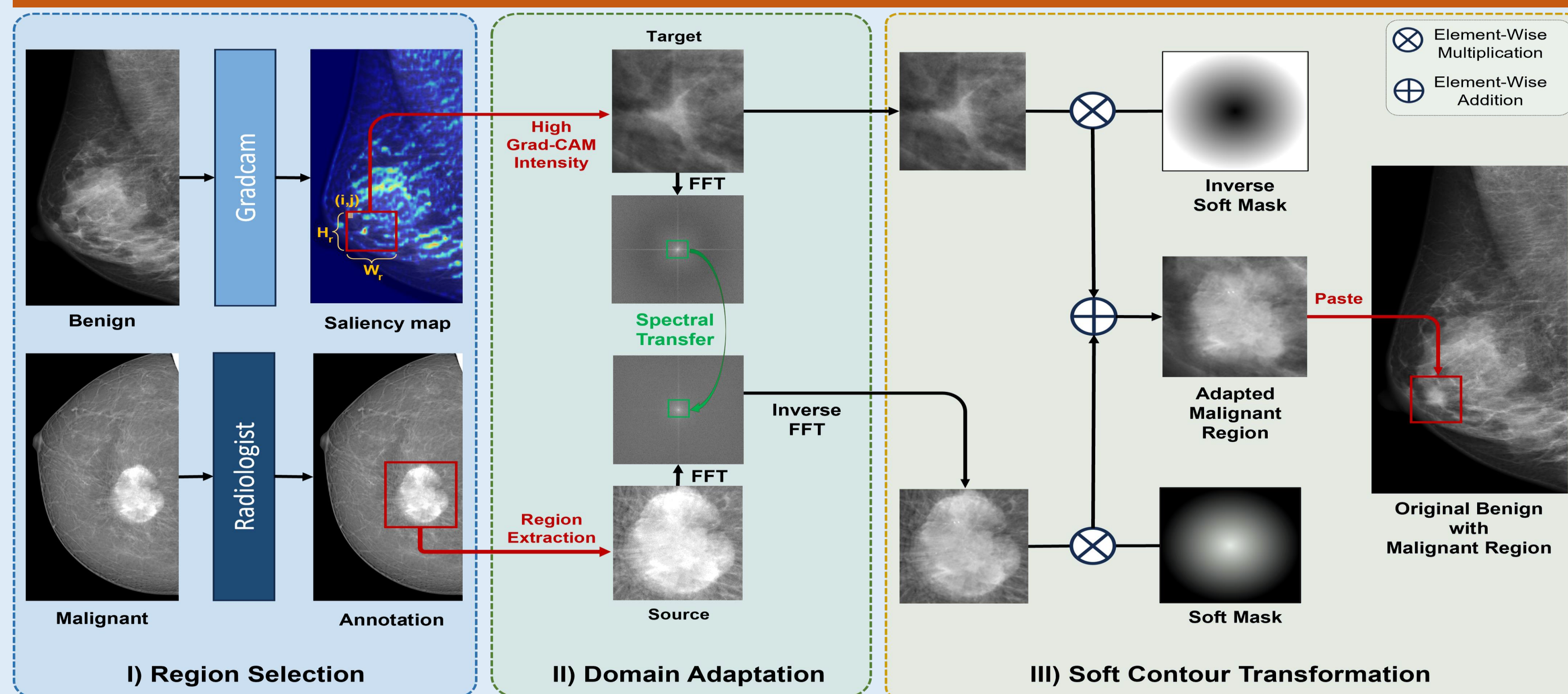
- Breast density classification.
- BI-RADS classification.
- Lesion detection.

## METHODOLOGY AND EXPERIMENTAL RESULTS

### Dual Ipsilateral Views Fusion Network (DIVF-Net) for Mammographic Cancer Diagnosis.



### Proposed Soft-Adapted Malignancy Synthesis Framework



### Data Acquisition

Datasets	Samples	Training Set	Testing Set
<b>CMMD</b>	5,202 mammograms (1,775 studies)	498 benign 1,157 malignant	88 benign 205 malignant
<b>VinDr-Mammo</b>	20,000 mammograms (5,000 studies)	1,870 benign 395 malignant	467 benign 99 malignant

