



# Learning to Zoom with Anatomical Relations for **Medical Structure Detection**

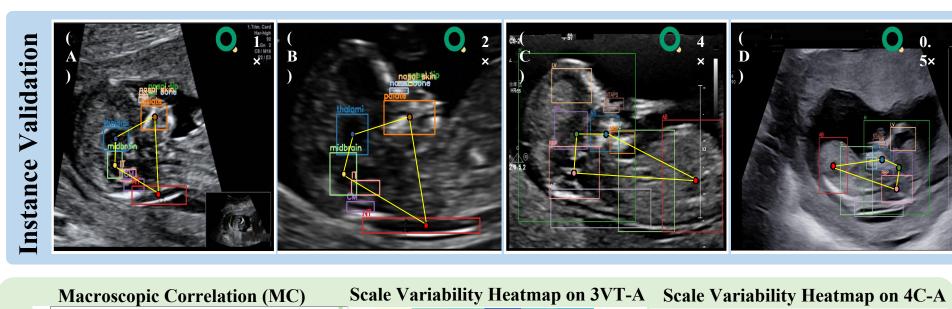


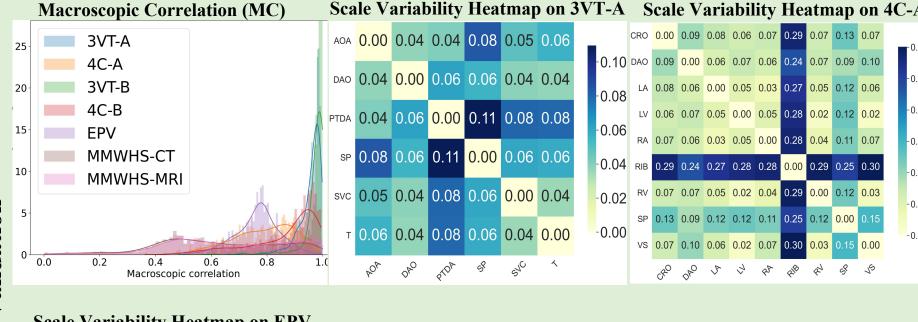
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### **Problem & Motivation**

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- Anatomical structures in medical images exhibit significant scale variability.
- Current algorithms **neglect zoom information** and structural topological priors.
- Models struggle with morphological variability, structural inconsistency, and detection uncertainty.





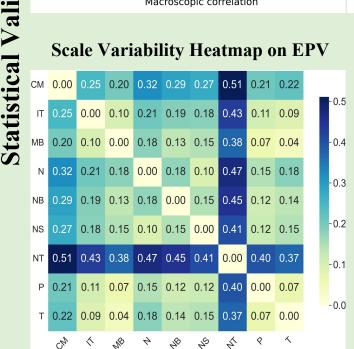


Figure 1: Motivation of our porposed ZR-DETR. The upper row illustrates the observed structural relation consistency and zoom patterns in our proposed ZR-DETR framework. The second row provides empirical validation through visualizations of Macroscopic Correlation (MC) across the employed datasets and category-wise scale variability heatmaps.

### **Our Contribution**

- 1 Introduced Zoom Relation Encoder and Anatomical Relation Constraints for scale-aware structural priors.
- (2) Proposed a Gaussian Process probabilistic detection framework for uncertainty quantification.
- 3 Significantly outperformed baselines in single-domain and domain adaptation scenarios.

