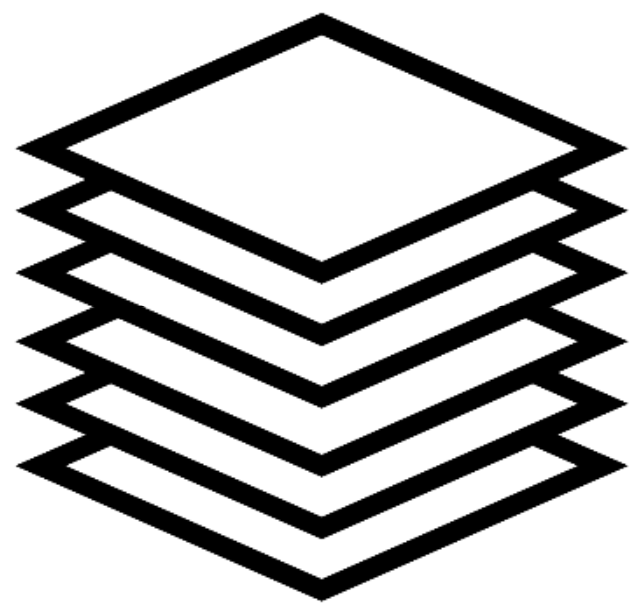


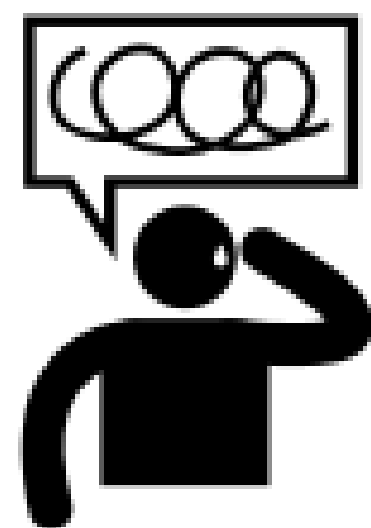
# MASS SEGMENTATION IN MAMMOGRAMS

## A CROSS-SENSOR COMPARISON OF DEEP AND TAILORED FEATURES

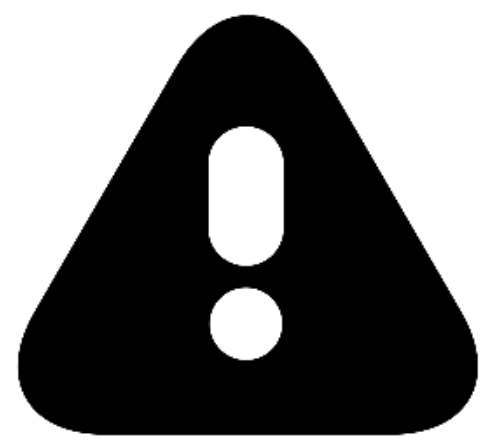
### The need for CAD in Breast Cancer Screening



Large no. of mammograms to be analyzed every day



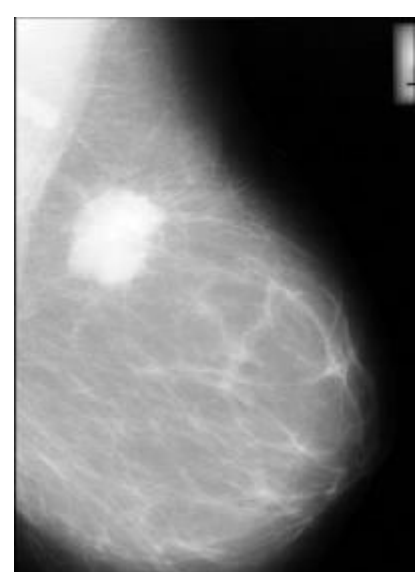
Exhaustive task, mammograms have low contrast



Prone to human errors / missing vital clues

Radiologists error rates are of **10%** to **30%** for detection of breast lesions in screening mammograms.