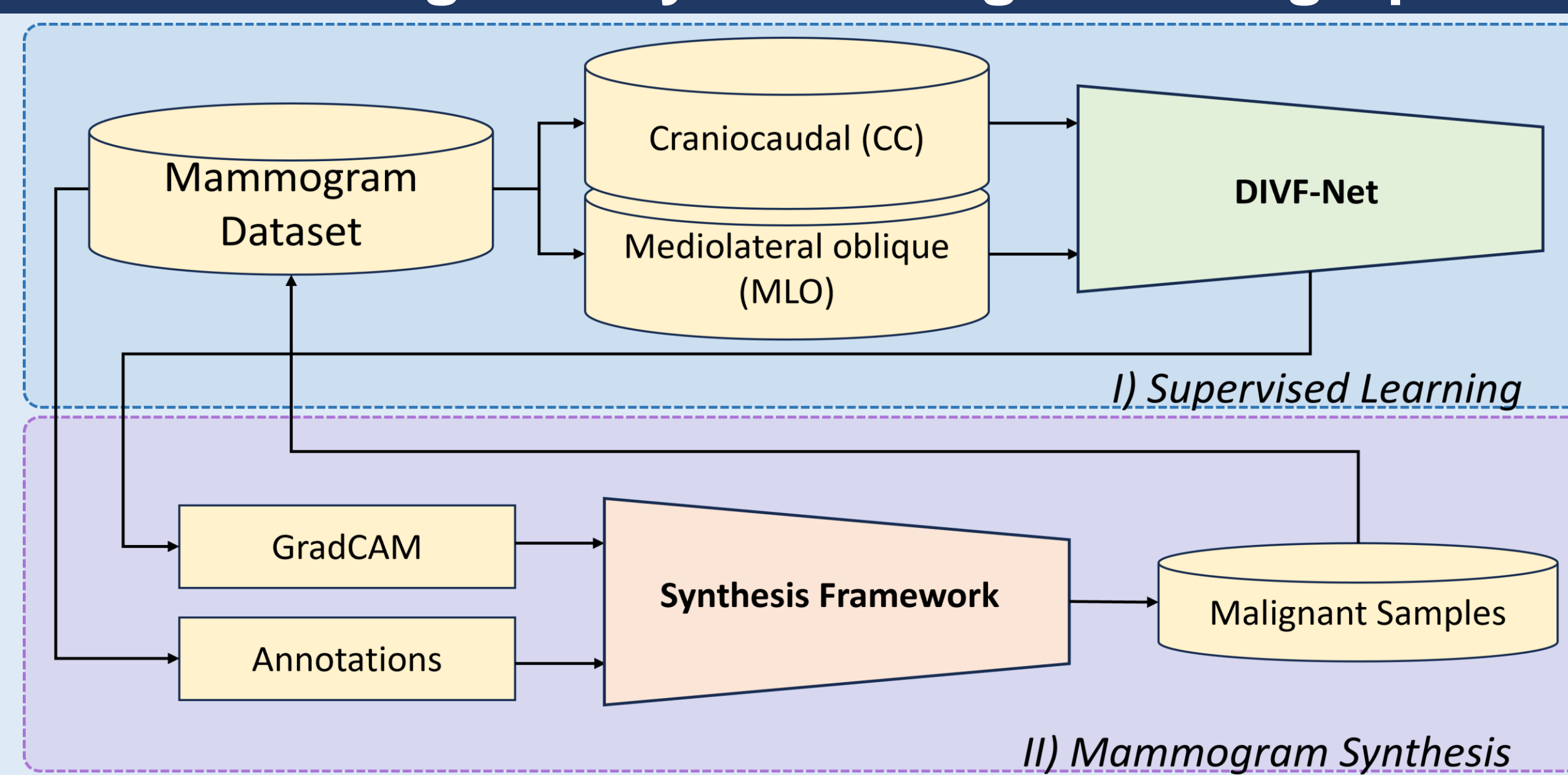
### Quantitative results (%) among our proposed DIVF-Net and other approaches.

