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# Short-term Solar Irradiance Forecasting from Sky Images



Hoang Chuong Nguyen, Miaomiao Liu  
Australian National University

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Tackling Climate Change with Machine Learning

# Background

- Global warming have become critical issue.



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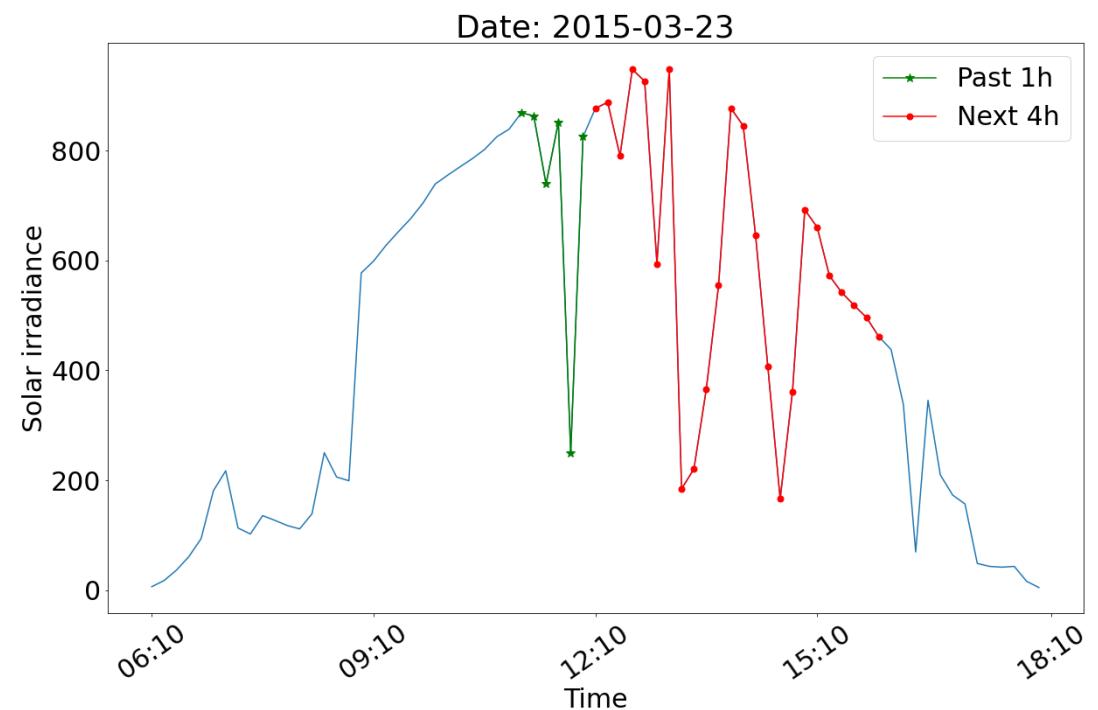
The background of the slide features a vast array of solar panels stretching across the frame. The panels are dark blue with white grid patterns. Above them, the sky is filled with large, billowing clouds. Sunbeams are visible through the clouds on the left side, creating a bright, golden glow.

Producing solar power is not fully manageable due to the unstable environmental factors.

→ Solar power forecasting helps the operation to be more stable.

# Background

- Forecasting: forecast future solar irradiance using historical sky images and auxiliary data.
  - Deterministic intra-hourly predictions.
  - Future solar irradiance diverse over a relatively long-term ( $> 1$  hour).



Example of forecasting: Use data in the past 1 hour to predict solar irradiance for the next 4 hours