# Methodology:ZR-DETR

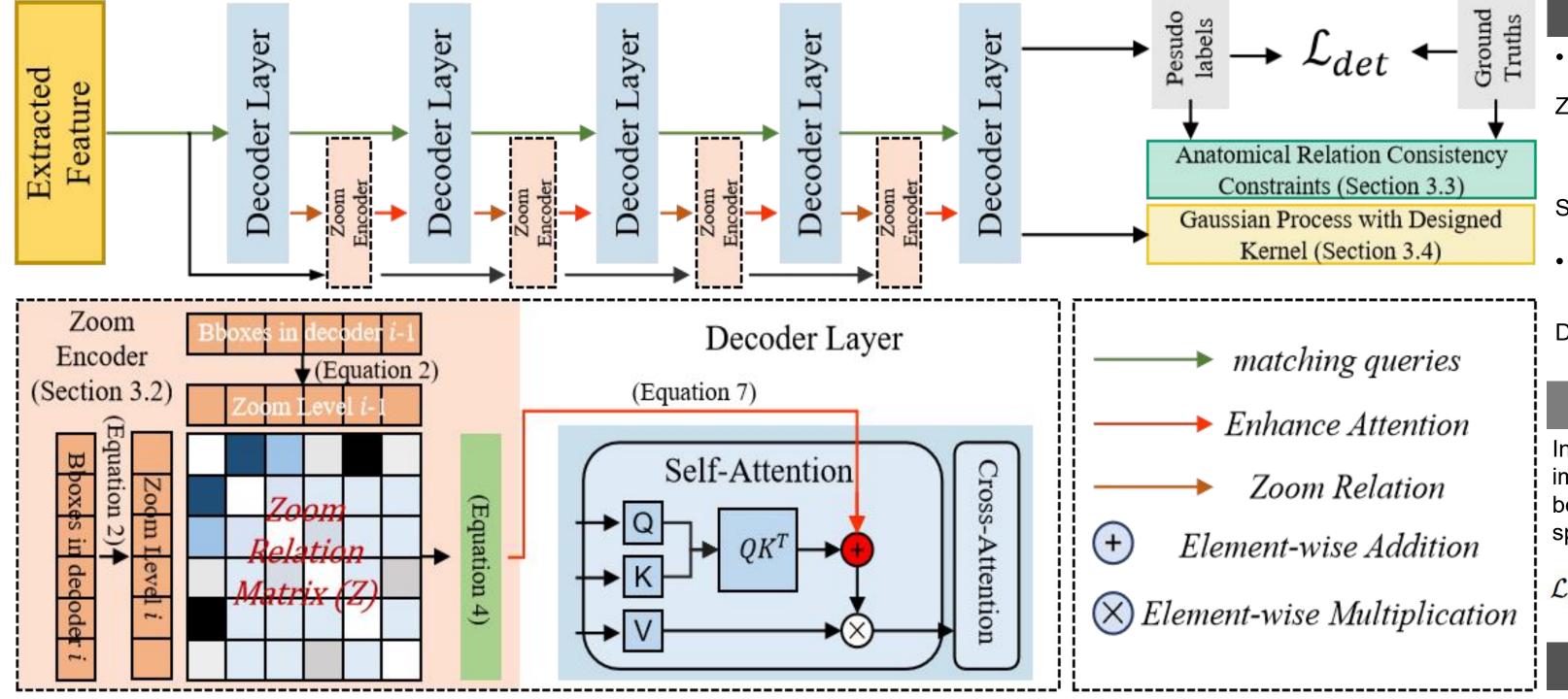


Figure 2: ZR-DETR integrates a Transformer-based encoder-decoder architecture, incorporating Zoom Embeddings to process multi-scale medical image features, Anatomical Relation Consistency Constraints Gaussian Process sample, using a composite kernel function to encode prior anatomical knowledge, and a Gaussian Process for uncertainty-aware detection.

#### **Zoom Relation Encoder**

Capture relative organ scale relationships

Zoom Relation Matrix(
$$Z$$
):  $\lambda_i = \log\left(\frac{A_i}{A_{\mathrm{img}}}\right), \;\; z_{ij} = \frac{|\lambda_i|}{|\lambda_j|}$ 

Embed $(z_{ij}, 2k) = \sin\left(\frac{s \cdot z_{ij}}{T^{2k/d_z}}\right)$ 

Scalar to high-dim vector:

Embed $(z_{ij}, 2k+1) = \cos\left(\frac{s \cdot z_{ij}}{T^{2k/d_z}}\right)$ 

Integrates zoom embeddings into attention

 $\operatorname{Attn}(Q_m^l) = \operatorname{Softmax}\left(\frac{Q_m Q_m^{\top}}{\sqrt{d_{\text{model}}}} + \operatorname{Zoom}(b^{l-1}, b^l)\right) Q_m,$ Dual-branch decoder:  $\operatorname{Attn}(Q_h^l) = \operatorname{Softmax}\left(\frac{Q_h Q_h^\top}{\sqrt{d_{l+1}}}\right) Q_h.$ 

#### **Anatomical Relation Consistency Constraints**

Incorporates prior topological knowledge, penalizing implausible predictions by comparing the spatial proximity between predicted bounding boxes with the prior expected spatial proximity estimated from training data

$$\mathcal{L}_{ ext{anatomy}} = rac{1}{|\mathcal{P}|} \sum_{i=1}^{T} \sum_{j=1}^{T} \mathbf{M}_{a_i,a_j} \left\| k_{ ext{spatial}}(\mathbf{b}_i, \mathbf{b}_j) - k_{ ext{prior}}(a_i, a_j) 
ight\|_2$$

