# U.S. Meat Production Analysis and Forecasting

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# STAT442 Unsupervised Machine Learning

### **Abstract**

This study analyzes U.S. meat production statistics and forecasts future production using machine learning techniques. A Variational Autoencoder (VAE) and a transformer-based forecasting model are employed to capture and predict trends.

#### **Data Source**

USDA Domestic Meat Production, Historical.

# **Analytical Objectives**

Correlation Analysis and Time-Series Predictions.

# **Data Cleaning**

Limited the analysis to the period of Jan-2001 to Jan-2025.

Normalized using MinMaxScaler to prevent negative values.

Standardized numeric and datetime formats

# **Statistical & Machine Learning Procedures**

Descriptive Analysis: Correlation heatmap, timeseries visualization, and scatter plots.

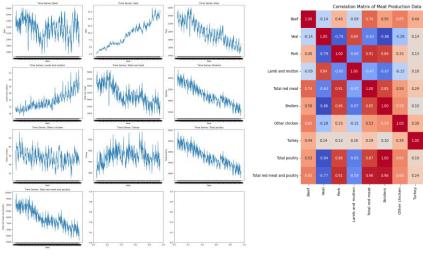
Predictive Analysis: Variational Autoencoder (VAE) and Transformer-based forecasting.

## **Software & Tools**

Python (Pandas, NumPy, Seaborn, Matplotlib, Keras, PyTorch).

# **Model Selection**

Captures latent patterns in historical data. Generates realistic production forecasts using deep learning.



### **VAE** results:

	PF	ealctea	TUTUTE	trends	(origina)	r scare):			
		Date	Fea	sture_1	Feature_2	2 Feat	cure_3	Feature_4	Feature_5
	0	Jan-25	2069	683594	6.779376	1886.9	89136	11.834916	4078.976562
	1	Feb-25	2178	073975	7.843806	1998.5	28687	12.213923	4221.109863
	2	Mar-25	2213	849854	8.120829	2031.9	905029	12.247731	4256.477539
	3	Apr-25	2226	565186	8.191932	2 2042.1	193604	12.242837	4266.317871
	4	May-25	2231	521729	8.209249	2045.5	660181	12.241272	4269.277344
	5	Jun-25	2233	714844	8.213400	2046.8	328735	12.244476	4270.285645
		Feat	ure 6	Feature	7 Featu	ire 8	Featur	e 9 Featu	ire_10
١	0	3055.5	31250	41.5156	02 457.27	74445 36	39.246	826 7594.3	73535
,	1	3227.75	56592	43.8292	85 466.04	16814 38	351.158	691 7978.1	41602
	2	3274.89	92822	44.6007	96 465.03	35797 39	914.104	736 8079.8	801270
	3	3289.00	32441	44.9027	18 463.49	9747 39	933.799	561 8104.3	31543
	4	3293.98	39746	45.0415	73 462.39	99292 39	940.562	988 8108.3	864258
	5	3296.39	93799	45.1150	40 461.79	94006 39	943.285	156 8107.6	557715

# Transformer results:

	8	1	2	3	4	5	6	7	1
0	1741.0094	32.8152	1309.6686	31.1271	2822.1050	1405.9308	380.1586	1956.3875	
1	1886.8129	26.1994	1275.5325	24.3964	3365.1445	1713.3722	391.3523	2178.2158	
2	1960.1750	21.9284	1373.6909	21.6222	3632.3477	1988.3779	406.1849	2433.3838	
3	2003.7330	18.0578	1505.2623	18.4523	3837.4453	2326.9846	429.4685	2763.1968	
4	2068.4082	14.1441	1656.1318	16.3812	3932.5452	2686.8577	449.2589	3133.7710	
5	2093.5562	10.8822	1793.6283	14.9633	3918.9602	2940.5046	460.7080	3417.1995	
6	2113.3562	8.9785	1896.0061	13.3408	3949.0161	3109.6438	468,7467	3625.2012	
7	2117.9890	8.1316	1961.2762	12.0100	4004.7803	3216.7402	474.0439	3765.1921	
8	2123.5330	7.9407	1994.4252	11.6604	4055,2300	3270.8459	475.9704	3838.4641	
9	2125,2400	7.8247	2006.6818	11.5188	4073.0515	3291.5183	475.9724	3863.3645	
10	2125.9209	7.7744	2012.1145	11.4549	4081.2314	3300.5908	475.9526	3874.3167	
11	2126.2141	7.7519	2014.5176	11.4265	4084.8560	3304.6008	475.9412	3879.1487	

#### **Model Architecture**

VAE: LSTM Layer with 50% dropout.

Transformer: Three RELU layers, two 10% dropout layers.

#### Results

Strong correlations observed across all meat categories. Annual peaks and dips are identifiable.

# **Forecasting Results**

VAE-generated forecasts captured broad trends but lacked precision in monthly fluctuations.

Transformer-based models improved performance by leveraging additional historical data but still lacks accuracy.

### Conclusion

Both models are capable of time-series forecasting, but are difficult to train, tune and understand with limited experience.

Transformer-based models offered improvements but require further refinement.

Future research should explore additional tuning and alternative deep-learning approaches.

### References

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https://hunterheidenreich.com/posts/modernvariational-autoencoder-in-pytorch/

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