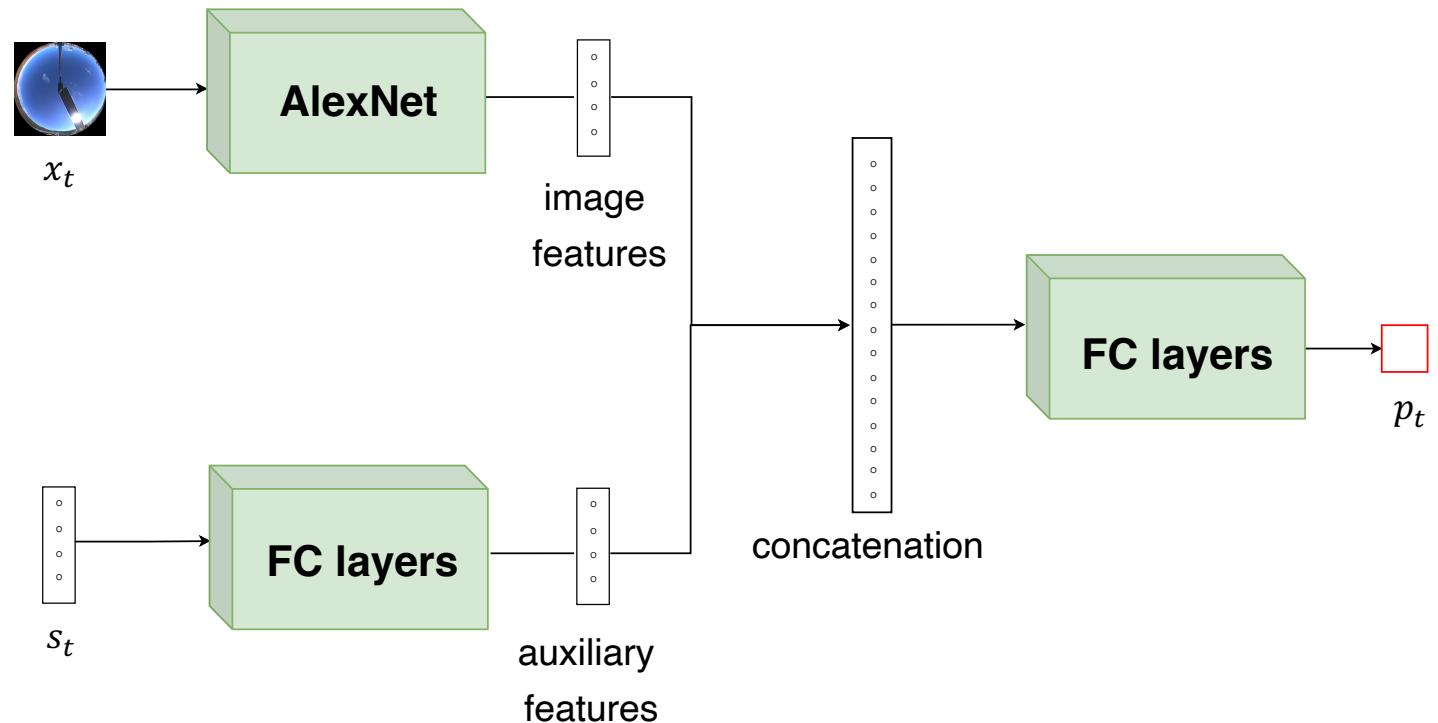
# Methodology

- We propose two models for solar irradiance forecasting:
  1. Deterministic forecasting model.
  2. Stochastic forecasting model.
- Each model contains three components:
  1. A nowcasting model
  2. An auxiliary LSTM
  3. A model for predicting future sky images.

# The nowcasting model

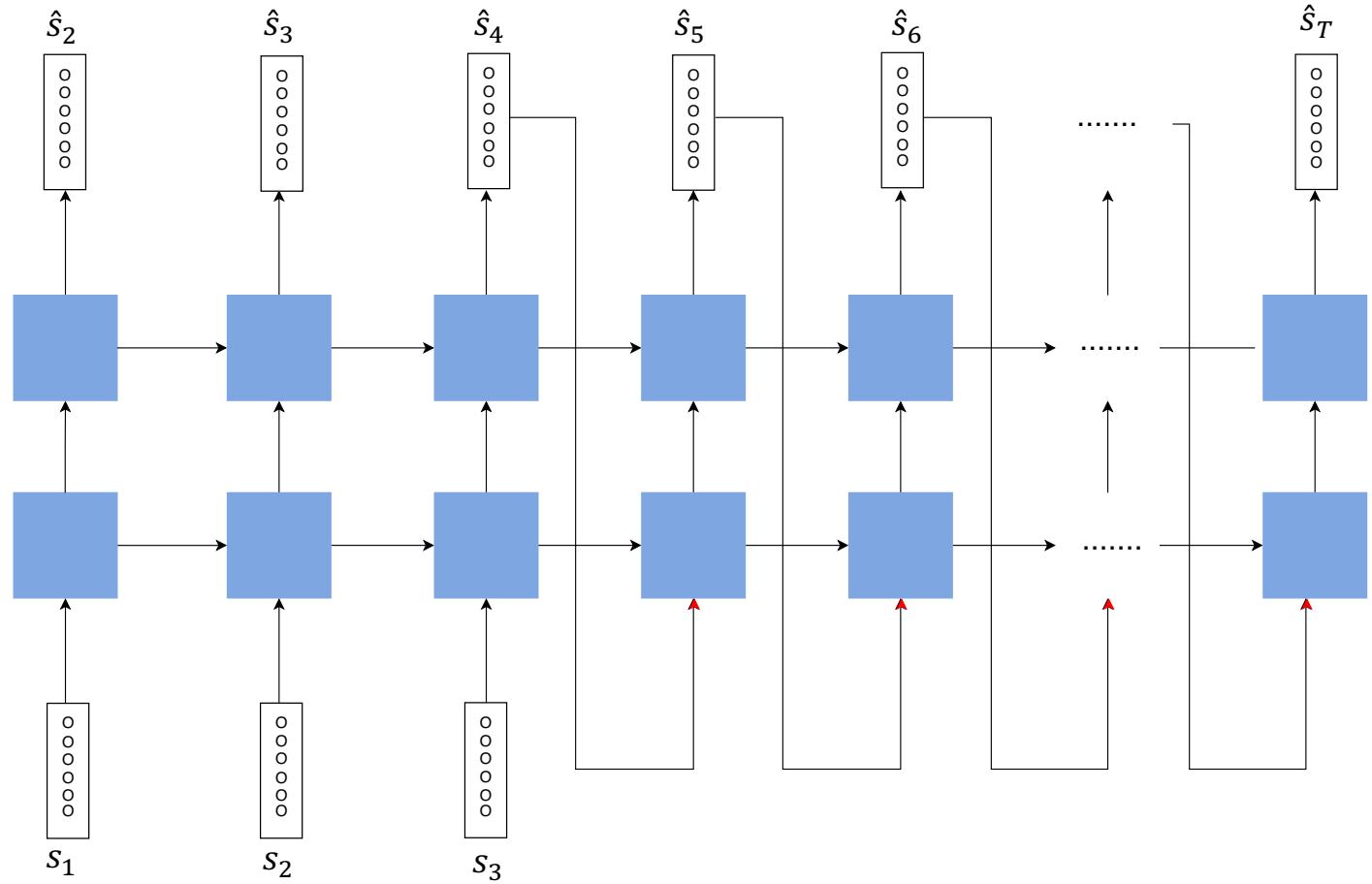
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- Estimate solar irradiance at a specific timestep using:
  - Sky image
  - Auxiliary data.



