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# Ammonia Emissions Enhancements with Deep Learning CTM (Forward-Backward) and Remote-sensing Observations

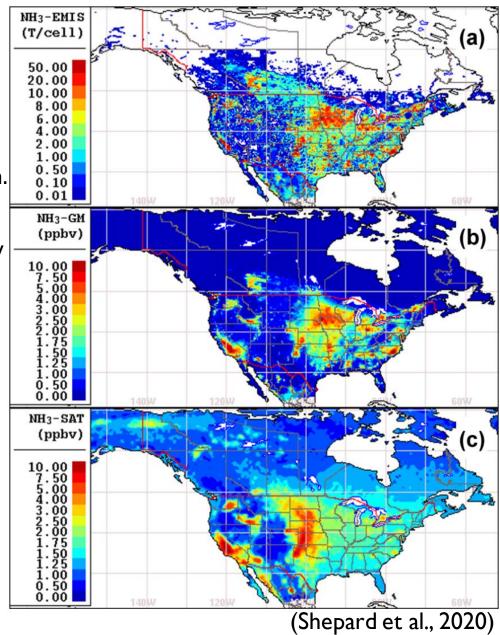
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#### **Motivation**

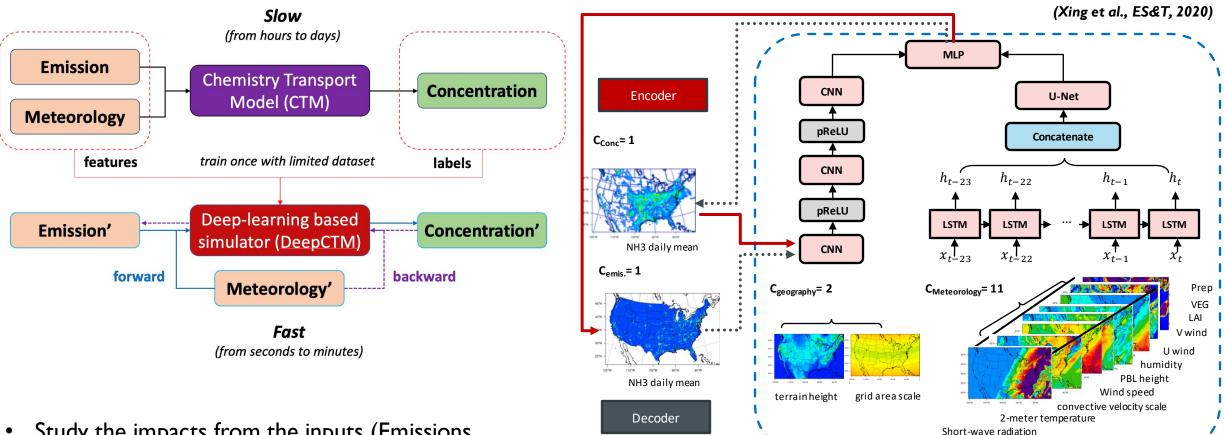
- **NH**<sub>3</sub>: Animal manure, production, and Application of fertilizers
  - Short lifetime of  $NH_3$  ( $\tau = 0.5-5$  days or less)
  - Deposited to the surfaces near its source with relatively high dry deposition.
  - Ecosystem eutrophication
- $NH_4^+$ : Neutralization process of acid gases ( $H_2SO_4$ ,  $HNO_3$  and HCI) by  $NH_3$  at the surface on  $PM_{fine} \le 2.5 \mu m$ 
  - NH<sub>4</sub>HSO<sub>4</sub>, [NH<sub>4</sub>]<sub>2</sub>SO<sub>4</sub>, [NH<sub>4</sub>]<sub>3</sub>H[SO<sub>4</sub>]<sub>2</sub>, NH<sub>4</sub>NO<sub>3</sub>, NH<sub>4</sub>Cl,,,,
  - Role of degradation of visibility in the local area
  - Longer lifetime of NH<sub>4</sub><sup>+</sup> approximately 5~10 days
  - Ecosystem acidification from ammonium oxidation in the atmosphere
  - Sensitive spatiotemporal patterns by meteorology, events, and acid gases.
- Top-down vs. Bottom-up NH<sub>3</sub> emissions are quite different
  - Swine NH<sub>3</sub> emissions from Iowa and NC relatively overestimated?
  - Beef  $\rm NH_3$  emissions from Texas, Kentucky, and Nebraska are the main  $\rm NH_3$  in Central US