#### **Probabilistic Detection with Gaussian Processes**

Probabilistic Detection with Gaussian Processes quantifies predictive uncertainty by modeling each detection score as a that integrates visual features, zoom levels, and anatomical categories, thereby providing both predictive mean and variance

## **Experimental Results**

#### Results on **FCS** dataset

Table 1: The performance of different detection methods in FCS Table 2: The performance of different detection methods in FCS (3VT) dataset [9]. (4C) dataset [9].

Method	LA ↑	RA↑	LV↑	RV↑	VS↑	CRO ↑	SP ↑	DAO ↑	RIB↑	mAP ↑
	Single Domain Structure Detection (Site A)									
FasterRCNN [37] CVPR16	92.0	96.6	94.1	91.5	96.9	97.0	94.2	93.1	73.0	92.1
Deformable-DETR [6] ICLR21	94.5	95.6	96.0	95.2	96.7	93.1	97.5	97.4	73.5	93.3
DAB-DETR [38] ICLR22	96.5	96.7	95.6	94.5	97.5	98.5	97.5	98.0	76.8	94.6
DN-DETR [39] CVPR22	97.2	98.0	98.9	97.7	99.8	97.7	99.2	98.5	85.4	96.9
DINO [41] ICLR23	97.3	97.8	99.4	97.5	99.9	97.9	98.1	98.5	86.4	97.0
Relation-DETR [30] ECCV24	97.6	98.4	97.8	97.2	99.6	97.9	99.9	99.6	87.8	97.3
ZR-DETR (Ours)	98.1	97.9	98.9	99.6	99.9	100	99.9	99.6	87.7	98.0
	Single Domain Structure Detection (Site B)									
FasterRCNN [37] CVPR16	71.3	91.3	86.7	80.5	90.4	87.4	88.1	89.1	82.6	85.3
Deformable-DETR [6] ICLR21	75.7	87.4	90.7	85.2	91.0	85.0	90.0	91.0	84.3	86.7
DAB-DETR [38] ICLR22	75.5	92.2	92.9	88.4	94.3	86.7	92.1	92.1	86.6	89.0
DN-DETR [39] CVPR22	77.1	92.3	92.0	87.6	96.0	87.9	93.2	92.4	85.0	89.3
DINO [41] ICLR23	83.4	95.2	96.4	92.9	97.0	92.2	94.3	93.0	89.5	92.7
Relation-DETR [30] ECCV24	87.9	95.1	97.5	94.2	94.9	92.7	96.6	93.3	88.4	93.4
ZR-DETR (Ours)	89.0	95.0	97.4	95.5	96.1	94.0	96.4	94.4	90.0	94.2
Cross Domain Structure Detection (A→B)										
SIGMA [42] CVPR22	50.1	62.1	49.5	51.3	58.9	55.6	46.7	54.0	47.9	52.0
SIGMA++ [43] TPAMI 23	57.1	57.0	60.2	58.3	56.1	58.9	55.5	59.1	60.1	57.8
$M^3$ -UDA [9] CVPR24	79.9	69.8	72.8	71.7	81.0	78.0	81.7	78.0	78.3	76.8
ToMo-UDA [10] ICML24	64.2	75.6	70.4	64.3	66.7	75.0	75.5	77.2	73.0	71.3
DATR [44] TIP25	81.4	62.4	68.9	74.1	81.8	73.9	82.6	83.7	62.3	74.6
ZR-DETR-UDA (Ours)	58.4	82.2	80.8	74.6	89.5	83.2	83.4	87.4	78.2	79.8