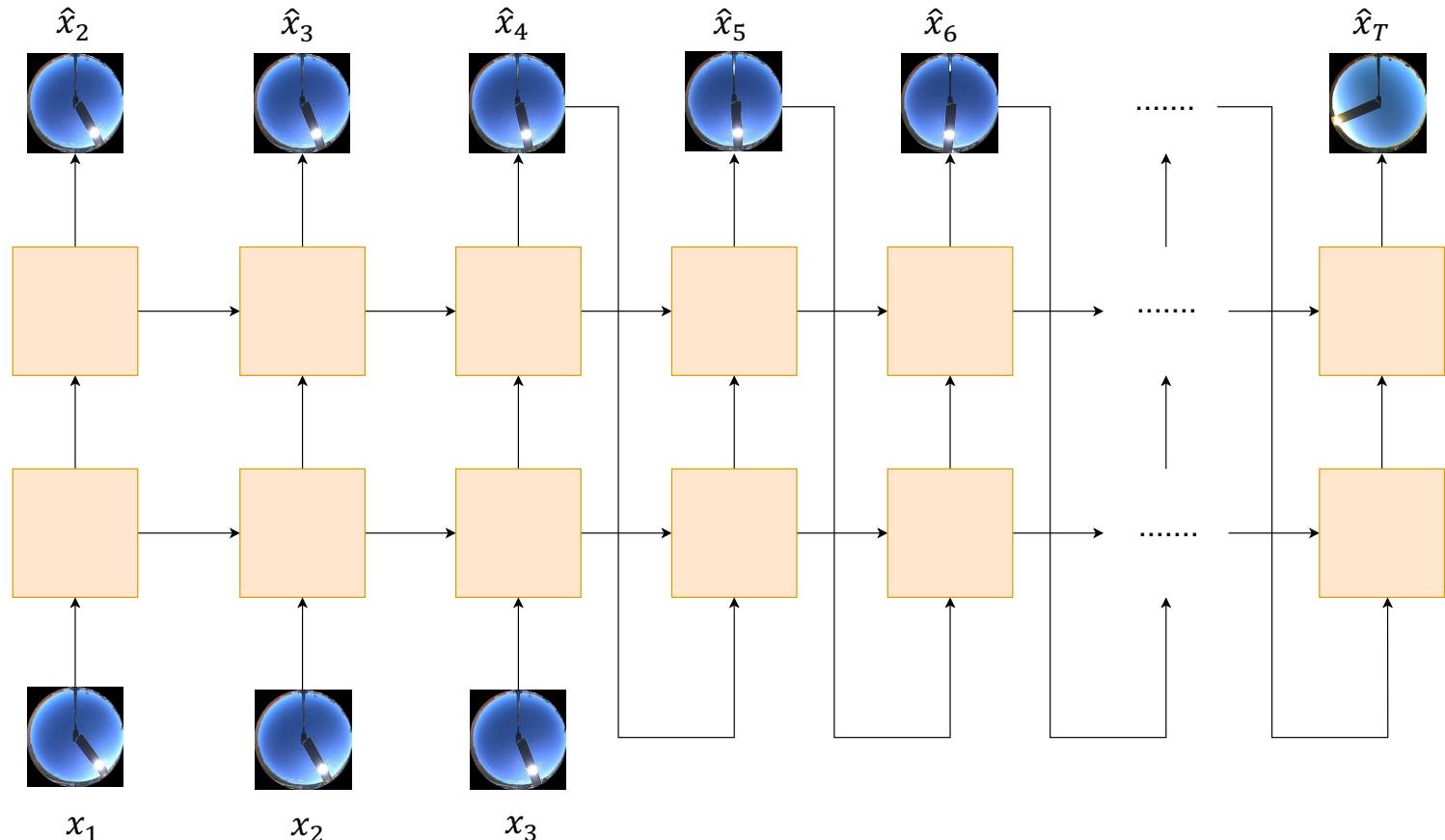
# The auxiliary LSTM model

- Has a LSTM structure
- Aim to predict future auxiliary data  
**autoregressively** given the historical ones.



