



北京航空航天大学
BEIHANG UNIVERSITY



Viewport Proposal CNN for 360° Video Quality Assessment

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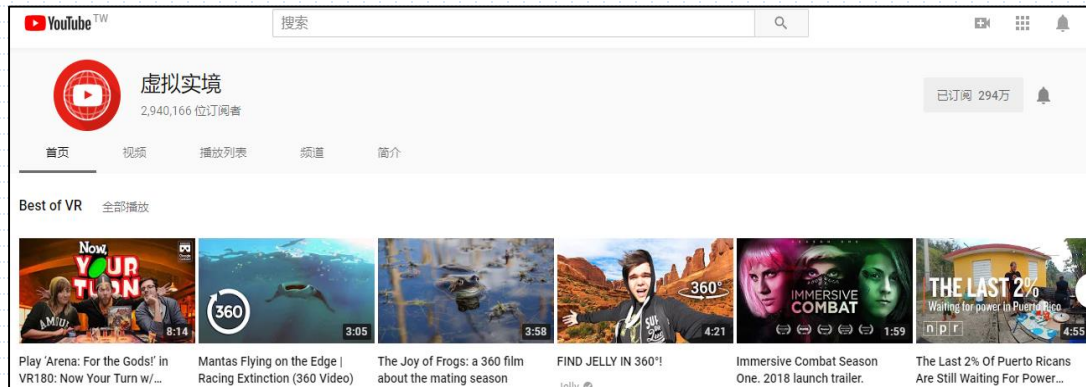
Collaborated with Chen Li, Lai Jiang, Shanyi Zhang and Xiaoming Tao

Outline

- ✚ **Background**
- ✚ **Related Works**
- ✚ **Our V-CNN Approach**
- ✚ **Experimental Results**

Background

- Most recently, 360° video has become part of our daily life.



