

SRVP:

Strong Recollection Video Prediction Model Using Attention-Based Spatiotemporal Correlation Fusion

Yuseon Kim

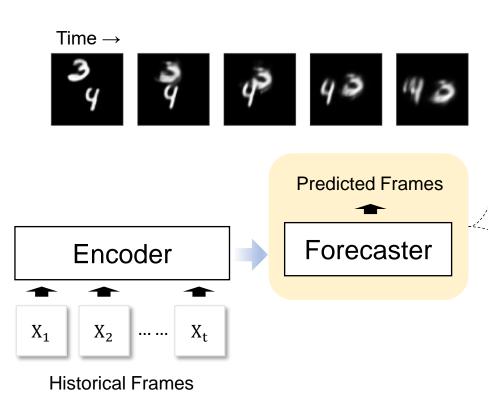
Kyongseok Park



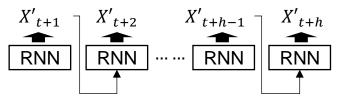


Video Prediction



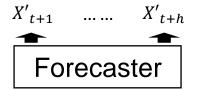


RNN-Based Approach



- ✓ Effectively Capture Temporal Changes
- ✓ ConvLSTM based Models
- ✓ Highly Sensitive to Object Motion
- ✓ Blurred Images

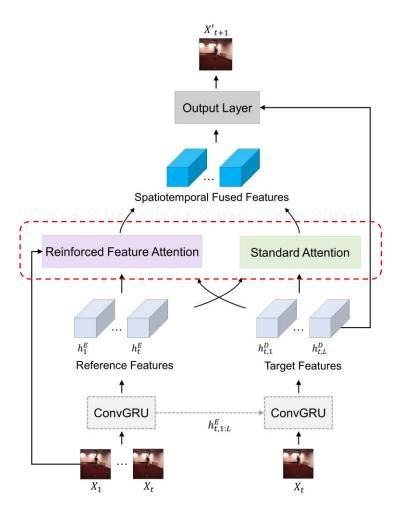
RNN-Free Approach



- ✓ Mitigates Cumulative Errors
- ✓ Vision Transformer based Models
- ✓ Absence of Explicit Mechanism for Temporal Variation

Strong Recollection Video Prediction

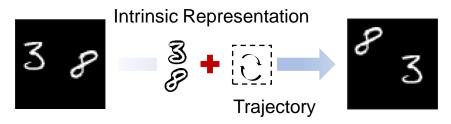




- ✓ Baseline Architecture
 - ConvGRU based Encoder-Forecaster
- ✓ Temporal Dynamics



✓ Spatiotemporal Fused Features

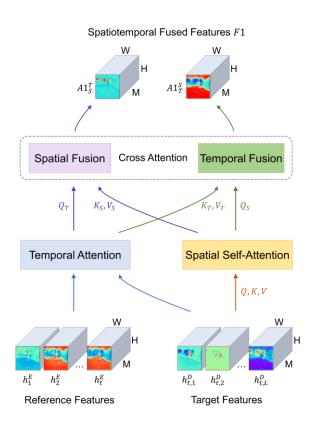


- ✓ Scaled-Dot Product Attention
 - Vaswani, Ashish, et al. "Attention is all you need." In NeurIPS (2017).

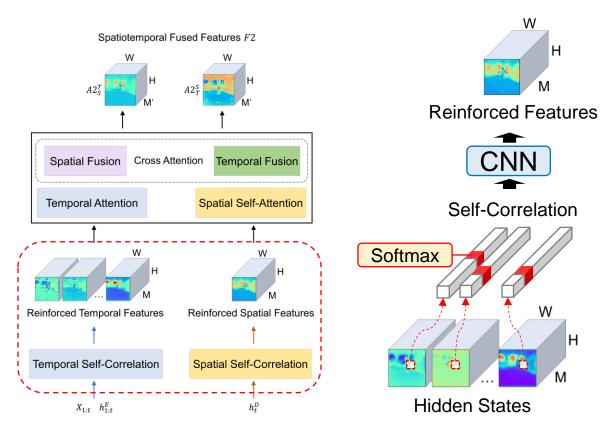
Attention-based Spatiotemporal Fusion



Standard Attention



Reinforced Feature Attention



Experiment Materials



✓ Benchmarks



Moving MNIST

- 64x64 Gray Image
- 10 Frames → 10 Frames



KTH Action

- 64x64 Gray Image
- 10 Frames → 10 Frames



Human3.6M

- 100x100 Color Image
- 4 Frames → 4 Frames

✓ Evaluation Metrics

- Mean Squared Error (MSE) ↓
- Peak Signal-to-Noise Ratio (PSNR) ↑
- Structural Similarity Index Measure (SSIM) ↑