# The PredRNN model [1]

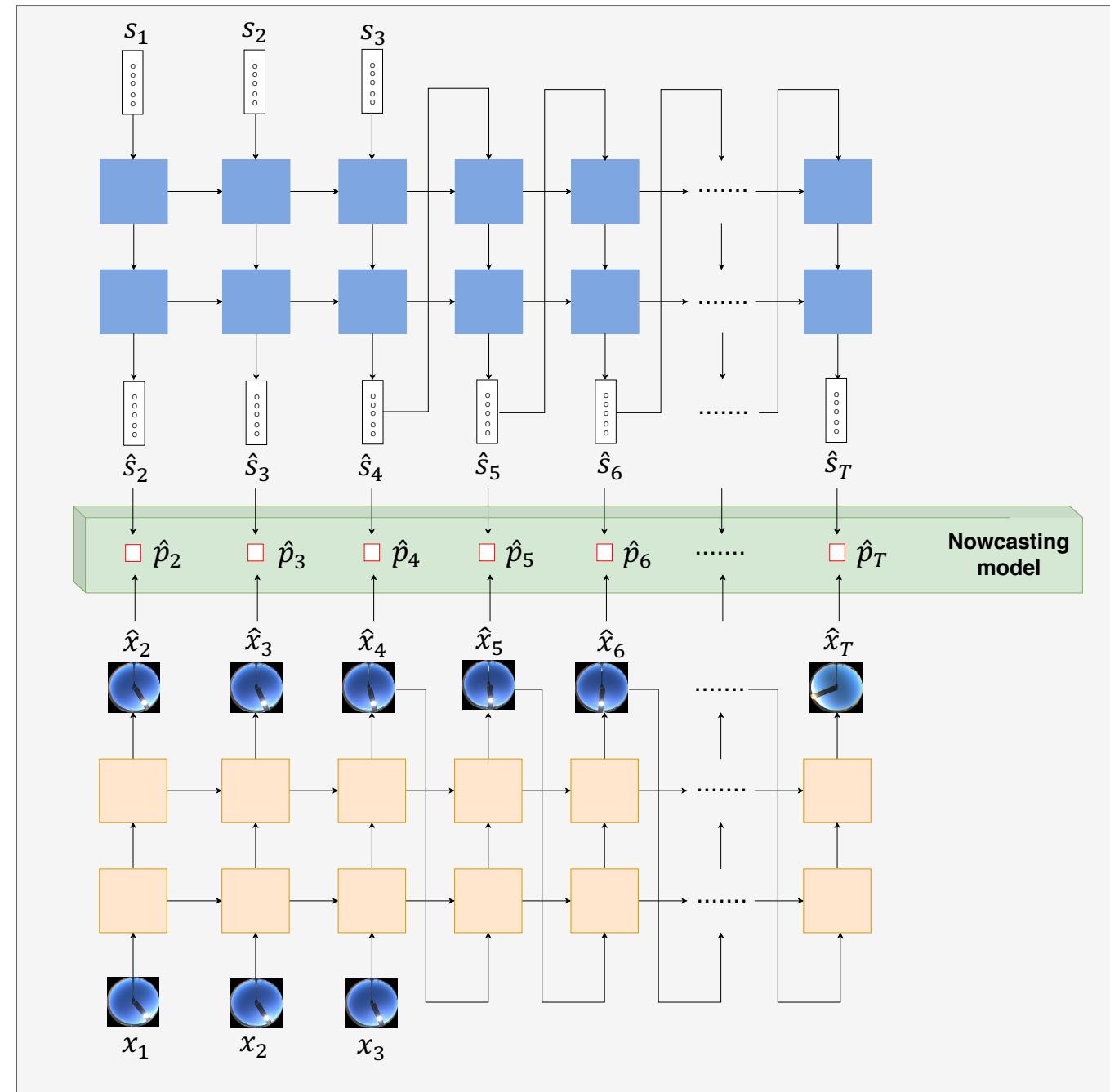
- Spatial-temporal model
- Aim to predict future sky images **autoregressively** given the historical ones.



[1] Wang, Y., Long, M., Wang, J., Gao, Z. and Yu, P.S., 2017, December. Predrnn: Recurrent neural networks for predictive learning using spatiotemporal lstms. In *Proceedings of the 31st International Conference on Neural Information Processing Systems* (pp. 879-888).

# The deterministic forecasting model

- Contains three components:
  - The nowcasting model
  - The auxiliary LSTM
  - The PredRNN model
- Prediction process:
  - Step 1: PredRNN predicts future images
  - Step 2: Auxiliary LSTM predicts future auxiliary data
  - Step 3: The nowcasting model predicts future solar irradiance.



# Loss function

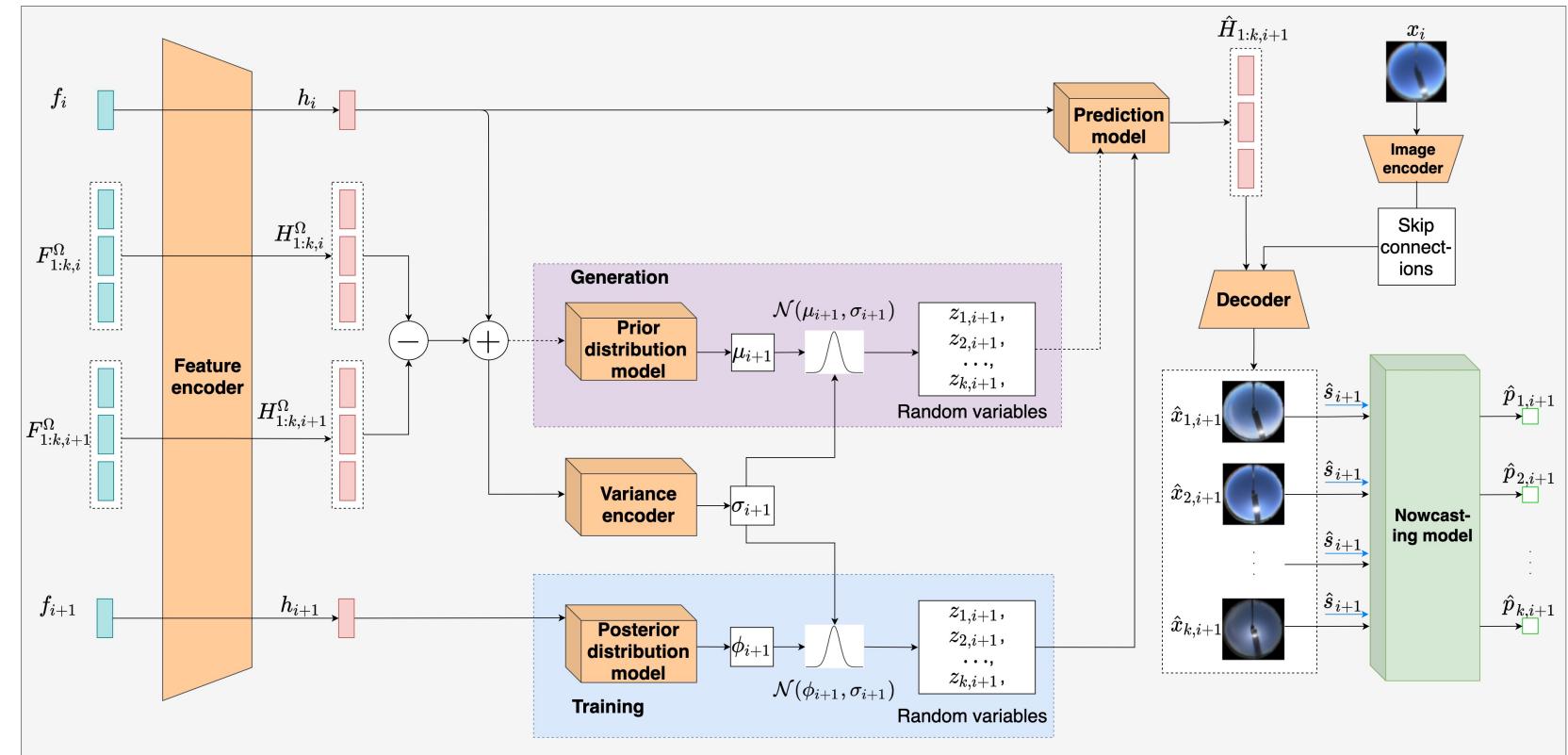
$$\mathcal{L}_{det} = \frac{1}{M + T - 1} \left( \sum_{i=1}^{M+T-1} \underbrace{\left| \left| p_{i+1} - \hat{p}_{i+1} \right| \right|_1}_{\text{Solar irradiance loss}} + \alpha \underbrace{\left| \left| x_{i+1} - \hat{x}_{i+1} \right| \right|_1}_{\text{Image loss}} \right)$$