Amount of 360° video on YouTube: **100 million+**



Live streaming of **CCTV Spring Festival Gala** in
360° format

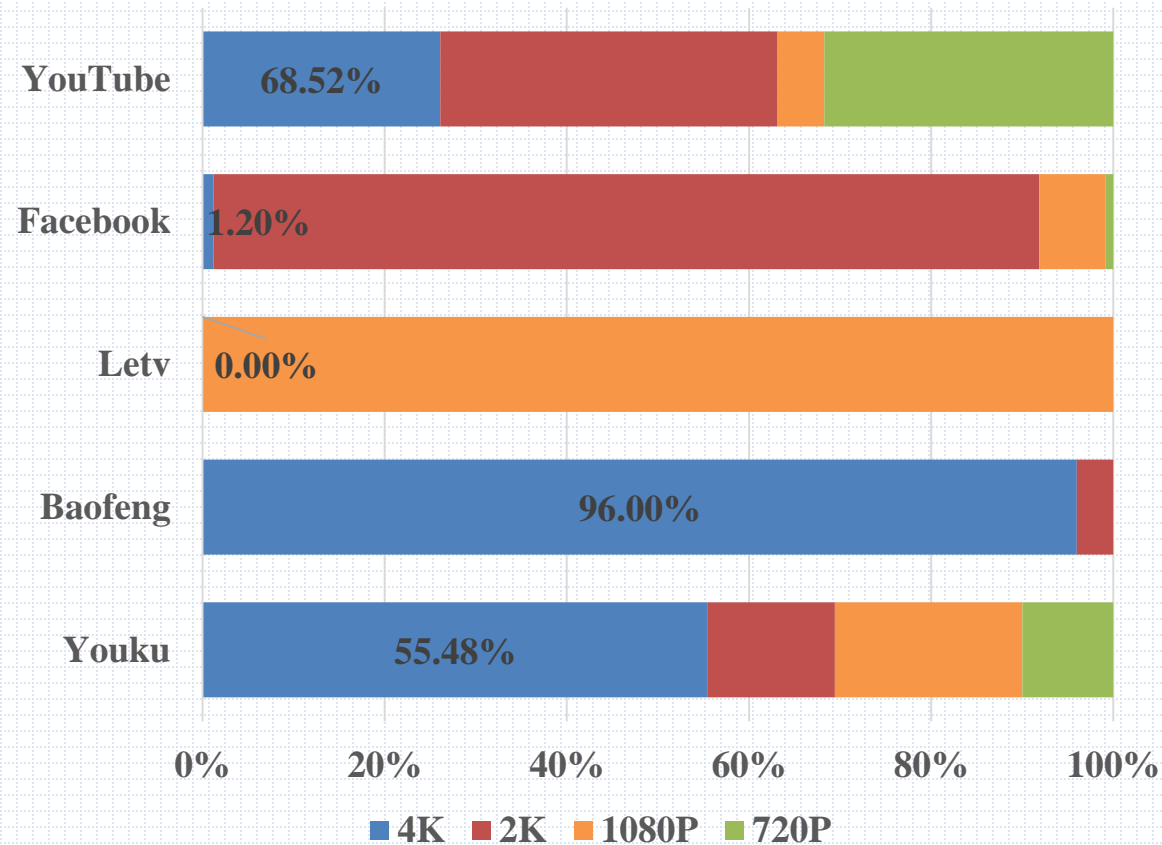
Video Demo: **Skydiving**



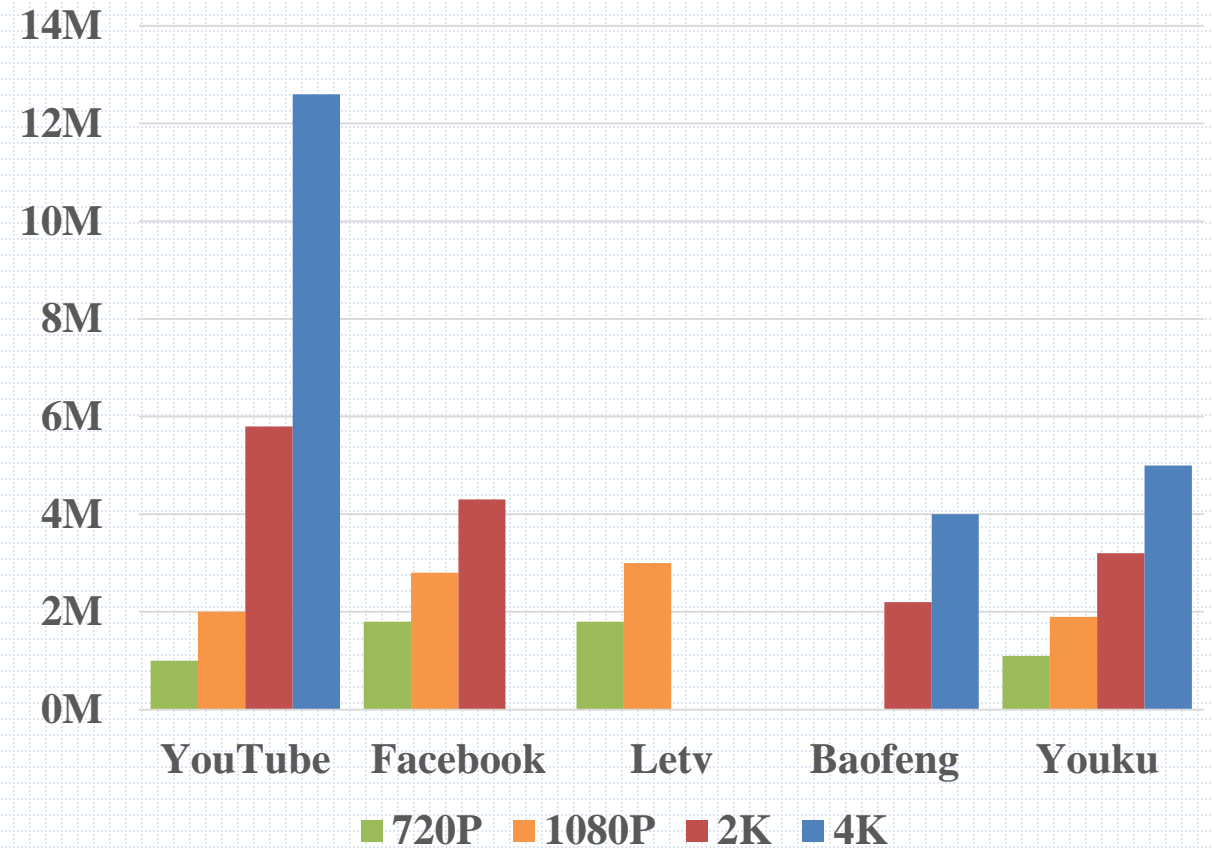
Background

- However, the source of 360° video varies.

Resolution of 360° video on several platforms



Bitrate of 360° video on several platforms



Background

- Meanwhile, the quality degradation in the viewport **is more noticeable.**



Thus, it is necessary to study **visual quality assessment (VQA)** for 360° video.

Related Works

➤ VQA on 360° video

Non-Deep-Learning : S-PSNR (Yu *et al.*, 2015), CPP-PSNR (Zakharchenko *et al.*, 2016), WS-PSNR (Sun *et al.*, 2017), AW-SPSNR (Xiu *et al.*, 2017).

Deep-Learning: VR-IQA-NET (Lim *et al.*, 2018).

No work considers that the visual quality highly depends on the viewports in 360° video.

