

# Modeling Bitcoin Price Trends Using Regression and Gradient Descent

CSCI 240 Final Project

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— Price

Objective: Predict Bitcoin prices using regression analysis

Models used:

Linear Regression

Quadratic Regression

Exponential Regression (normalized and unnormalized)

Price

100000

80000

60000

40000

20000

Jan

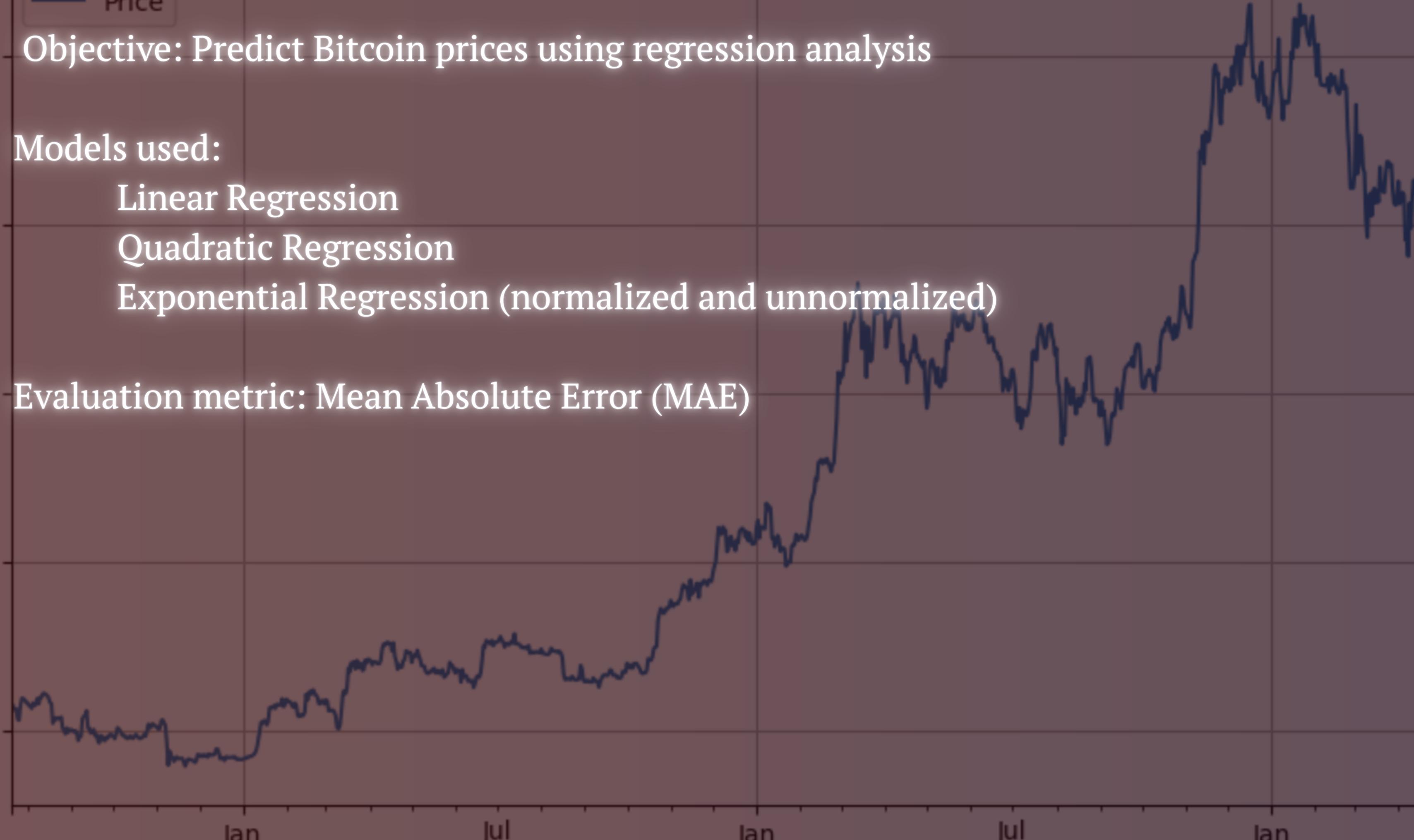
Jul

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Evaluation metric: Mean Absolute Error (MAE)