# DeepCTM-VAE Applications with Observations

- Objectives
  - Bottom-up NH<sub>3</sub> Emissions Comparison against Top-down NH<sub>3</sub> Emissions
  - Understand the differences and identify the solution to enhance the bottom-up Inventory
- Deep-Learning Chemical Transport Model (Decorder)
  - Development of DeepCTM (forward) based on the CMAQ CONUS I 2 simulation to understand the non-linear relationship between emissions and air quality
- Variational AutoEncoder (VAE) Applications with CrIS Remote-sensing Observations
  - Development of VAE (Backward) over the U.S.
  - Data Assimilation: Surface NH<sub>3</sub> concentrations based on the Crls remote-sensing and monitoring observations.
  - Development of top-down NH<sub>3</sub> emissions inventory over the U.S. and compare them against the bottom-up emissions (USEPA NEI)

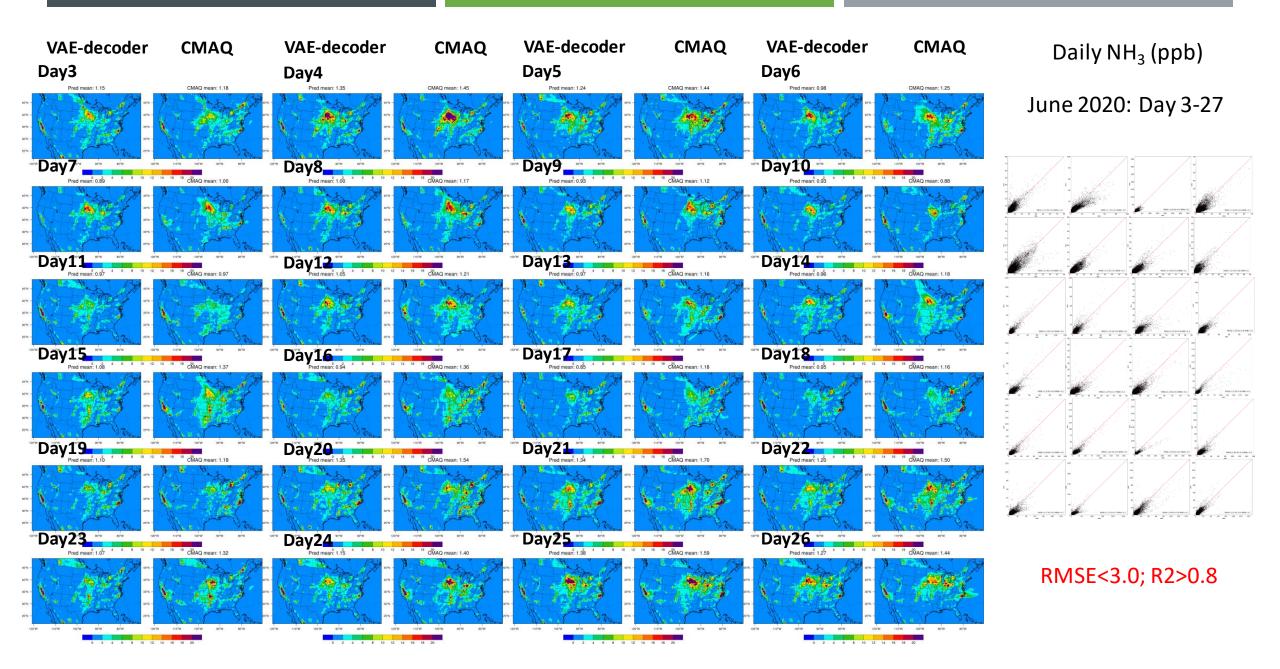
## DeepCTM: Deep-Learning CTM



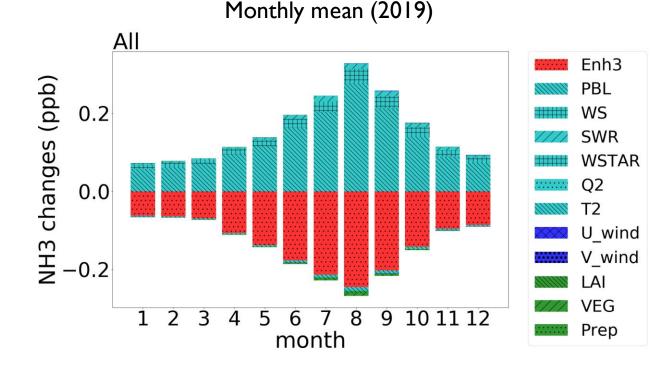
- Study the impacts from the inputs (Emissions, Meteorology, Concentrations, and Time)
