

# Multi-view analysis of unregistered medical images using cross-view transformers

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## Motivation

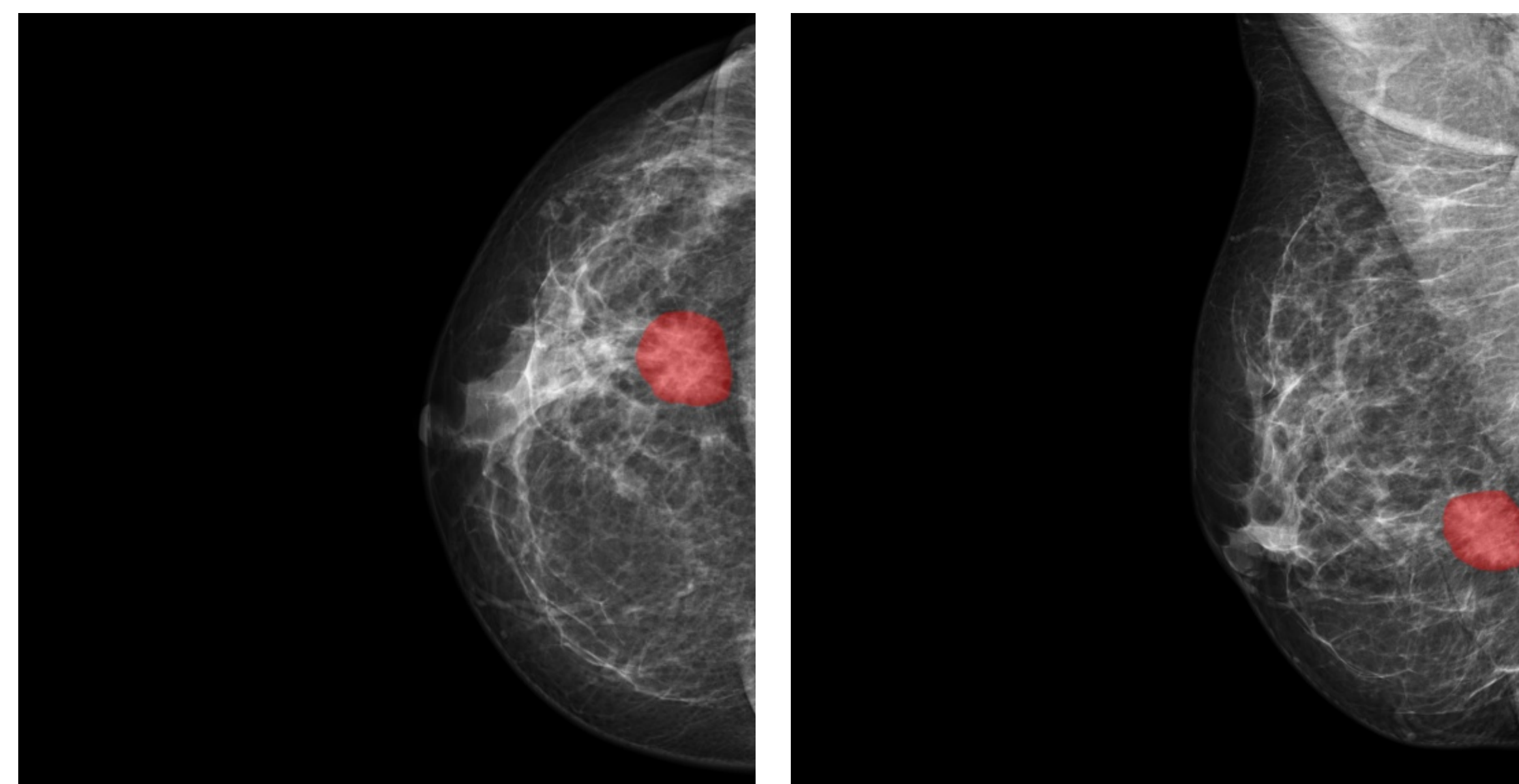
Multi-view medical image analysis often depends on the combination of information from multiple views.

Differences in perspective and other forms of misalignment can make it difficult to combine views effectively.

## Data

CBIS-DDSM mammography images with CC and MLO views.

Breast tumor classification benefits from information from both views.