Method	DAO ↑	SP ↑	PTDA 1	T↑	<b>SVC</b> ↑	AOA ↑	mAP 1	
	Single Domain Structure Detection (Site A)							
FasterRCNN [37] CVPR16	93.7	92.0	96.9	87.1	87.5	94.5	91.9	
Deformable-DETR [6] ICLR21	97.1	90.2	96.9	91.4	91.0	98.8	94.2	
DAB-DETR [38] ICLR22	96.5	96.3	96.8	92.0	89.0	97.9	94.7	
DN-DETR [39] CVPR22	97.2	92.7	96.7	89.5	93.7	98.8	94.8	
<b>DINO [41] ICLR23</b>	95.8	97.4	97.7	94.5	93.0	99.4	96.6	
Relation-DETR [30] ECCV24	96.5	94.0	96.5	94.8	93.0	99.7	95.8	
MI-DETR [30] CVPR25	97.8	97.4	97.7	94.5	93.0	99.4	96.6	
ZR-DETR (Ours)	98.7	96.9	98.7	95.3	94.3	99.8	97.3	
Single Domain Structure Detection (Site B)								
FasterRCNN [37] CVPR16	54.8	30.2	30.0	19.9	39.5	41.3	35.9	
Deformable-DETR [6] ICLR21	51.3	22.4	31.9	20.5	36.9	56.0	36.8	
DAB-DETR [38] ICLR22	44.6	35.7	29.6	27.2	37.8	62.4	39.6	
DN-DETR [39] CVPR22	51.9	33.4	31.0	23.4	44.9	60.0	40.8	
<b>DINO [41] ICLR23</b>	51.8	39.0	29.0	26.6	52.7	62.5	43.6	
Relation-DETR [30] ECCV24	56.9	42.2	32.6	22.5	57.9	56.1	44.7	
MI-DETR [30] CVPR25	53.2	43.2	32.7	22.3	62.4	54.8	44.8	
ZR-DETR (Ours)	57.4	39.1	33.2	28.7	73.2	61.2	48.8	
Cross Domain Structure Detection (A→B)								
SIGMA [42] CVPR22	42.9	42.8	59.4	39.6	41.7	60.0	47.7	
SIGMA++ [43] TPAMI 23	42.3	37.4	45.4	29.0	32.0	42.9	38.1	
$M^3$ -UDA [9] CVPR24	59.5	59.7	70.1	51.9	52.4	68.9	60.4	
ToMo-UDA [10] ICML24	45.4	60.1	81.5	27.6	45.6	63.7	54.0	
DATR [44] TIP25	57.4	59.2	61.7	56.9	50.7	62.7	58.1	
ZR-DETR-UDA (Ours)	61.1	67.1	67.8	60.8	49.4	67.9	62.4	

#### Results on MM-WHS dataset

#### Results on **EPV** dataset

Table 3: The performance of different detection methods in EPV dataset [33].

Method		NB ↑	P ↑	NS ↑	NT ↑	MB↑	NT ↑	IT ↑	CM ↑	mAP
	Single Domain Structure Detection									
FasterRCNN [37] CVPR16	98.5	80.2	97.0	62.1	86.0	98.3	66.4	94.7	73.5	84.1
Deformable-DETR [6] ICLR21	98.6	87.3	95.5	81.5	92.3	99.6	73.0	95.2	78.2	89.0
DAB-DETR [38] ICLR22	97.7	93.0	96.4	85.5	92.2	98.4	82.8	94.4	77.5	90.9
DN-DETR [39] CVPR22	98.7	93.7	97.3	85.2	94.6	97.4	83.1	96.7	79.1	91.7
<b>DINO [41] ICLR23</b>	98.3	92.3	97.1	86.8	93.8	99.8	82.4	94.6	81.5	91.8
Relation-DETR [30] ECCV24	98.7	94.9	96.7	85.9	95.2	98.3	84.8	96.0	87.8	93.1
MI-DETR [40] CVPR25	98.9	94.6	98.6	89.2	94.5	100	87.2	94.2	84.2	93.5
ZR-DETR(Ours)	99.0	95.1	98.4	91.6	95.4	100	90.1	94.6	85.2	94.4

#### **Computational Efficiency**

Table 6: Comparison of computational efficiency across different detection methods.

Metric	Method										
	FasterRCNN [37]	Deformable-DETR [6]	DAB-DETR[38]	<b>DN-DETR</b> [39]	<b>DINO</b> [41]	Relation-DETR [30]	MI-DETR [40]	ZR-DETR (Ours)			
Times (ms) ↓	41.405	40.101	41.558	46.133	42.102	46.2	55.6	45.172			
Params (M) ↓	41.1	65.2	48.5	88.5	90.6	111.2	138.1	115.4			
<b>TFlops</b> ↓	0.192	0.18	0.0874	0.244	0.256	0.258	0.292	0.268			

Code & data: https://github.com/LiwenWang919/ZR-DETR

Please contact pubin@hnu.edu.cn for further information.