←-----→

➤ Attention models on 360° video

HM prediction: Cheng *et al.*, 2018, Spherical U-Net (Zhang *et al.*, 2018), DHP (Xu *et al.*, 2019), etc.

EM prediction: Ling *et al.*, 2018, Xu *et al.*, 2018, etc.

No existing 360° VQA work benefits from the auxiliary tasks of attention modelling.

←-----→

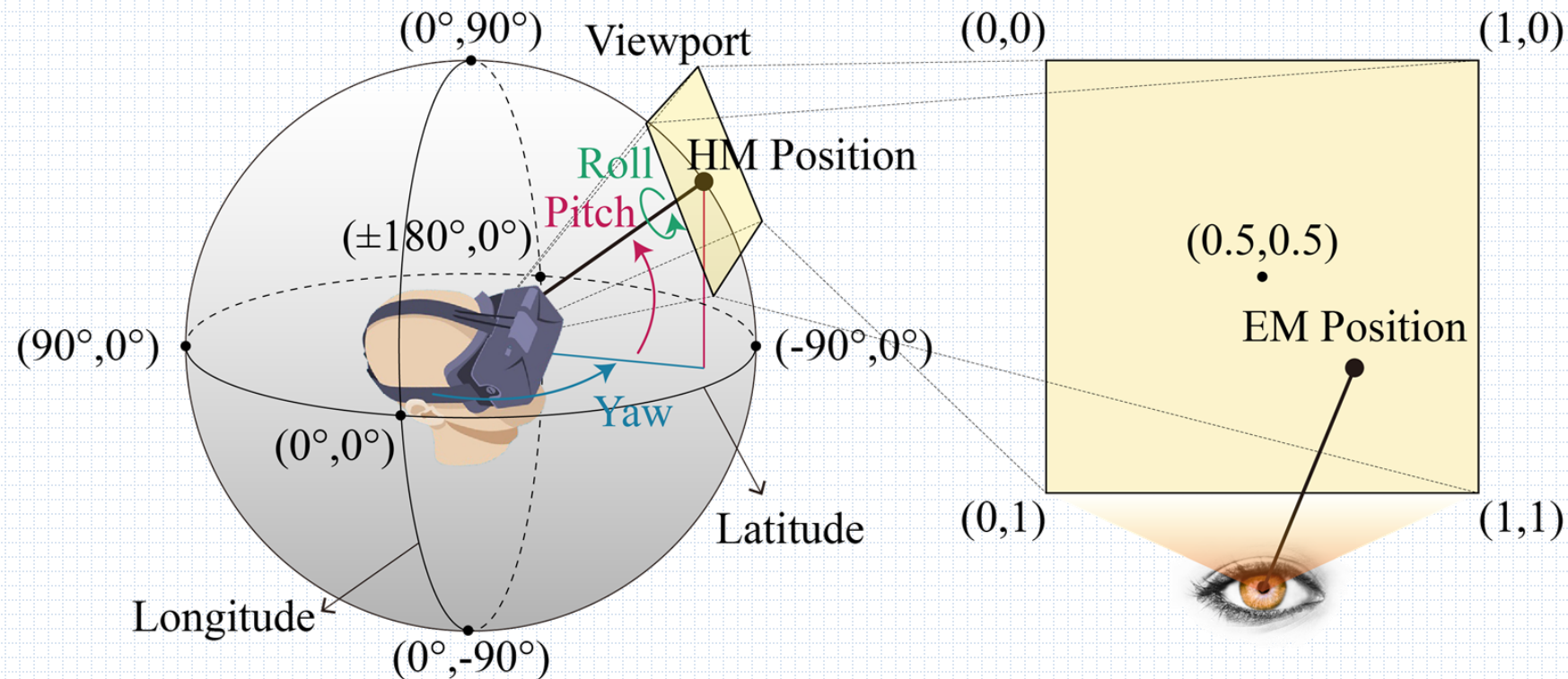
➤ Great success of **R-CNN** approaches in object detection

R-CNN (Girshick *et al.*, 2014), Fast R-CNN (Girshick *et al.*, 2015), Faster R-CNN (Ren *et al.*, 2017), Mask R-CNN (He *et al.*, 2017)

Two staged, multi-task approaches with region proposal.