✓ Reference Models

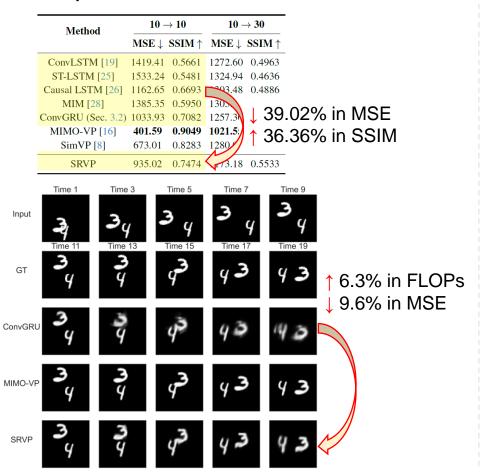
Approach	Model	Conference	
RNN-based	ConvLSTM	NIPS 2015	
	ST-LSTM	NIPS 2017	
	Causal LSTM	ICML 2018	
	MIM	CVPR 2019	
	ConvGRU	-	
RNN-free	MIMO-VP	AAAI 2023	
	SimVP	CVPR 2022	

- MIMO-VP → Transformer-based VP
- SimVP → CNN-based VP

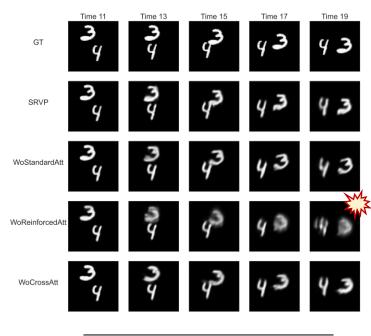
Notable Findings on SRVP



✓ Outperform than RNN-based Models



✓ Impact of Feature Reinforcement



Method	$\mathbf{MSE}\downarrow$	PSNR ↑	SSIM ↑
SRVP	935.02	18.6160	0.7474
Without-SA	968.64	18.4719	0.7365
Without-RFA	1023.84	18.2027	0.6983
Without-CrossAtt	973.16	18.4690	0.7314

Outperforming on Complex Motion

KISTI "UST

SRVP Achieved Superior Performance

Method	MSE ↓	PSNR ↑	SSIM ↑	_
ConvLSTM [19]	936.69	17.1198	0.5322	_
ST-LSTM [25]	386.21	21.9665	0.7294	
Causal LSTM [26]	408.14	21.5750	0.7158	
MIM [28]	445.27	21.3214	0.6947	
ConvGRU (Sec. 3.2)	418.09	21.5636	0.7115	L FO 000/ :: MCF
MIMO-VP [16]	502.15	21.0167	0.7276	↓ 59.09% in MSE
SimVP [8]	452.65	21.4365	0.7179	↑ 29.69% in PSNR
SRVP	383.16	22.0322	0.7360	↑ 38.29% in SSIM

Outperform than RNN-free Models

Method	Horizon			
nieurou.	1	2	3	4
ConvLSTM [19]	0.8960	0.8799	0.8617	0.8440
ST-LSTM [25]	0.9656	0.9519	0.9397	0.9285
Causal LSTM [26]	0.9639	0.9502	0.9375	0.9262
MIM [28]	0.9599	0.9446	0.9313	0.9185
ConvGRU (Sec. 3.2)	0.9643	0.9489	0.9354	0.9235
MIMO-VP [16]	0.9303	0.9154	0.9034	0.8924
SimVP [8]	0.9415	0.9286	0.9194	0.9122
SRVP	0.9659	0.9512	0.9386	0.9272

1.64 to 3.9% in SSIM



Stronger Robustness and Consistency

















VS

















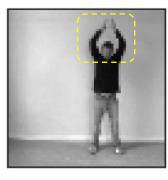


Producing Errors even in Static Regions

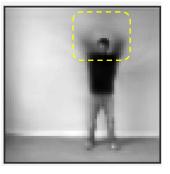
Conclusion



Ground Truth

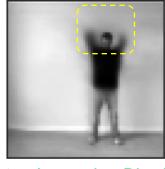


RNN-based

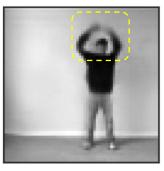


- Error Accumulation
- Blurring

SRVP



RNN-free



- Improving Blurring Problems
- Consistent Performance
- Future Work
 - → Enhancing Recollection Ability
- Temporal Inconsistency (Inaccurate Position)
- Sensitive to image size changes





Thank you for your attention!

https://github.com/yuseonk/SRVP