- **False Positive cases:** women undergo further unnecessary clinical evaluation or breast biopsy, which can lead to needless anxiety.
- **False Negative Cases:** the best time interval for the treatment of cancer can be missed, thus potentially endangering the patient.

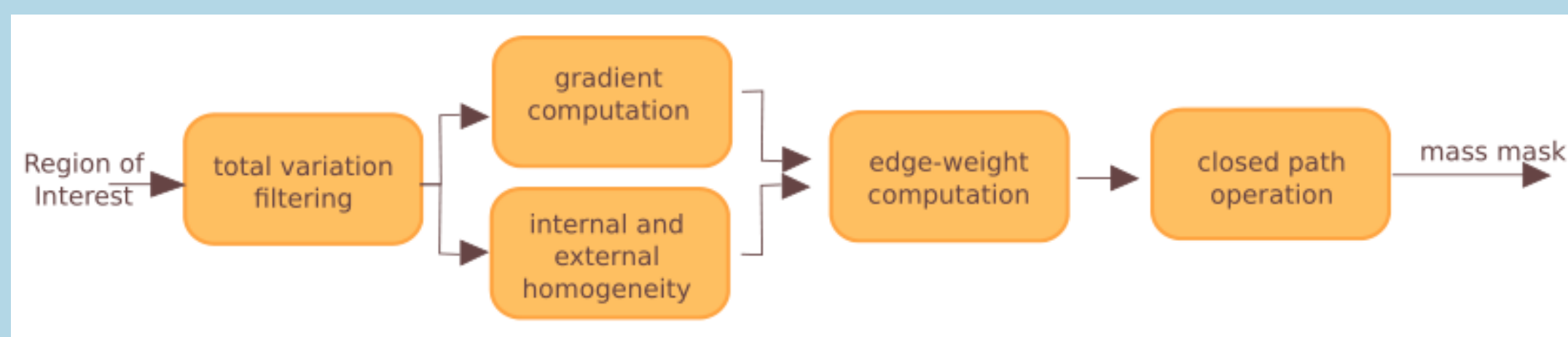


A fundamental stage in typical CAD systems is the **segmentation of masses** in regions of interest (ROIs)

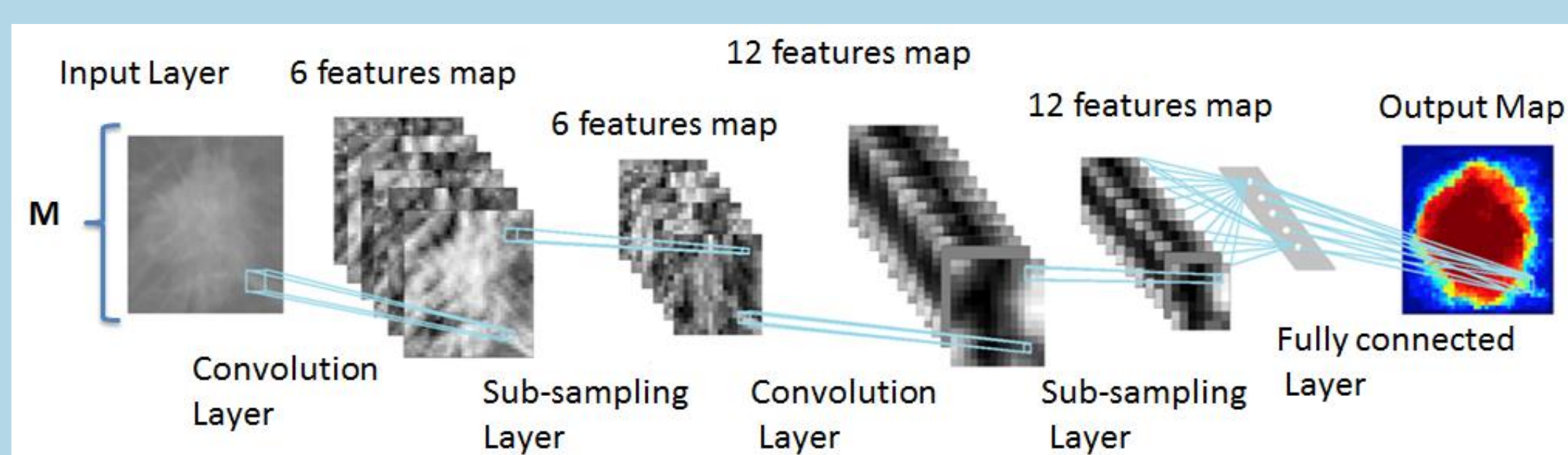
### Limitations of Current CAD Approaches