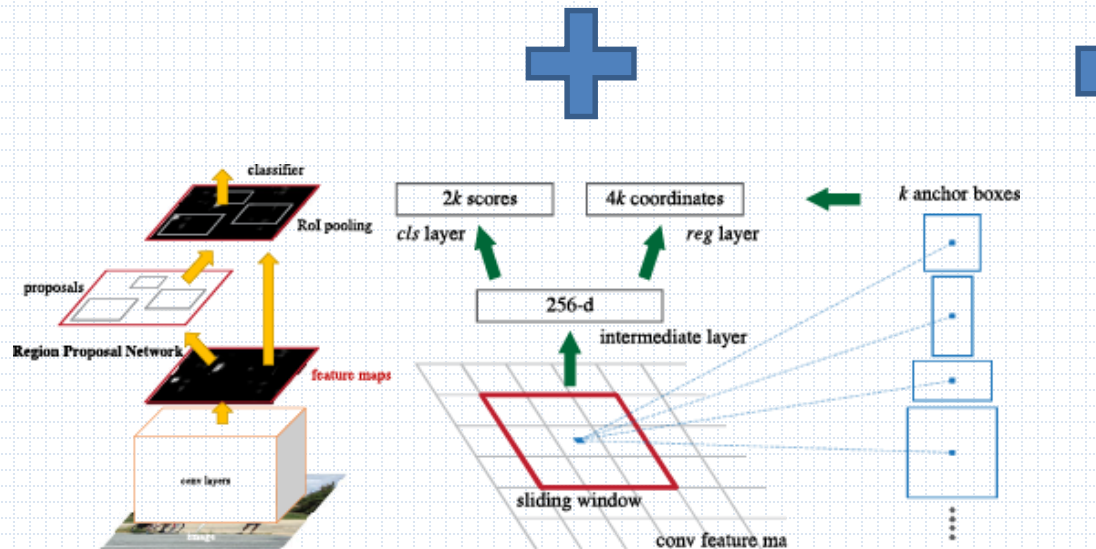
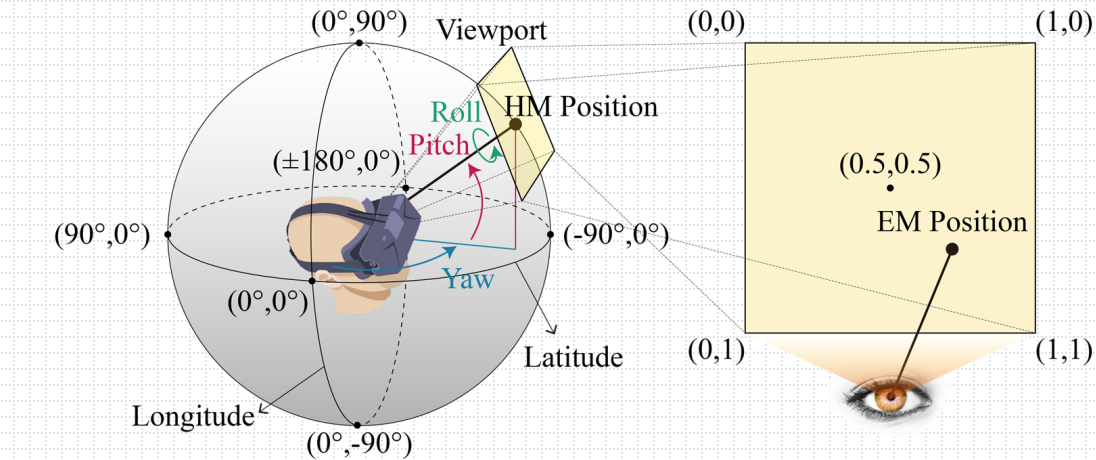
Our V-CNN Approach

- **Motivation: Human's way of viewing 360° video:**
Head Movement (HM) + Eye Movement (EM)

