Abstract

Introduction

- The goal of SR(Super Resolution) is to enhance a LR(Low Resolution) image to a HR(High Resolution) image.
- The domain of SR research can be divided into 3 main areas: SISR(Single Image SR),
 MISR(Multi Image SR) and VSR(Video SR).
- Consider an LR video which consists of a sequence of frames: $LR_{t-n}, \ldots, LR_t, \ldots, LR_{t+n}$, where we super-resolve a target frame LR_t .
 - \circ SISR: Only use LR_t , so obviously it fails to utilize temporal information.
 - MISR: Align all the neighboring frames and fuses them to utilize the mising details observed only in one or some of the frames.
 - VSR: Almost similar to MISR, but the alignment progress is done with concern for temporal smoothness, which is not concerned in MISR.
- IQA(Image Quality Assessment): PSNR(Peak Signal to Noise Ratio) and SSIM(Structural Similarity).
 - PSNR:

$$egin{aligned} MSE &= rac{1}{H imes W} \sum_{i=1}^{H} \sum_{j=1}^{W} \left(X(i,j) - Y(i,j)
ight)^2 \ PSNR &= 10 \lg(rac{(2^n-1)^2}{MSE}) \end{aligned}$$

- Larger is better, the unit is decible(db).
- **n** is number of bits per pixel.
- o SSIM:

$$\begin{split} \mu_X &= \frac{1}{H \times W} \sum_{i=1}^H \sum_{j=1}^W X(i,j), \, \sigma_X^2 = \frac{1}{H \times W} \sum_{i=1}^H \sum_{j=1}^W (X(i,j) - \mu_X)^2 \\ \sigma_{XY} &= \frac{1}{H \times W} \sum_{i=1}^H \sum_{j=1}^W ((X(i,j) - \mu_X)(Y(i,j) - \mu_Y)) \\ l(X,Y) &= \frac{2\mu_X \mu_Y + C_1}{\mu_X^2 + \mu_Y^2 + C_1}, \, c(X,Y) = \frac{2\sigma_X \sigma_Y + C_2}{\sigma_X^2 + \sigma_Y^2 + C_2}, \, s(X,Y) = \frac{\sigma_{XY} + C_3}{\sigma_X \sigma_Y + C_3} \\ SSIM(X,Y) &= l(X,Y) \cdot c(X,Y) \cdot s(X,Y) \end{split}$$

- Larger is better.
- Light, Contrast and Structure.
- To avoid the case where the denominator is 0, normally we use $C_1 = (K_1 \cdot L)^2, C_2 = (K_2 \cdot L)^2, C_3 = C_2/2$, and $K_1 = 0.01, K_2 = 0.03, L = 255$.
- In most cases, **CNNs** are well qualified for the job, but once the upscaling factors are large enough, **GAN** approach is nearly indispensable to render the "naturality".
- Deep VSR can be divided into 5 types based on the approach to preserving temporal information.
 - Temporal Concatenation: DUF
 - Temporal Aggregation: Robust Video Super-Resolution with Learned Temporal
 Dynamics

- o RNNs: Detail-revealing deep video super-resolution, Bidirectional recurrent convolutional networks for multi-frame super-resolution
 - many-to-one
 - many-to-many
- Optical Flow: Framerecurrent video super-resolution
 - warp
 - Video super-resolution with convolutional neural networks
 - Total variation regularization of local-global optical flow
 - VESPCN: replace the optical flow model with a trainable motion compensation network.
- Pre-training then Fine-tuning
 - VESPCN
 - Detail-revealing deep video super-resolution
 - Robust Video Super-Rsolution with Learned Temporal Dynamics
- 2D-conv & 3D-conv: https://www.zhihu.com/question/266352189

Related Work

timeline

- SRCNN[2014], VDSR[2016]
- DUF[2018], TDAN[2018], EDVR[2019]
- ESPCN(calc in lr space)[2016], VESPCN(video espcn)[2017]
- DBPN[2018], RBPN[2019]
- SRGAN[2017]
- iSeeBetter(RBPN+SRGAN)[2020]

results

SRCNN

Video Super-Resolution	Ultra Video Group HD - 4x upscaling	SRCNN	Average PSNR	37.52	# 2	Ð	Compare
Image Super- Resolution	Urban100 - 4x upscaling	SRCNN	PSNR	24.52	# 35	Ð	Compare
			SSIM	0.7221	#31	Ð	Compare
Video Super-Resolution	Vid4 - 4x upscaling	SRCNN	PSNR	24.68	# 14	Ð	Compare
			SSIM	0.7158	# 13	Ð	Compare
			MOVIE	6.90	# 4	Ð	Compare
Video Super-Resolution	Xiph HD - 4x upscaling	SRCNN	Average PSNR	31.47	# 2	Ð	Compare