Dataset	Backbone	Method	ResNet-18		ResNet-34	
			F1-Score	AUC-ROC	F1-Score	AUC-ROC
VinDr-Mammo	No Fusion		70.12	68.79	71.48	70.22
	Average Fusion		72.54	74.20	73.25	72.88
	Concatenate Fusion		73.22	70.66	74.63	72.18
	DIVF(Average)		74.00	72.15	74.17	71.67
	DIVF(Concatenate)		<b>75.34</b>	<b>74.24</b>	<b>75.98</b>	<b>74.86</b>
CMMD	No Fusion		73.26	76.70	75.52	77.18
	Average Fusion		79.22	79.13	79.97	81.80
	Concatenate Fusion		75.86	77.10	78.12	77.67
	DIVF(Average)		<b>81.45</b>	<b>84.14</b>	<b>82.44</b>	80.92
	DIVF(Concatenate)		77.77	80.42	79.51	<b>81.97</b>

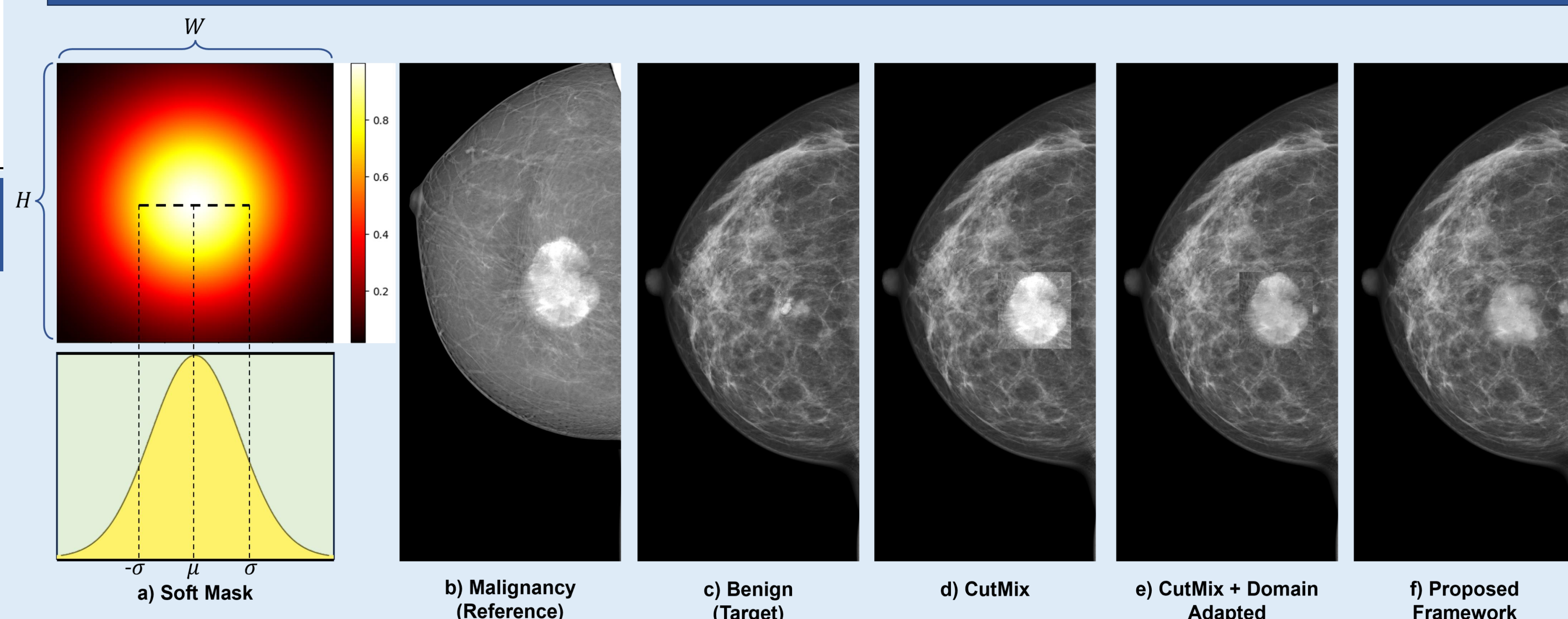
### Ablation studies of the proposed Soft-Adapted Malignancy Synthesis Framework on DIVF Concatenate+ResNet-34 on VinDr-Mammo Dataset

	DIVF	Region Selection & CutMix	Fourier Adaptation	Soft Mask	Macro F1-Score
Baseline	✓				75.98
Proposed Methods	✓	✓			76.54
	✓	✓	✓		76.96
	✓	✓		✓	76.78
	✓	✓	✓	✓	77.32

### Pipeline for Training and Synthesizing Mammographic Images



### Comparison of mammogram images synthesized using other algorithms with our proposed framework



## REFERENCES

- Liu et al. (2021) IEEE Transactions on Pattern Analysis and Machine Intelligence. 2021 Jun 1;44(10):5947-61.
- Yang, Y., & Soatto, S. (2020). Proceedings of the IEEE/CVF conference on computer vision and pattern recognition (pp. 4085-4095).

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