Craniocaudal (CC)

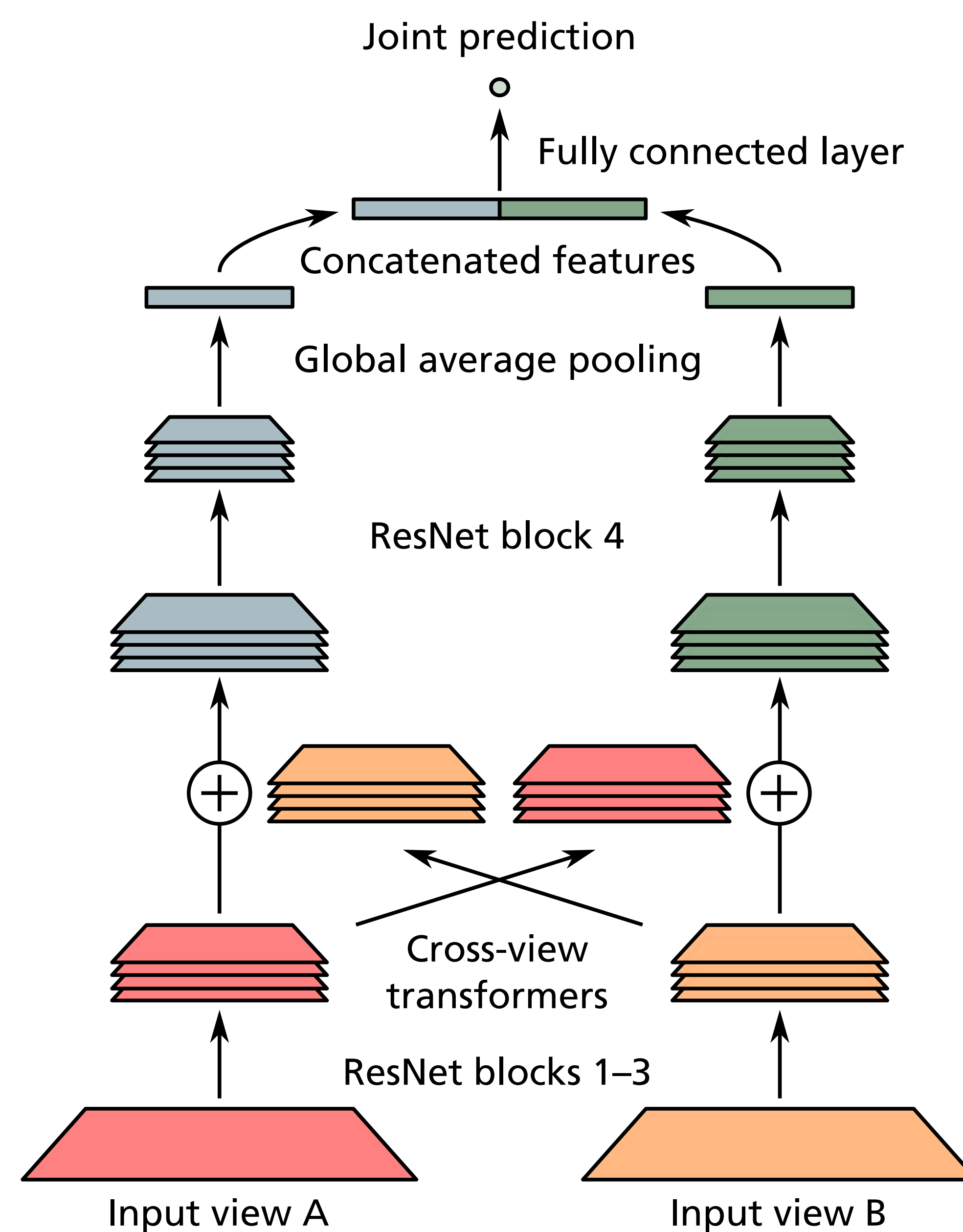
Mediolateral-oblique (MLO)

## Method

A cross-view transformer that can transfer features between unregistered views at the level of spatial feature maps.

## Cross-view transformer

Link views and transfer features at an intermediate, spatial level.