- Computationally highly efficient (<2 minutes) to execute DeepCTM once the training of DeepCTM is complete

**UNet:** capture the cross-space interaction **LSTM:** capture the temporal variation

### **Performance of VAE-decoder (forward: emis->conc)**



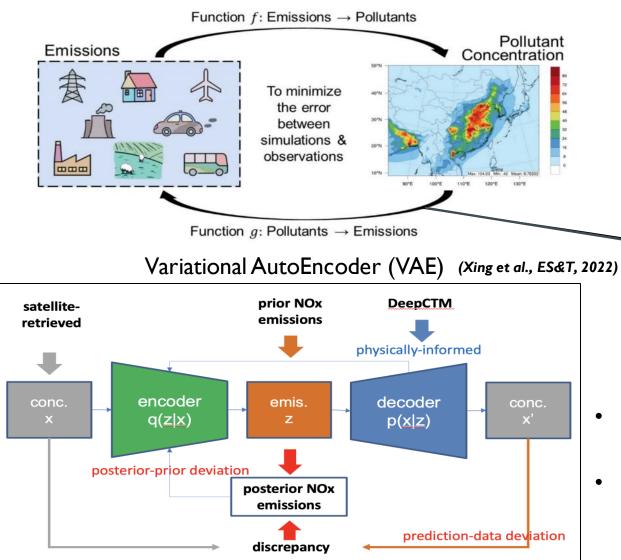
#### The Feature Impacts on NH<sub>3</sub> Concentrations

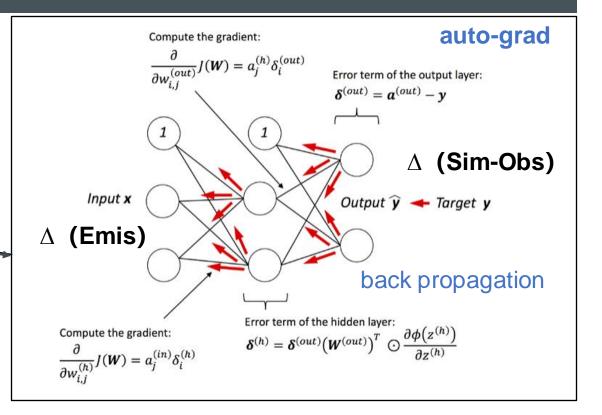


#### modulating each feature one-by-one

- **T2**: -2K
- others: -20%
- NH<sub>3</sub> concentration is contributed by NH<sub>3</sub> emissions (Enh3) and initial NH<sub>3</sub> condition (Inh3)
- Met variables will modulate the NH3 concentration sensitivity to NH3 emissions
- DeepCTM exhibits high efficiency in identifying the dominant factor to photochemical formation

### Variational AutoEncoder (VAE):Top-down Emissions





- Based on emission-concentration relationship from DeepCTM, we can accurately estimate the differences of emissions
- Because of the in-depth knowledge over various impactors (emissions, meteorology, initial conc.), we can proportionally correct the emissions only.