EDVR



• ESPCN

Video Super-Resolution	Ultra Video Group HD - 4x upscaling	ESPCN	Average PSNR	37.91	#1	Ð	Compare
Video Super-Resolution	Ultra Video Group HD - 4x upscaling	bicubic	Average PSNR	36.20	#3	Ð	Compare
Video Super-Resolution	Vid4 - 4x upscaling	ESPCN	PSNR	25.06	# 13	Ð	Compare
			SSIM	0.7394	# 11	Ð	Compare
			MOVIE	6.54	#3	Ð	Compare
Video Super-Resolution	Xiph HD - 4x upscaling	bicubic	Average PSNR	30.30	#3	Ð	Compare
Video Super-Resolution	Xiph HD - 4x upscaling	ESPCN	Average PSNR	31.67	# 1	Ð	Compare

VESPCN

Ranked #10 on Video Super-Resolution on Vid4 - 4x upscaling							
TASK	DATASET	MODEL	METRIC NAME	METRIC VALUE	GLOBAL RANK	RESULT	BENCHMARK
Video Super-Resolution	Vid4 - 4x upscaling	VESPCN	PSNR	25.35	# 12	Ð	Compare
			SSIM	0.7557	# 10	Ð	Compare
			MOVIE	5.82	# 2	Ð	Compare
Video Super-Resolution	Vid4 - 4x upscaling	bicubic	PSNR	23.82	# 16	Ð	Compare
			SSIM	0.6548	# 15	Ð	Compare
			MOVIE	9.31	# 5	Ð	Compare

DBPN

TASK	DATASET	MODEL	METRIC NAME	METRIC VALUE	GLOBAL RANK	RESULT	BENCHMARK
Image Super-Resolution	BSD100 - 4x upscaling	D-DBPN	PSNR	27.72	# 11	Ð	Compare
			SSIM	0.740	# 18	Ð	Compare
Image Super-Resolution	Manga109 - 4x upscaling	D-DBPN	SSIM	0.914	# 14	Ð	Compare
Image Super-Resolution	Set14 - 4x upscaling	D-DBPN	PSNR	28.82	# 11	Ð	Compare
			SSIM	0.786	# 19	Ð	Compare
Image Super-Resolution	Set5 - 4x upscaling	D-DBPN	PSNR	32.47	# 12	Ð	Compare
			SSIM	0.898	# 16	Ð	Compare
Image Super-Resolution	Urban100 - 4x upscaling	D-DBPN	SSIM	0.795	# 15	Ð	Compare
Video Super-Resolution	Vid4 - 4x upscaling	DBPN	PSNR	25.37	# 11	Ð	Compare
			SSIM	0.737	# 12	Ð	Compare

• RBPN

Results from the Paper

I Edit ∨

Ranked #6 on Video Super-Resolution on Vid4 - 4x upscaling → Get a GitHub badg							
TASK	DATASET	MODEL	METRIC NAME	METRIC VALUE	GLOBAL RANK	RESULT	BENCHMARK
Video Super-Resolution	Vid4 - 4x upscaling	RBPN/6-PF	PSNR	27.12	# 6	Ð	Compare
			SSIM	0.818	# 6	Ð	Compare

• iSeeBetter

Results from the Paper

I Edit ∨

Ranked #1 on Video Su	per-Resolution on Vimeo90k					→l Get a GitHub badge
TASK	DATASET	MODEL	METRIC NAME	METRIC VALUE	GLOBAL RANK	BENCHMARK
Video Super-Resolution	Vid4 - 4x upscaling	iSeeBetter	PSNR	27.43	#3	Compare
			SSIM	0.835	# 1	Compare
Video Super-Resolution	Vimeo90k	iSeeBetter	PSNR	40.17	# 1	Compare