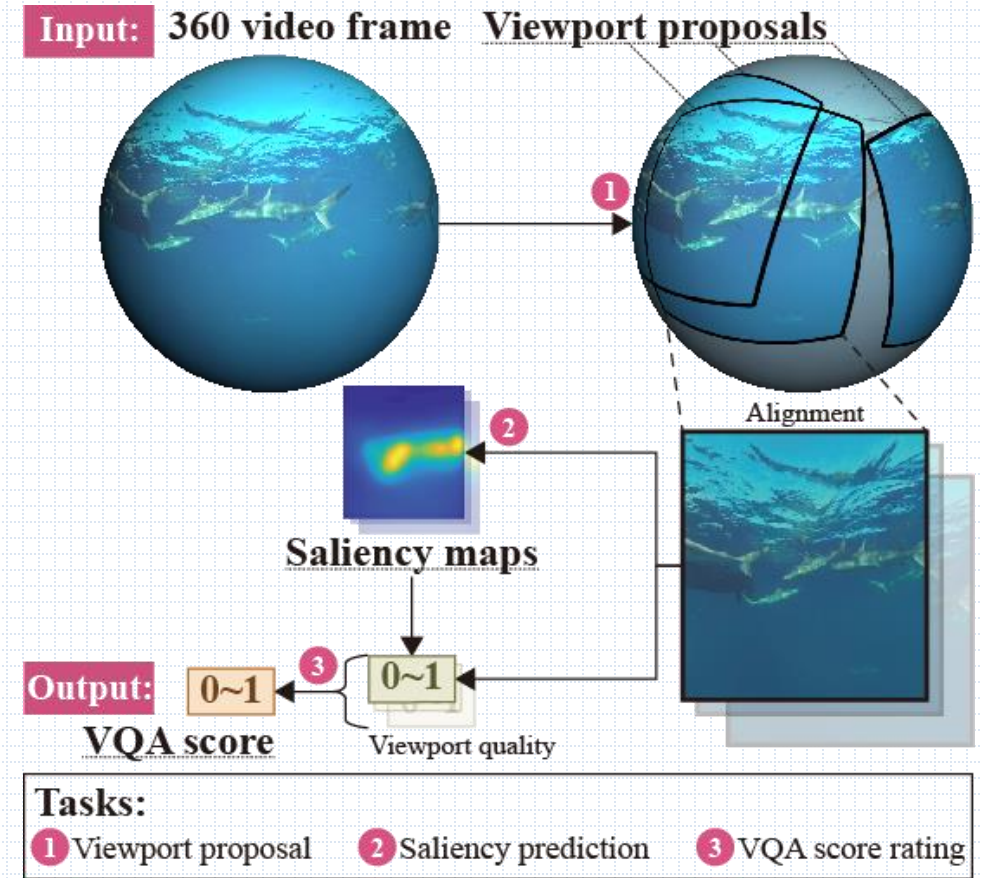


Our V-CNN Approach

- Motivation: Two-staged procedure** of viewing 360° video with viewport.