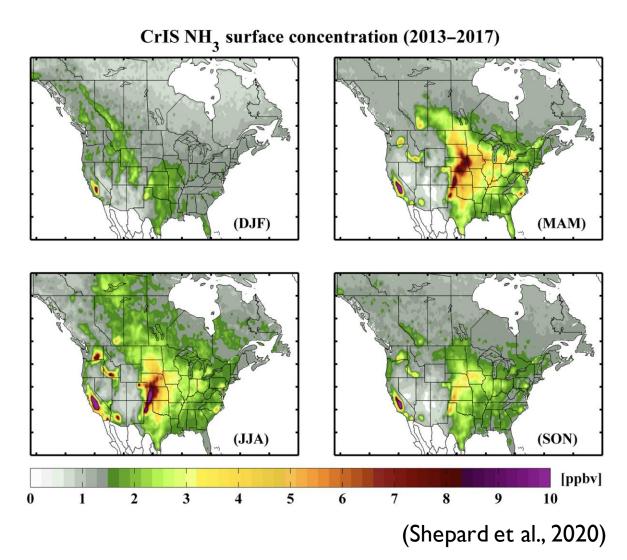
# CrIS Fast Physical Retrieval (CFPR) Algorithm for NH<sub>3</sub>

(Shepard et al., 2020)

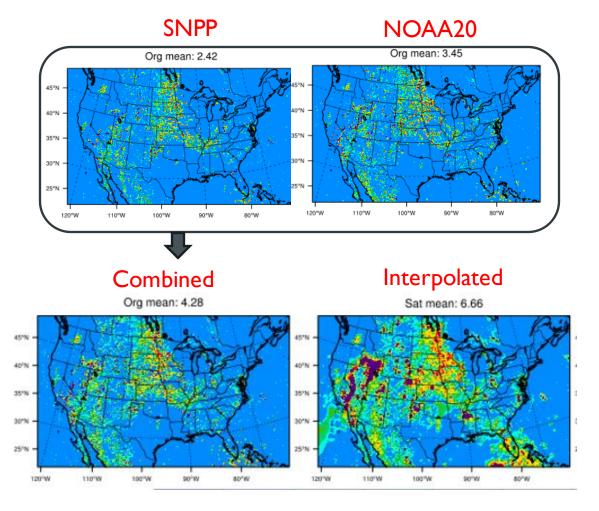
	Satellite	Spatial Resolution (km²)	Spatial Coverage	Temporal Coverage	Retrieved Quantity	Comments
NH <sub>3</sub>	Cross-track Infrared Sounder <b>(CrIS)</b> (2012-2038)	14 (diameter)	Global	Twice a day CrIS-SNPP @ ~13:30 ( <b>day</b> ) ~1:30 (night) CrIS-NOAA20@ ~14:25 ( <b>day</b> ) 2:25 (night)	Profile level volume mixing ratio values (ppbv)	<ul> <li>Limited vertical resolution</li> <li>Version 1.6</li> <li>Shephard et al., ACPD, 2020</li> <li>CFPR CrIS SNPP : May 2012 to May 2021</li> <li>CFPR CrIS NOAA : March 2019 to present</li> </ul>

- CrIS is most sensitive to NH<sub>3</sub> between 950 and 700 mb (~0.5 to 3 km) with minimum 0.3 ppbv detectability
  - Not equally sensitive in the vertical and varies from profile-to-profile
  - Surface retrieved values are driven by sensitivity in the boundary layer
- CFPR has been validated with U.S. AMoN (Ammonia Monitoring Network) and other ground– based FTIR (Fourier Transform Infrared) spectroscopy observations to determine the errors
  - AMoN: Crls ~15% higher with a good correlation of r ~ 0.8 and a slope of 1.02
  - FTIR: Crls ~30% higher (SD~40%)
- Applications to data assimilations, data fusion and model-based emissions inversions for Confined Animal Feeding Operation (CAFO) facilities located remotely.