- Evaluated in Small Datasets
- Optimistic estimation of performance

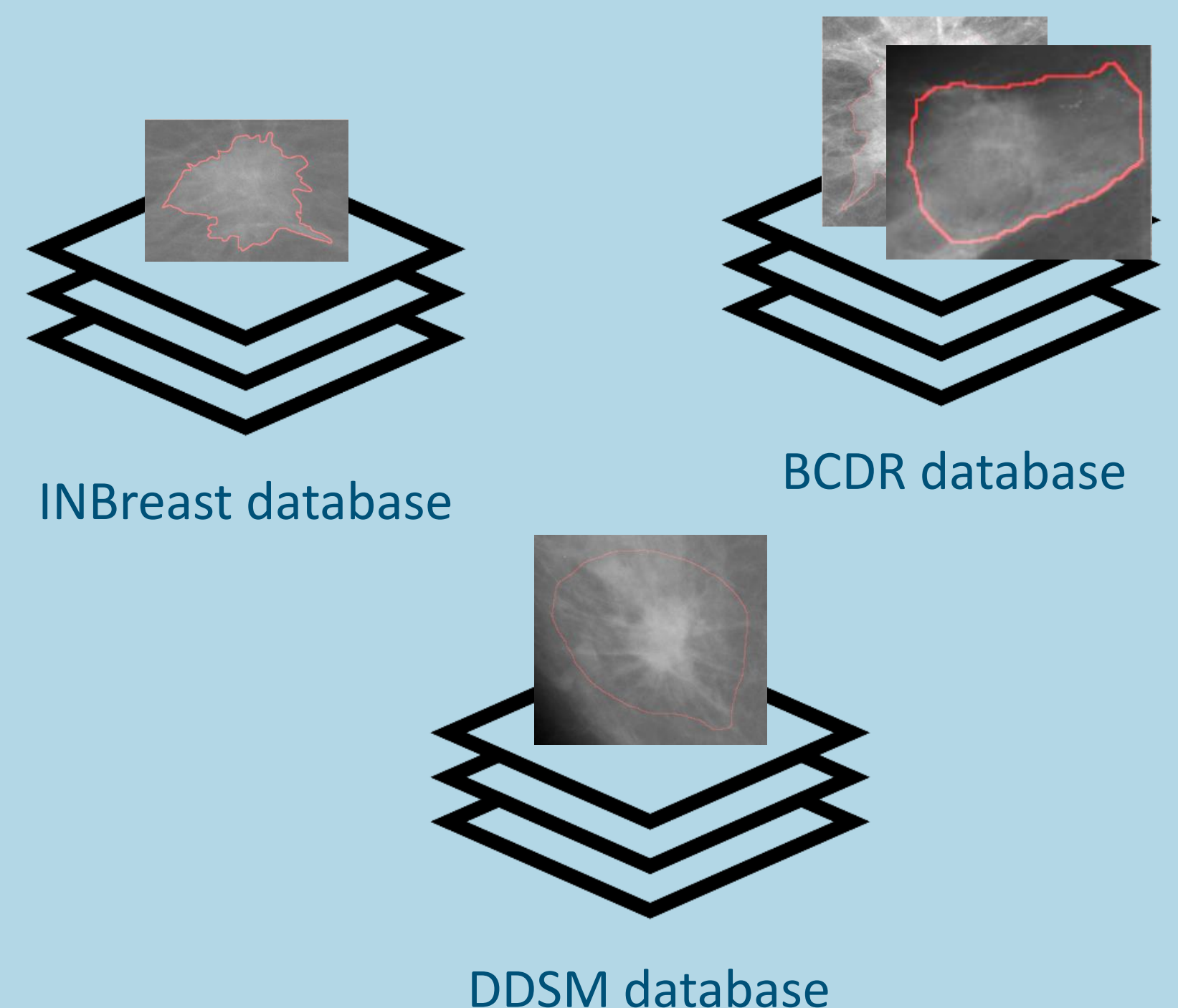
### Material and Methods



Block diagram of the **improved closed path approach**



Block diagram of the **deep learning based approach**



### Results

**Table 1:** Mass segmentation on Mammograms: Intra-sensor results. Results are the mean of the Dice metric (the higher the better).

| Database  | Original Closed Path | Improved Closed Path | SSVM | CRF  |
|-----------|----------------------|----------------------|------|------|
| INBreast  | 0.88                 | 0.89                 | 0.90 | 0.90 |
| BCDR-D01  | 0.84                 | 0.87                 | 0.88 | 0.89 |
| BCDR-F02  | 0.72                 | 0.77                 | 0.83 | 0.82 |
| DDSM-BCRP | 0.52                 | 0.87                 | 0.90 | 0.90 |

**Table 2:** Mass segmentation on Mammograms: Cross-sensor results. Results are the mean of the Dice metric (in brackets is the decrease from the intra-sensor performance).

| Train Database | Test Database | Improved Closed Path | SSVM        | CRF         |
|----------------|---------------|----------------------|-------------|-------------|
| BCDR-D01       | INBreast      | 0.89 (0.00)          | 0.82 (0.08) | 0.81 (0.09) |
| BCDR-F02       | INBreast      | 0.83 (0.06)          | 0.88 (0.02) | 0.87 (0.03) |