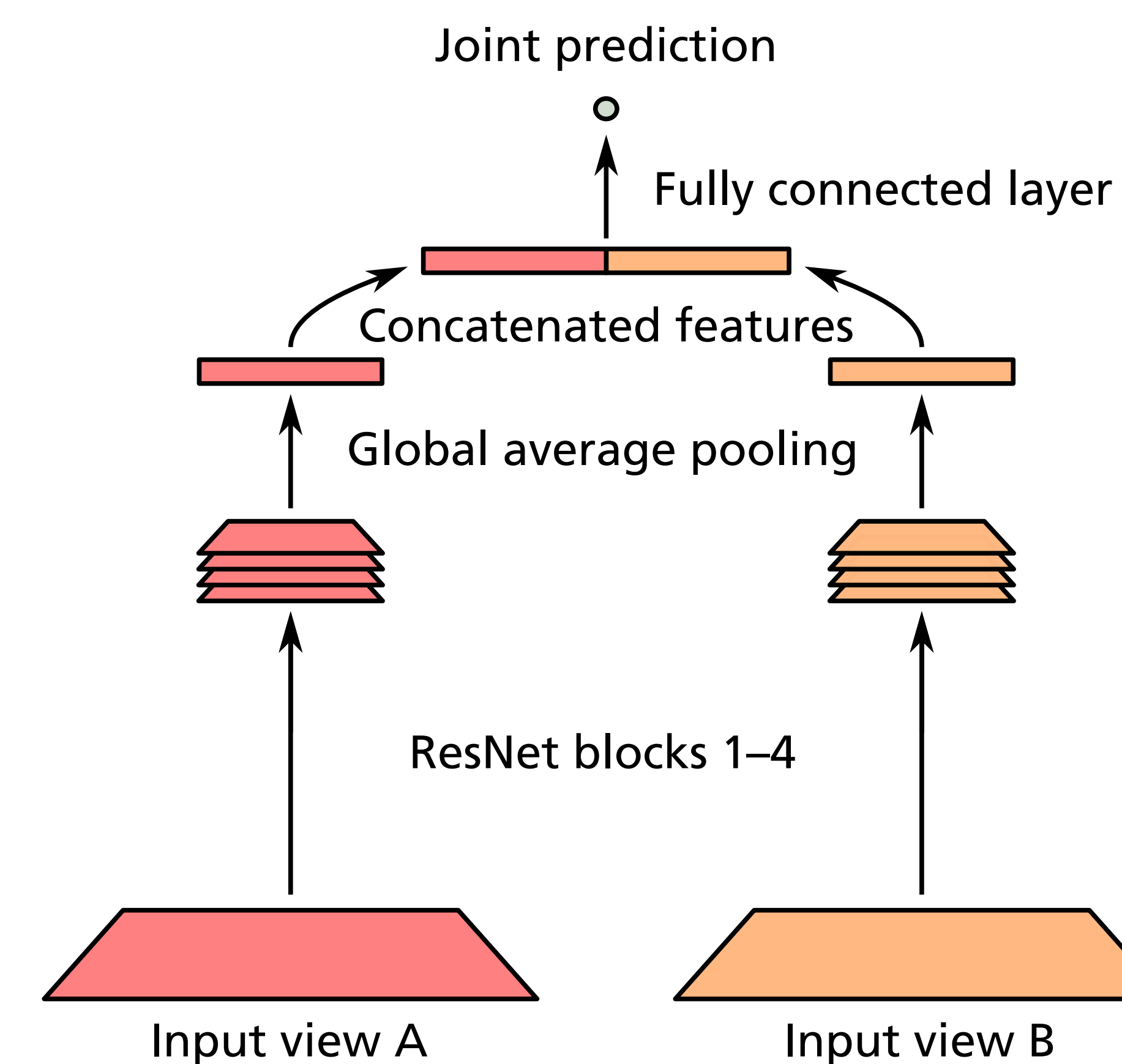
## Results for CBIS-DDSM breast tumor classification

Model	Views	ROC-AUC $\pm$ std.dev.	p-value
Single view	CC	0.750 $\pm$ 0.007	0.005
	MLO	0.763 $\pm$ 0.003	0.036
Late join	CC + MLO	0.788 $\pm$ 0.008	
Cross-view transformer (tokens)	CC + MLO	0.803 $\pm$ 0.007	0.061
Cross-view transformer (pixels)	CC + MLO	0.801 $\pm$ 0.003	0.006

Table 1: Area under the ROC curve for the CBIS-DDSM dataset. Mean and standard deviation computed over three runs. p-values for a two-sided Wilcoxon signed-rank test against the late-join baseline model.

## Baseline: Late join network

Merge information from views only after global pooling.



## Additional experiments

CheXpert chest X-ray with frontal and lateral views.

## Conclusion

Combining multi-view information on a spatial level can outperform merging features after global pooling.

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