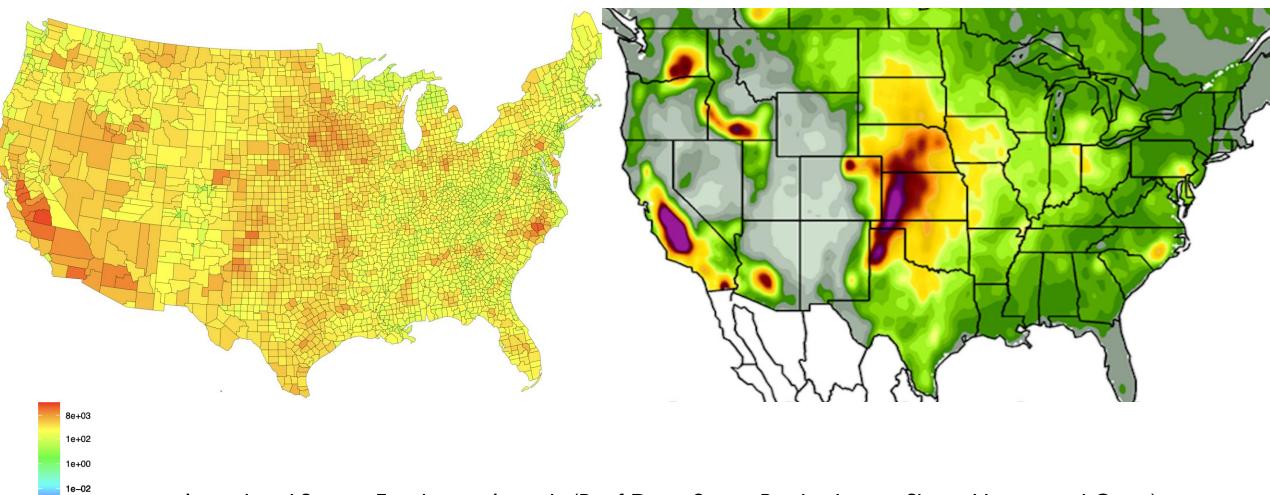
### NH<sub>3</sub> Crls CFPR Spatiotemporal Patterns



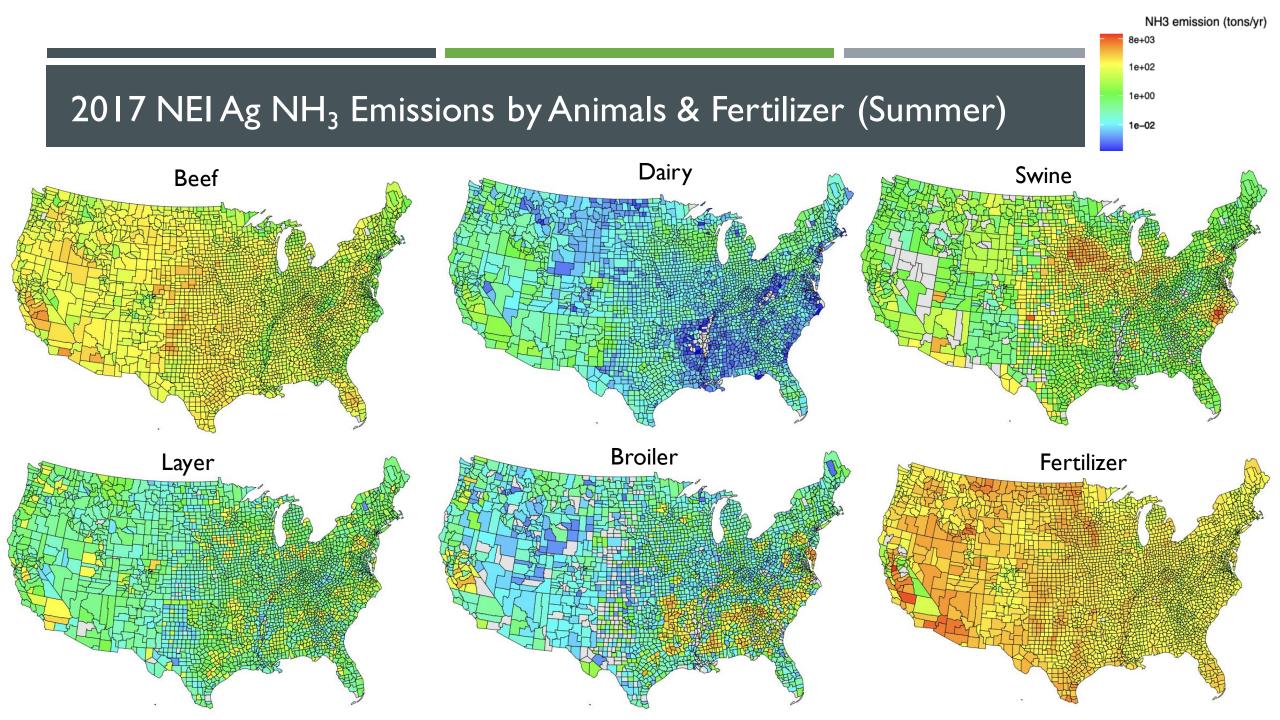
#### NOAA Satellites Surface NH<sub>3</sub> Measurements