Where:

- $p_{i+1}$  and  $\hat{p}_{i+1}$  are ground-truth and predicted solar irradiance at timestep  $i + 1$
- $x_{i+1}$  and  $\hat{x}_{i+1}$  are ground-truth and predicted image at timestep  $i + 1$

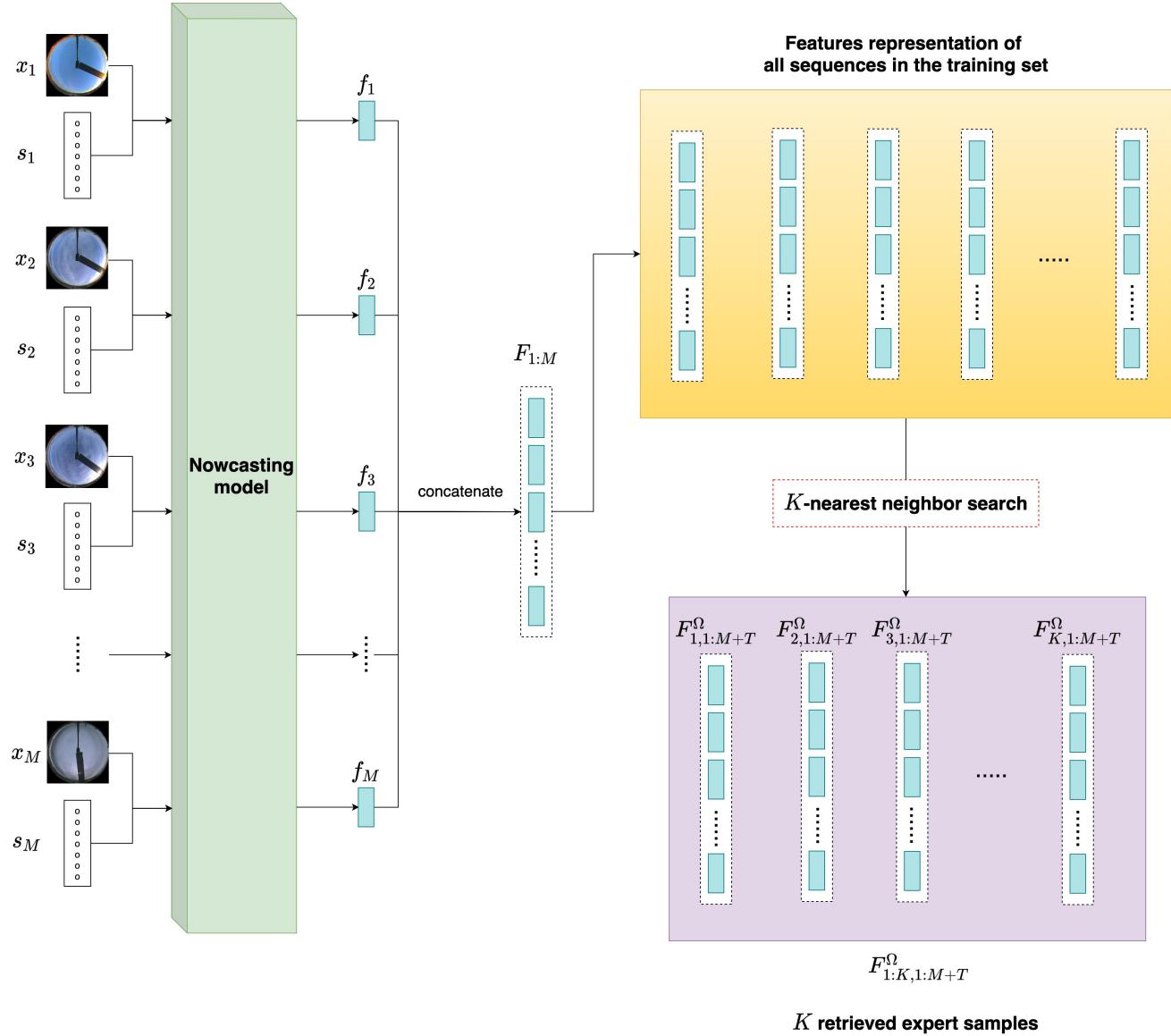
# The stochastic forecasting model

- Contains three components:
  - The nowcasting model
  - The auxiliary LSTM
  - The VPEG model
- VPEG model [2] aims predicts a distribution of future sky images.
- Contain three phases:
  - Expert samples retrieval
  - Training phase
  - Generation phase