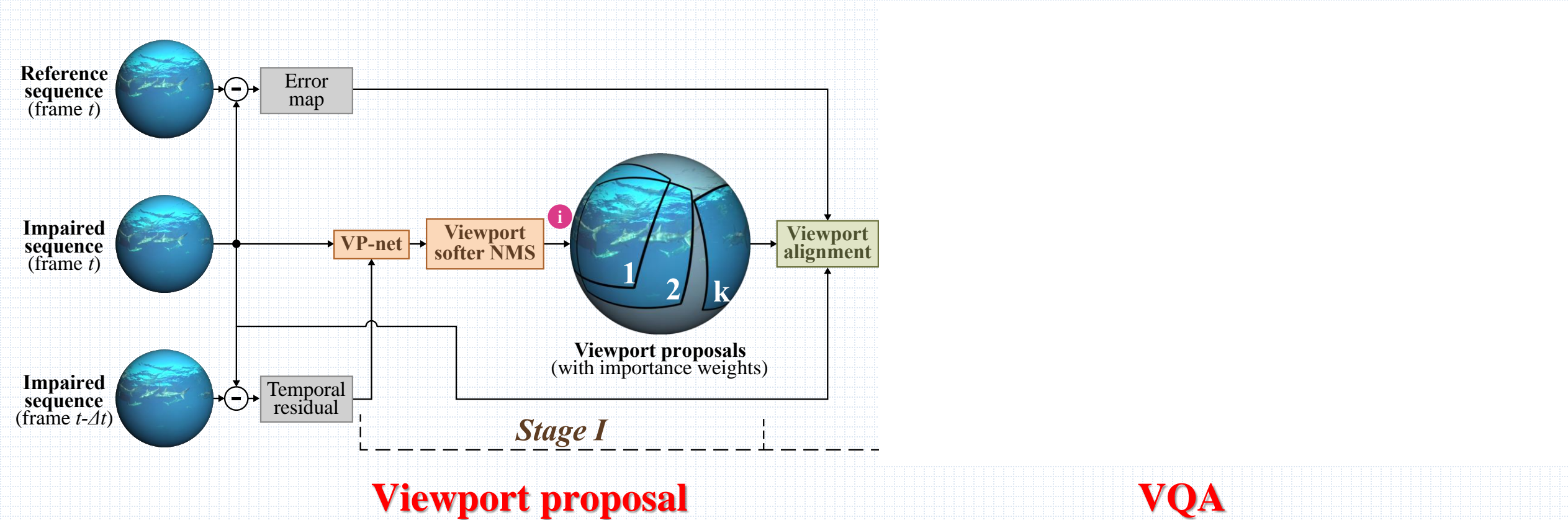
Faster R-CNN



Our V-CNN Approach

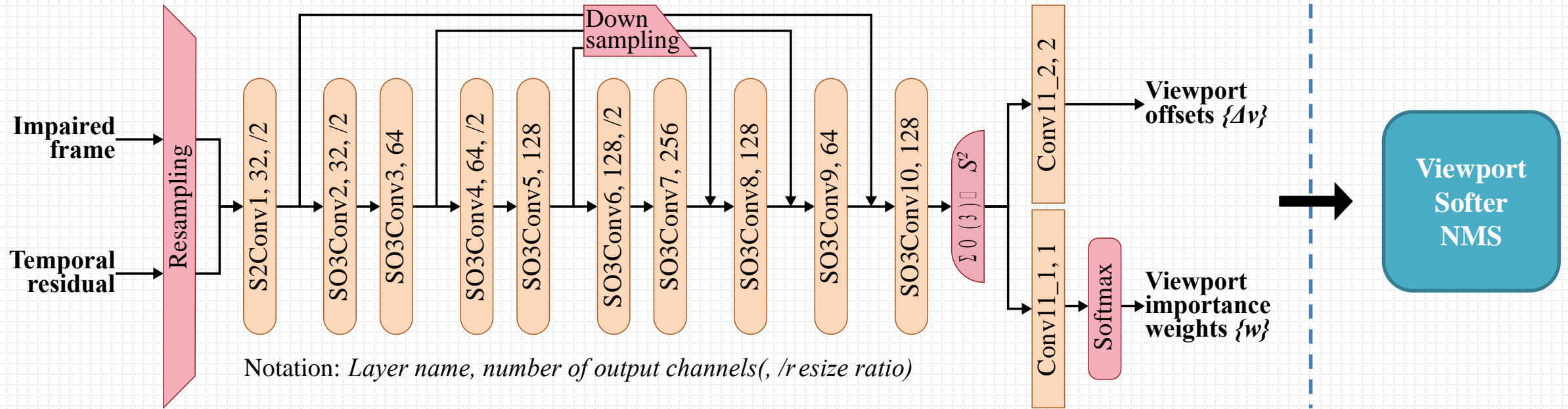
Our V-CNN Approach

■ Framework



Our V-CNN Approach

VP-net: Viewport proposal.



VP-Net

Loss: $\mathcal{L}^I = \lambda_w \mathcal{L}_w + \lambda_v \mathcal{L}_v$

Importance Weight
Anchor Location

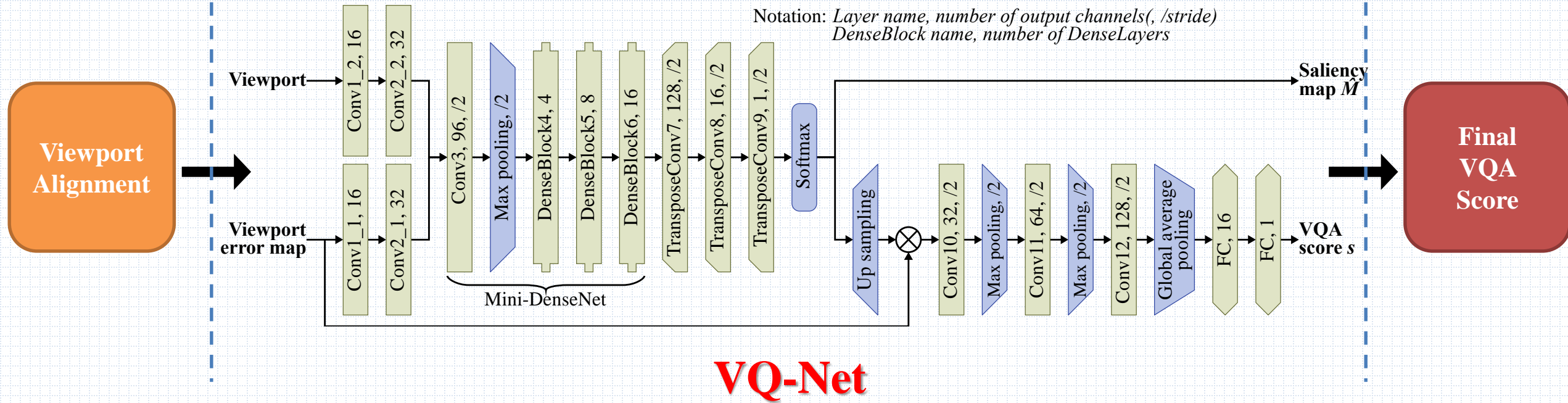
where $\mathcal{L}_w = D_{\text{KL}}(\{w_i\}_{i=1}^I \| \{\hat{w}_i\}_{i=1}^I) = \sum_{i=1}^I w_i \log \left(\frac{w_i}{\hat{w}_i} \right)$

$$\mathcal{L}_v = \sum_{i=1}^I w_i \cdot L_1^{\text{smooth}}(\Delta \mathbf{v}_i, \Delta \mathbf{v}_i^g)$$

$$\Delta \mathbf{v}_i^g = \arg \min_{\{\mathbf{v}_j^h\}_{j=1}^J} d(\mathbf{v}_i^a, \mathbf{v}_j^h) - \mathbf{v}_i^a$$

Our V-CNN Approach

■ VQ-net: Viewport quality assessment.