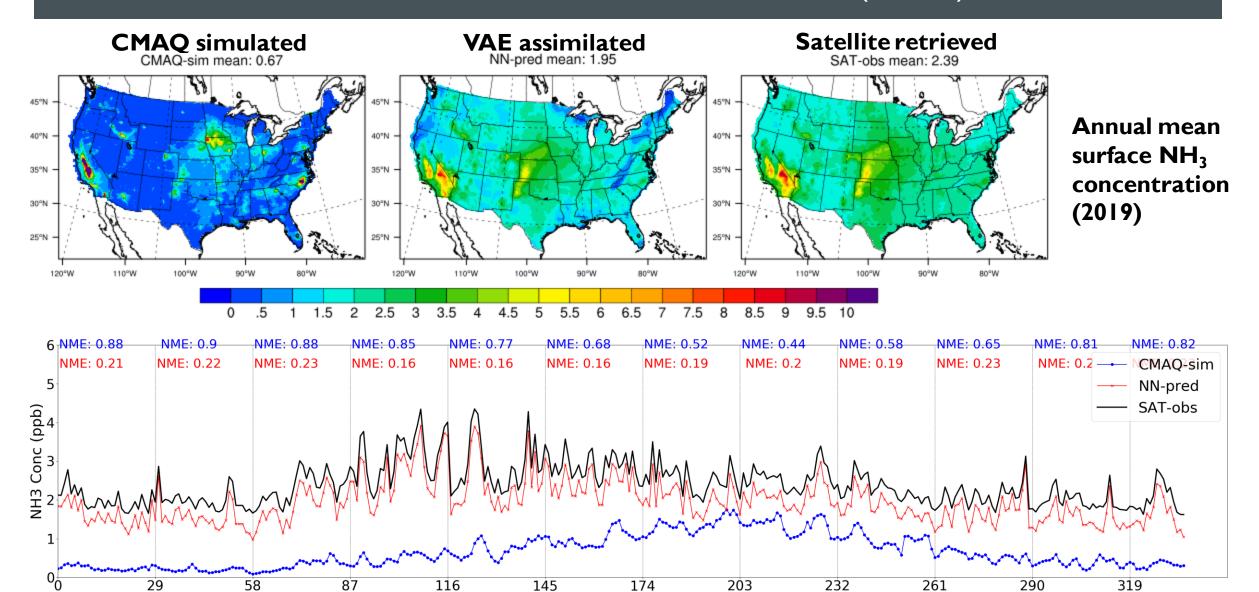
### 2017 NEI NH<sub>3</sub> Emissions from Agricultural Sector



Agricultural Sector: Fertilizer + Animals (Beef, Dairy, Swine, Broiler, Layers, Sheep, Horses, and Goats)