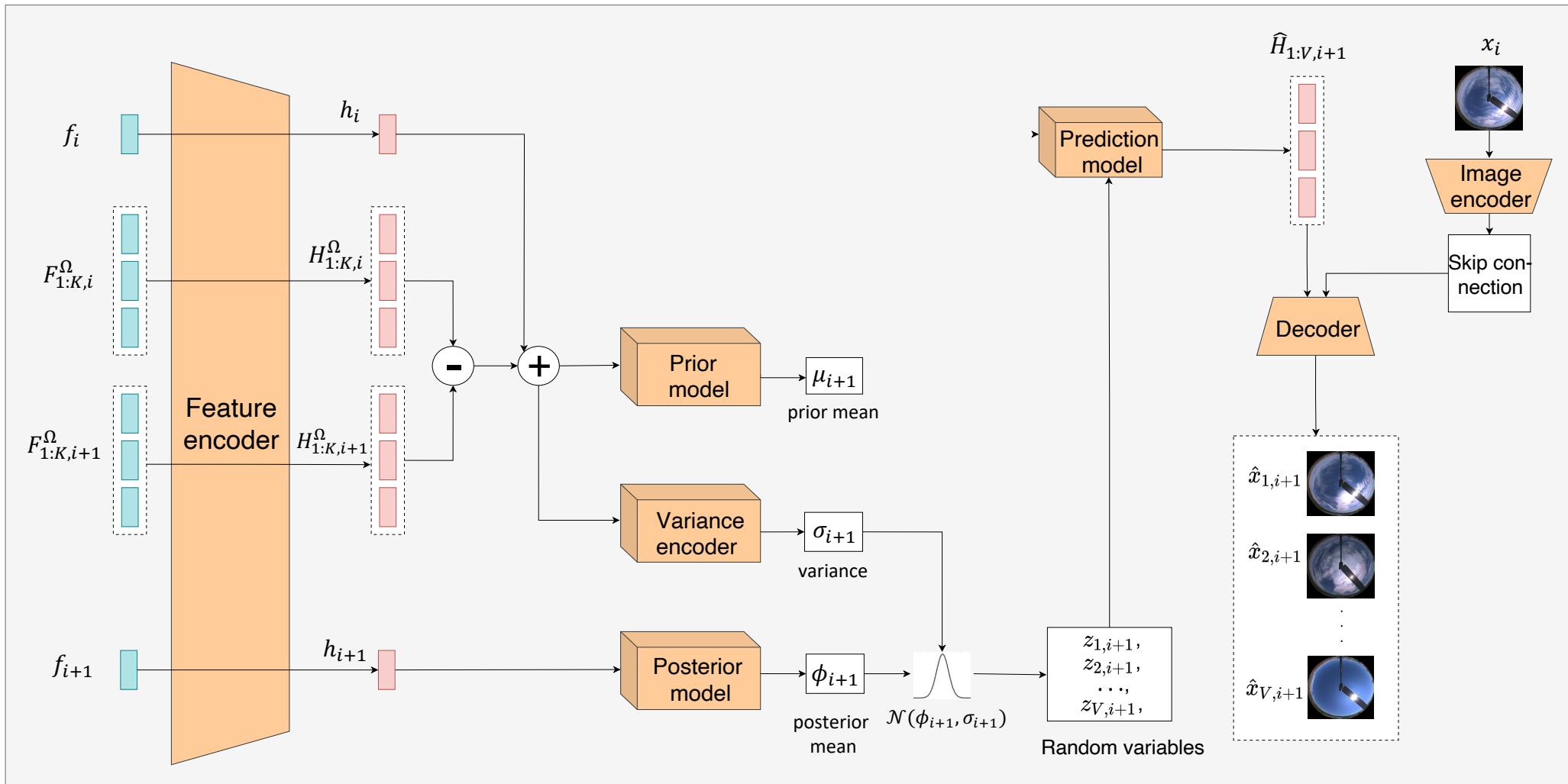
# Retrieval phase

- Use the output of the last hidden layer of the nowcasting model.
- Each sequence is represented as a sequence of features  $f$ .
- Perform  $K$ -nearest neighbor search to retrieve  $K$  expert examples.