| DDSM-BCRP      | INBreast      | 0.83 (0.06)          | 0.87 (0.03) | 0.87 (0.03) |
| INBreast       | BCDR-D01      | 0.87 (0.00)          | 0.82 (0.06) | 0.81 (0.08) |
| BCDR-F02       | BCDR-D01      | 0.84 (0.03)          | 0.80 (0.08) | 0.79 (0.10) |
| DDSM-BCRP      | BCDR-D01      | 0.84 (0.03)          | 0.84 (0.04) | 0.83 (0.05) |
| INBreast       | BCDR-F02      | 0.75 (0.02)          | 0.77 (0.06) | 0.80 (0.02) |
| BCDR-D01       | BCDR-F02      | 0.75 (0.02)          | 0.77 (0.06) | 0.76 (0.06) |
| DDSM-BCRP      | BCDR-F02      | 0.77 (0.00)          | 0.81 (0.02) | 0.81 (0.01) |
| INBreast       | DDSM-BCRP     | 0.65 (0.22)          | 0.77 (0.12) | 0.81 (0.09) |
| BCDR-D01       | DDSM-BCRP     | 0.65 (0.22)          | 0.83 (0.07) | 0.81 (0.09) |
| BCDR-F02       | DDSM-BCRP     | 0.87 (0.00)          | 0.85 (0.05) | 0.83 (0.07) |

### Discussion and Conclusions

- Improved Closed Path is much better than the original method
- The worst performances are obtained when transferring from INBreast to DDSM and from BCDR-D01 to BCDR-F02.
  - One of the reasons behind this performance drop lies in the annotation differences between those databases.
- The results improve from the film based to the digital mammography
  - the higher data quality of the digital mammograms pays off in the segmentation task.
- The fine-detailed segmentation of the (digital) INBreast database yields the best automatic segmentation model.