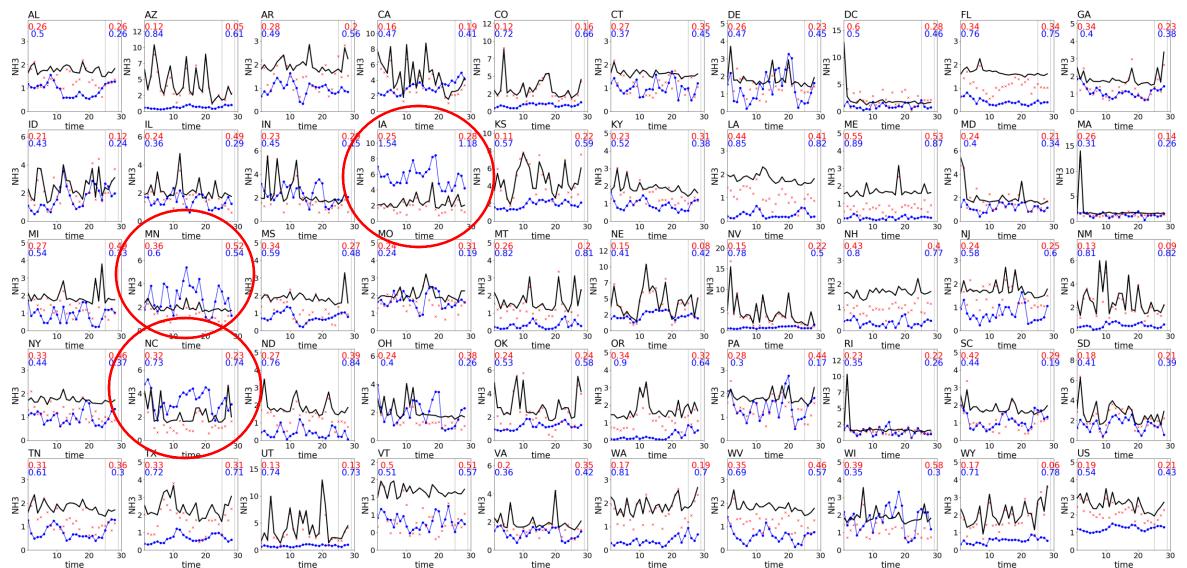


### Performance of VAE-assimilation (2019)

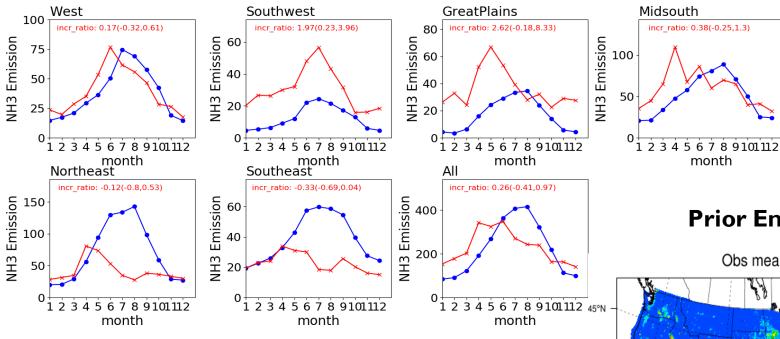


### Performance of VAE-assimilation (June 2019)

Blue: CMAQ Simulation Red:VAE-assimilation Black: CFPR Satellite



# NH<sub>3</sub> Emissions Comparison (2019)



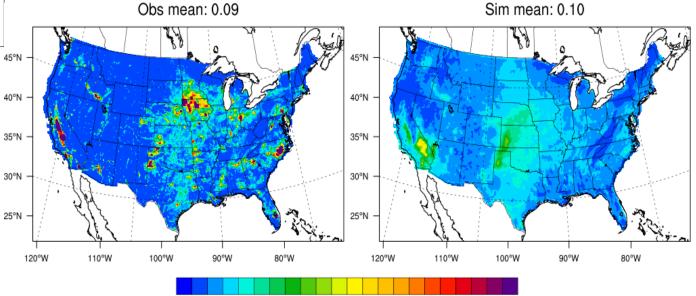
- Important to utilize the spatial and temporal ٠ coverages from the CrIs NH<sub>3</sub>
- **<u>Relatively</u>** underestimated NH<sub>3</sub> from Beef vs • overestimated NH<sub>3</sub> from Swine