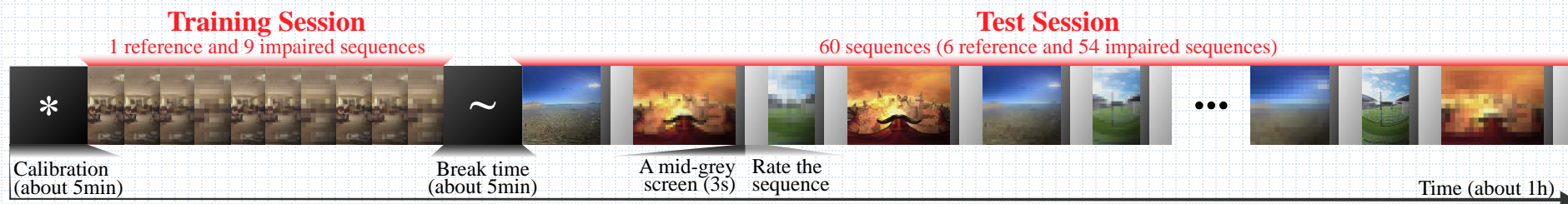
$$\text{Loss: } \mathcal{L}_k^{\text{II}} = \underbrace{\lambda_M \mathcal{L}_{M_k}}_{\text{Saliency Prediction}} + \underbrace{\lambda_s \mathcal{L}_{s_k}}_{\text{VQA Score}} \quad \text{where } \mathcal{L}_{M_k} = D_{\text{KL}}(\mathbf{M}_k \| \hat{\mathbf{M}}_k) = \sum_{(x', y')} M_k(x', y') \log \left\{ \frac{M_k(x', y')}{\hat{M}_k(x', y')} \right\}$$

$$\mathcal{L}_{s_k} = (s - s_k)^2$$

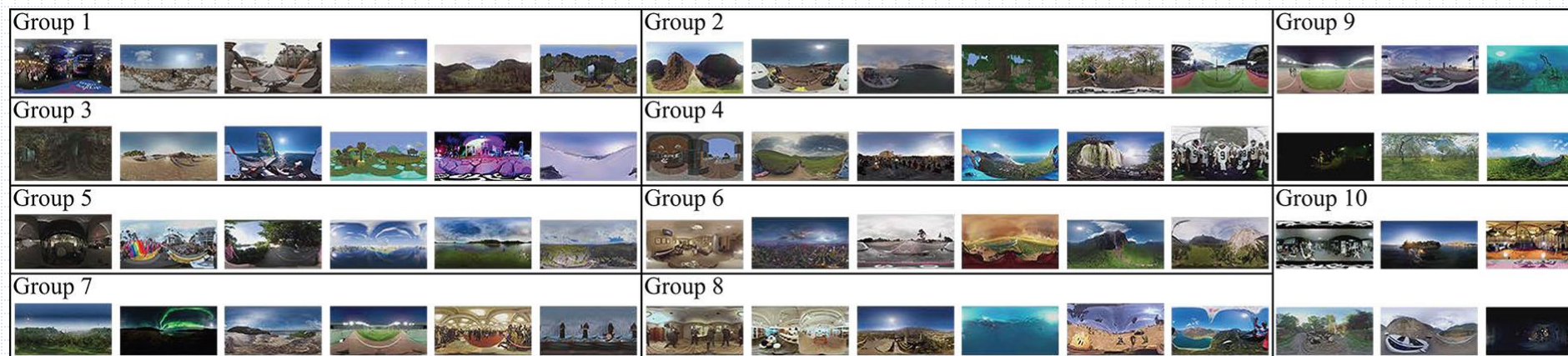
Experimental Results

■ Dataset

□ Collection Procedure



□ Dataset Details



- Sequences in total: **600**
- Reference sequences: **60**
- Projection format: ERP, RCMP, TSP
- Quality level: QP = 27, 37, 42
- Subjects in total: **221**
- Including Data: Subjective quality scores, HM and EM tracking data