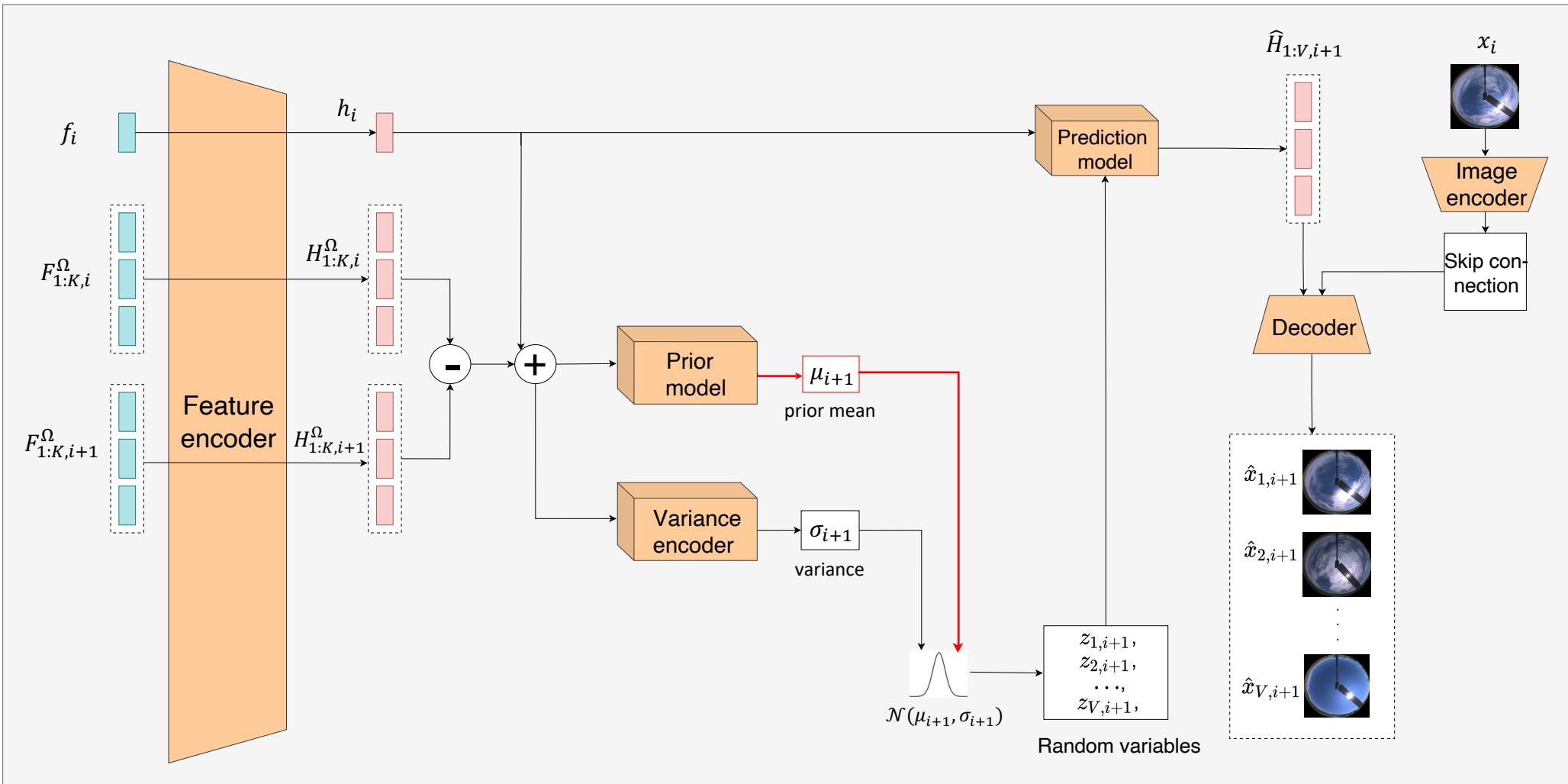
# Training phase

The prior mean is also predicted



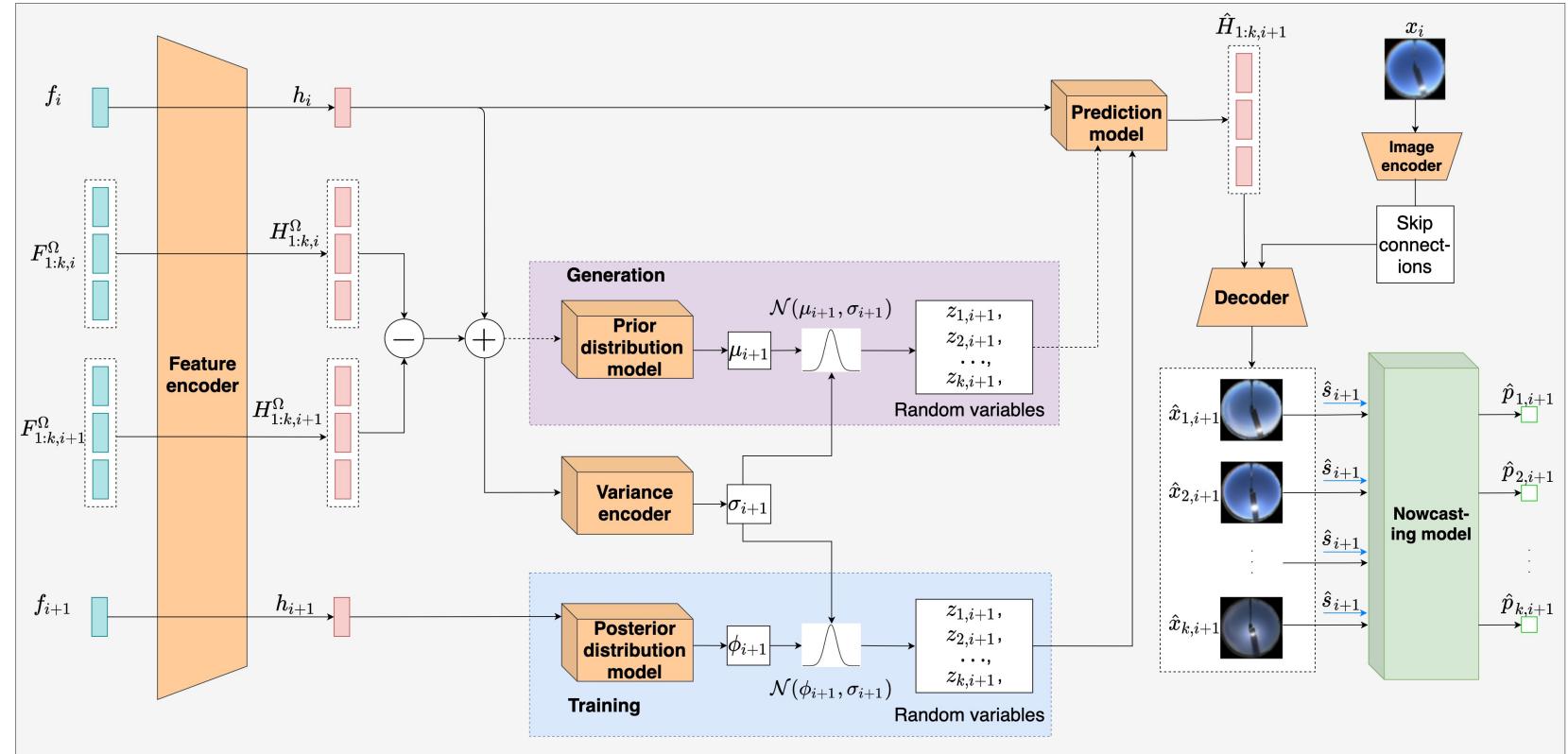
# Generation phase

Use the prior distribution (instead of the posterior distribution)



# The stochastic forecasting model

- Prediction process:
  - Step 1: VPEG generates multiple future images.
  - Step 2: Auxiliary LSTM predicts future auxiliary data.
  - Step 3: The nowcasting model predicts multiple future solar irradiance.



# Loss function



$$\textbf{Loss} = \lambda_1 \textbf{image loss} + \lambda_2 \textbf{expectation loss} + \lambda_3 \textbf{variance loss} + \lambda_4 \textbf{solar loss}$$

$$\text{image loss} = \| \text{best predicted image} - \text{ground truth image} \|_2^2$$

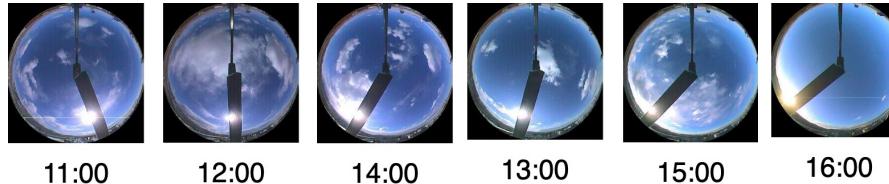
$$\text{expectation loss} = \| \text{prior mean} - \text{posterior mean} \|_2^2$$

$$\text{variance loss} = \| \text{variance of predictions} - \text{variance of expert samples} \|_2^2$$

$$\text{solar loss} = \| \text{best predicted solar irradiance} - \text{ground truth solar irradiance} \|_1$$

# Experiment setups

- Golden, Colorado Dataset:
  - Contains sky images and auxiliary data recorded from 2004 to 2016.
  - Auxiliary data contains date, time, clear-sky irradiance, azimuth angle and zenith angle
  - Data in 2015 and 2016 is used as test sets.



- Evaluation metrics:
  - Normalized mean absolute percentage error (nMAP):
