Review Series of Recent Deep Learning Papers: Parameter Prediction Paper: Dynamic Filter Networks

Bert De Brabandere, Xu Jia, Tinne Tuytelaars, Luc Van Gool NIPS 2016

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- Given related views humans can predict the next view
- ② Given a video frame humans can predict the next frame
- Oeep Networks trained for the same tasks use the same trained filtering operation for all samples.
- But, for example in a video frame prediction, different videos may have different motion patterns.
- **O** Dynamic filters for flexibility : filters conditioned on some input.

Dynamic Filter Network



- I Filter Generating Network generates filter conditioned on Input A
- Interpretation of the provide the second state of the second st
- Input A and B may be the same or different

Filter Generating Network



1 Input A $\mathbb{R}^{h \times w \times c_A}$

2 Input B $\mathbb{R}^{h \times w \times c_B}$

③ Generated *n* Filters F_{θ} parameterized by $\theta \in \mathbb{R}^{s \times s \times n \times c_B}$ if c_B is the number of channels in Input B

Dynamic Local Filtering Layer



Filters are not same for all spatial locations.

Generated Filters vary for positions on the image

Generated Filters $F_{\theta}^{(i,j)}$ where $\theta \in \mathbb{R}^{s \times s \times n \times c_B \times h \times w}$ if c_B is the number of channels in Input B

Helps to model local position specific transformation and

Dynamic Local Filtering Layer



If I_A and I_B are both images, use a CNN for the fiter generating network. Generated Filters conditioned on corresponding positions in I_A . $G(i,j) = F_{\theta}^{(i,j)}(I_B(i,j))$

Results: Video Prediction

1 Task: predict next frame given a sequence of video frames



Model for Video Prediction



Synthetic moving MNIST

	Moving MNIST	
Model	# params	bce
FC-LSTM [19]	142,667,776	341.2
Conv-LSTM [18]	7,585,296	367.1
Spatio-temporal [15]	1,035,067	179.8
Baseline (ours)	637,443	432.5
DFN (ours)	637,361	285.2

Results: Moving MNIST dataset

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