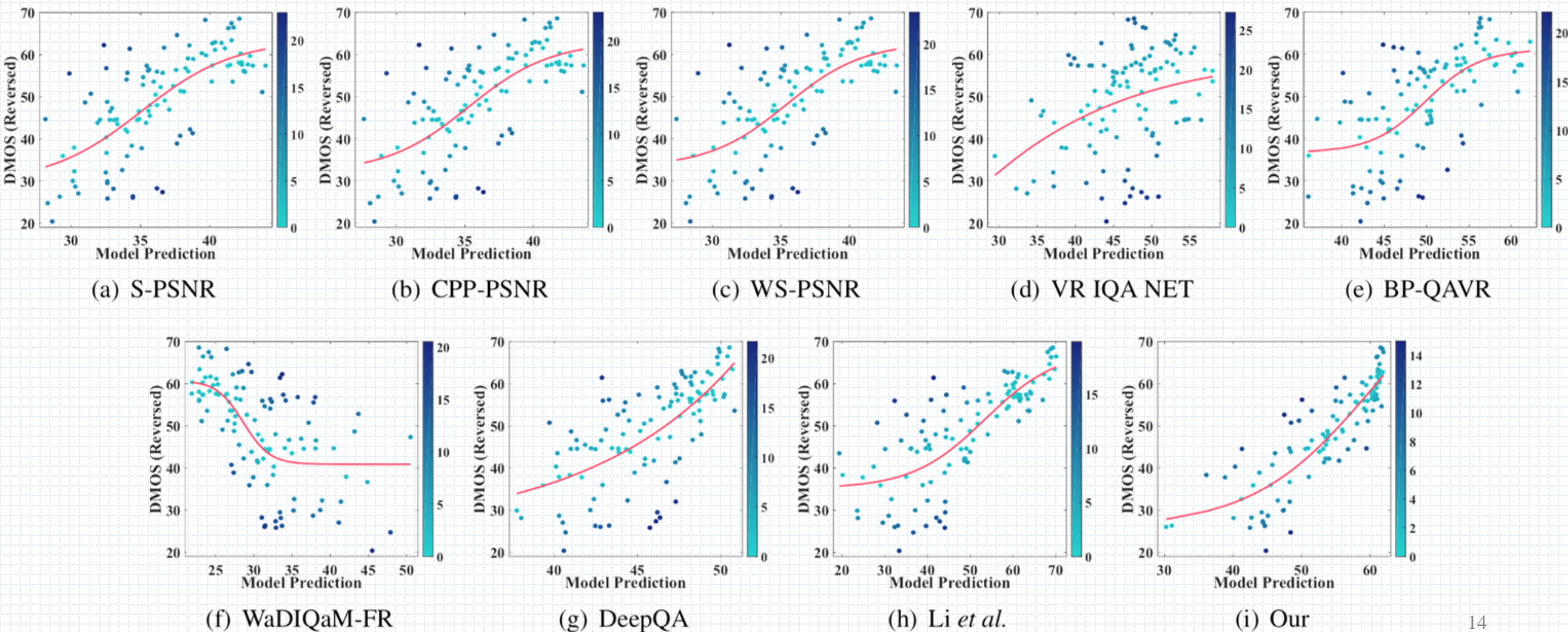
Experimental Results

■ Evaluation on VQA (Quantitative).

Approaches	Attributes				Evaluation on VQA-ODV [19]				
	For 360° video	Full reference	Deep learning	Re-trained*	PLCC	SROCC	KROCC	RMSE	MAE
S-PSNR	✓	✓			0.6929	0.6976	0.4981	8.5407	6.6810
WS-PSNR	✓	✓			0.6721	0.6839	0.4860	8.7707	6.9089
CPP-PSNR	✓	✓			0.6812	0.6896	0.4912	8.6718	6.7932
BP-QAVR	✓	✓	✓	✓	0.6588	0.6801	0.4780	8.9112	7.0823
Li et al.	✓	✓	✓	✓	0.7821	0.7953	0.5902	7.3817	5.7793
VR-IQA-NET	✓		✓		0.3713	0.3379	0.2260	10.9984	9.1016
DeepQA		✓	✓	✓	0.6936	0.7296	0.5213	8.5325	6.7720
WaDIQaM-FR		✓	✓		0.6207	0.6162	0.4206	9.2868	7.4574
V-CNN (Ours)	✓	✓	✓	✓	0.8740	0.8962	0.7137	5.7551	4.4893

Experimental Results

■ Evaluation on VQA (Scatter Plots).



Experimental Results

■ Evaluation on other auxiliary tasks.

Viewport Proposal

Approaches	NSS	CC	KL
Cheng <i>et al.</i>	1.96	0.35	2.50
DHP	1.98	0.37	2.41
GBVS360	1.16	0.22	2.31
BMS360	1.85	0.34	1.88
V-CNN (Ours)	2.65	0.63	2.38

Saliency Prediction

Approaches	NSS	CC	KL
DeepVS	1.10	0.36	1.95
BMS	0.89	0.26	1.42
GBVS	0.81	0.30	1.34
PQFT	0.73	0.19	1.60
V-CNN (Ours)	0.97	0.37	1.33



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Thank You!

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