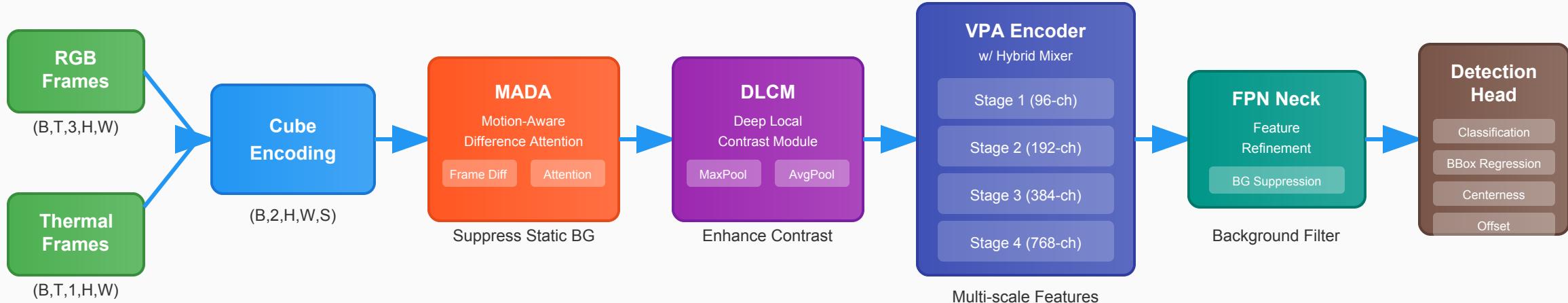


TLCFormer: Temporal-Local-Contrast Transformer Architecture



Core Innovations

MADA: Motion-Aware Difference Attention

1. Temporal Gradient:

$$\begin{aligned} D_{\text{pre}} &= |I_t - I_{t-1}| \\ D_{\text{next}} &= |I_{t+1} - I_t| \end{aligned}$$

2. Motion Saliency Map:

$$M_{\text{raw}} = D_{\text{pre}} * D_{\text{next}}$$

3. Feature Modulation:

$$I'_t = I_t * (1 + a * \text{sigmoid}(F(M_{\text{raw}})))$$

DLCM: Deep Local Contrast Module

1. Background Estimation:

$$\mu_{\text{bg}} = \text{AvgPool}_{9 \times 9}(X)$$

2. Target Intensity:

$$L_{\text{max}} = \text{MaxPool}_{3 \times 3}(X)$$

3. Contrast Response:

$$C = L_{\text{max}}^2 / (\mu_{\text{bg}} + \epsilon)$$

$$X_{\text{out}} = X + \beta * C$$

Hybrid Energy-Preserving Mixer

1. Dual-Path Pooling:

$$\begin{aligned} P_{\text{max}} &= \text{MaxPool2d}(X) \\ P_{\text{avg}} &= \text{AvgPool2d}(X) \end{aligned}$$

2. Feature Fusion:

$$P_{\text{hybrid}} = \text{Concat}(P_{\text{max}}, P_{\text{avg}})$$

3. Channel Compression:

$$X_{\text{out}} = X + \text{GELU}(\text{Conv1x1}(P_{\text{hybrid}}))$$