Indicating that the significant • differences over the Great Plains and Southwest regions

**Prior Emissions** 



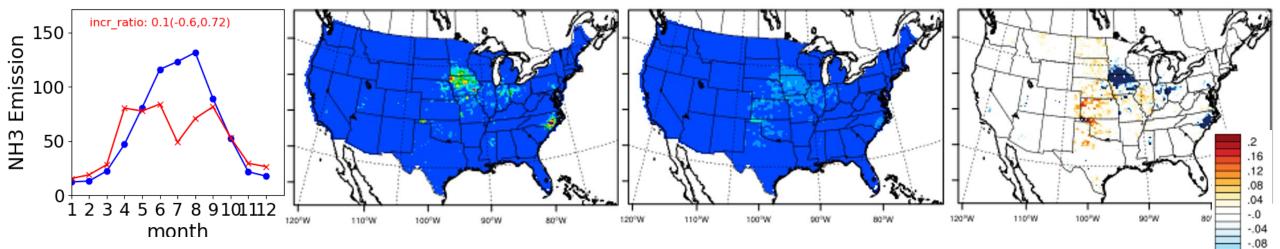
Sim mean: 0.10



0 .05 .1 .15 .2 .25 .3 .35 .4 .45 .5 .55 .6 .65 .7 .75 .8 .85 .9 .95 1

### Sector-level NH<sub>3</sub> Emissions Comparison (2019)

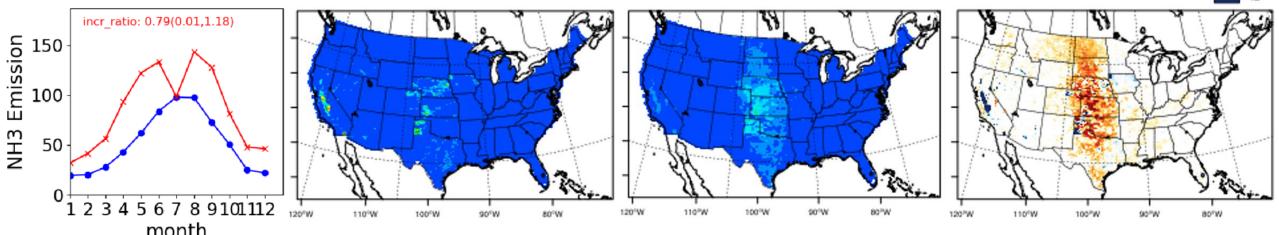
#### **Swine**



-.12

-.16 -.2

#### **Beef Cattle**



# Findings

- A significant amount of surface NH<sub>3</sub> concentrations changes by location and time/dates
- Requires a better spatiotemporal representation of NH<sub>3</sub> from livestock wastes
- DeepCTM/VAE performed well in capturing the spatial and temporal differences of NEI Livestock Waste NH<sub>3</sub> against the observations
- Identified the sector-level (anima-specific) NH<sub>3</sub> differences for the calibrations of animal-specific NEI Emissions from the Farm Emissions Model used in NEI development
- Swine and Beef Cattle shows the lack of spatiotemporal representations
  - Swine: <u>Overestimations</u> of NH<sub>3</sub> over the areas in Iowa and NC (most swine-dense area)
  - Beef: <u>Underestimation</u> of NH<sub>3</sub> over the areas in TX, KY, CO, NE, SD, ND, and MT, while the <u>overestimation</u> occurred over CA.
    - Regional farm management practices input to the FEM need an update to address this issue.

# Thank you!

#### ACKNOWLEDGEMENT

- NOAA Atmospheric Chemistry, Carbon Cycle and Climate (AC4) [Grant# NA21OAR4310225]
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