Projective RNN

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1 Projective RNN

Receiving a sequence of inputs RNN models maintain hidden state as recurrent function $h_t = \text{RNN}(x_t, h_{t-1})$.

The idea of projective RNN is to couple arbitrary RNN (e.g. LSTM, GRU) module with two dynamically updated matrices X_t and H_t , consisting of important events in input and state spaces respectively.

Having an input x we project it to linear span of dictionary X. The optimal projection $\pi_X(x)$ is found by minimizing $||x - X\alpha||^2$ and is equal to:

$$\pi_X(x) := X(X^T X)^{-1} X^T x.$$

We now define projective RNN:

$$X_{t} = \begin{cases} X_{t-1} & \text{if } ||x_{t} - \pi_{X_{t-1}}(x_{t})|| < \epsilon_{x}, \\ [X_{t-1}, x_{t}] & \text{otherwise}, \end{cases}$$

$$h_{t} = \text{RNN}(\pi_{X_{t}}(x_{t}), \pi_{H_{t-1}}(h_{t-1})). \tag{1}$$

$$H_{t} = \begin{cases} H_{t-1} & \text{if } ||h_{t} - \pi_{H_{t-1}}(h_{t})|| < \epsilon_{h}, \\ [H_{t-1}, h_{t}] & \text{otherwise}. \end{cases}$$

where X_t and H_t are dictionaries of inputs and hidden states, and ϵ_x and ϵ_h are two thresholds controlling their updates.