$$nMAP = \frac{1}{N} \sum_{n=1}^N \frac{|p_n - \hat{p}_n|}{\frac{1}{N} \sum_{n=1}^N p_n} \times 100$$
  - Diversity: Average L1 difference of each pair of predictions.

# Result

	nMAP							
	Test 2015				Test 2016			
	+1h	+2h	+3h	+4h	+1h	+2h	+3h	+4h
Siddiqui [3]	<b>17.9</b>	25.2	31.6	39.1	<b>16.9</b>	25.0	31.9	39.5
Our deterministic model	21.6	25.7	30.1	35.6	19.2	23.3	27.2	32.7
Our stochastic model (best prediction)	19.7	<b>21.2</b>	<b>22.5</b>	<b>27.8</b>	17.4	<b>19.1</b>	<b>21.2</b>	<b>25.5</b>

- Our deterministic model outperform the state-of-the-art model for predictions in the far future.
- The best prediction of the stochastic model better than the that of the deterministic models.

[3] Siddiqui, T.A., Bharadwaj, S. and Kalyanaraman, S., 2019, January. A deep learning approach to solar-irradiance forecasting in sky-videos. In *2019 IEEE Winter Conference on Applications of Computer Vision (WACV)* (pp. 2166-2174). IEEE.

# Diversity in the predictions of the stochastic model

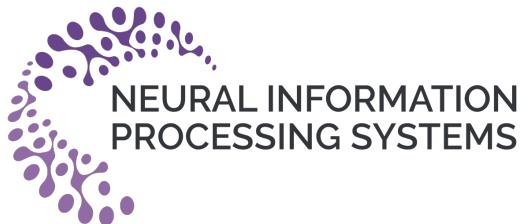
	Diversity							
	Test 2015				Test 2016			
	+1h	+2h	+3h	+4h	+1h	+2h	+3h	+4h
Our stochastic model	77.4	91.5	97.5	100.2	70.1	82.2	88.0	90.0

- The diversity increases as we predict further into the future.
- Our stochastic model is able to capture uncertainties in the future.

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