- Linear – models a straight-line relationship

$$y = mx + b$$

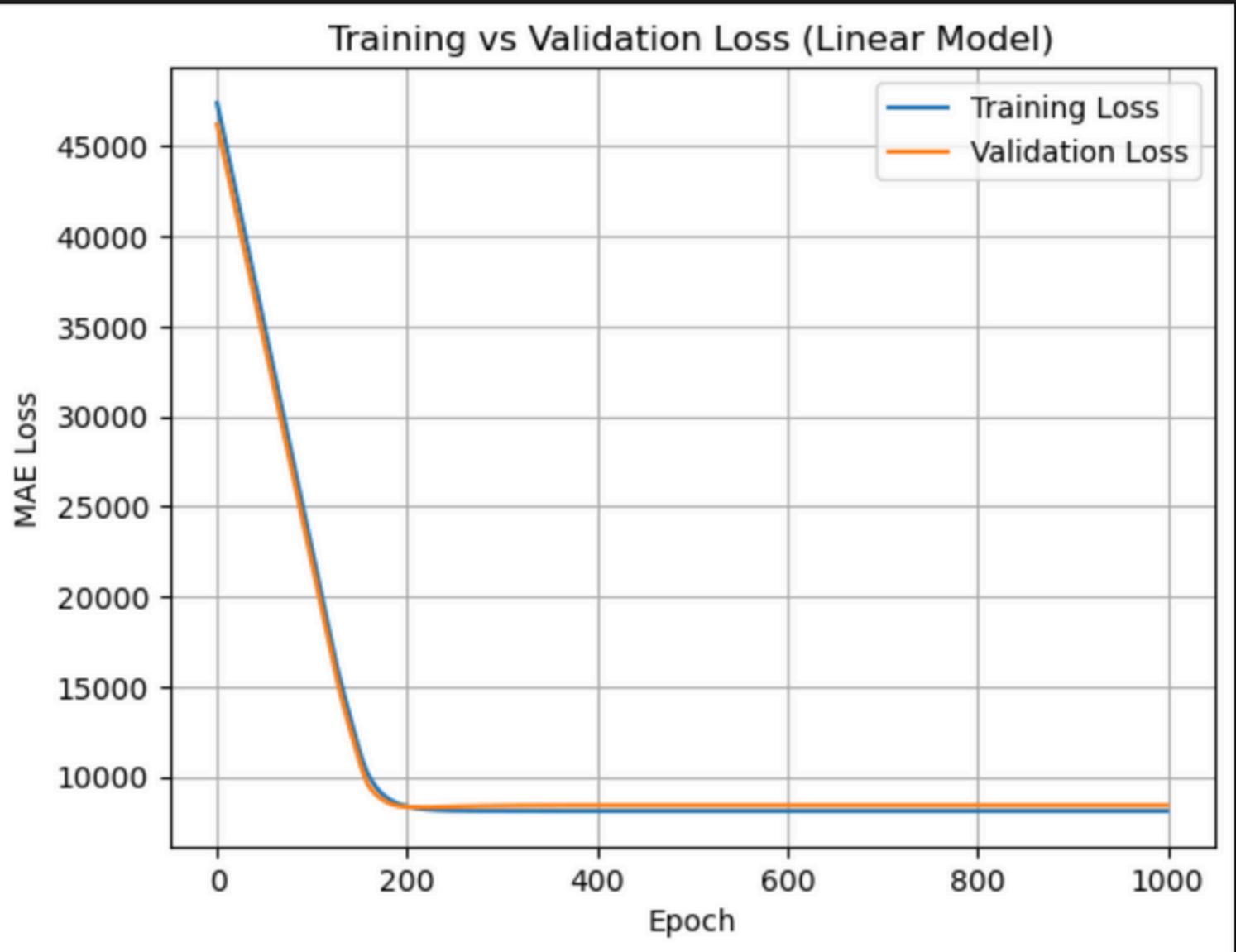
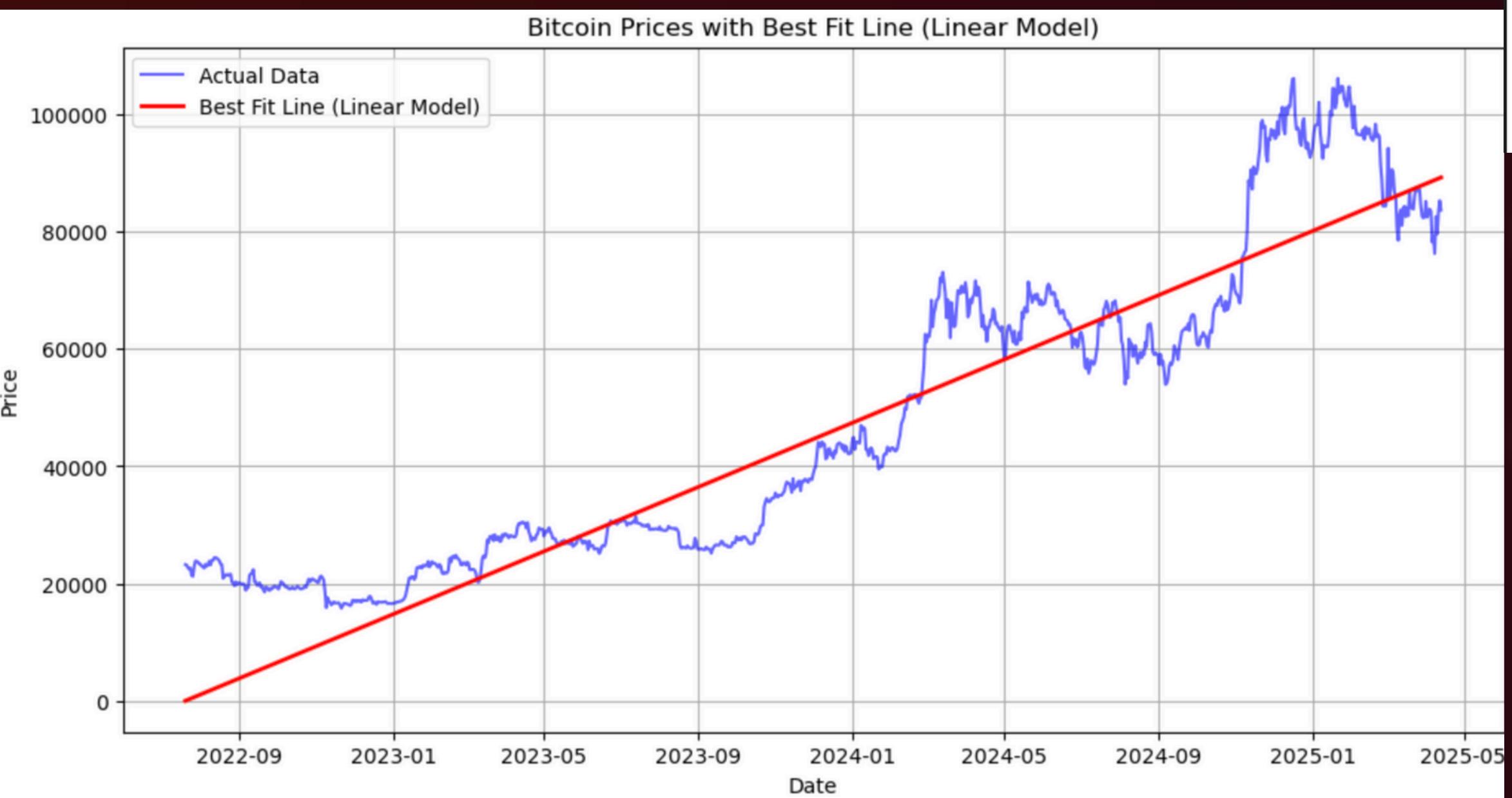
- Quadratic – captures curvature in the trend

$$\hat{y} = ax^2 + bx + c$$

- Exponential – captures compounding growth (sensitive to input scale)

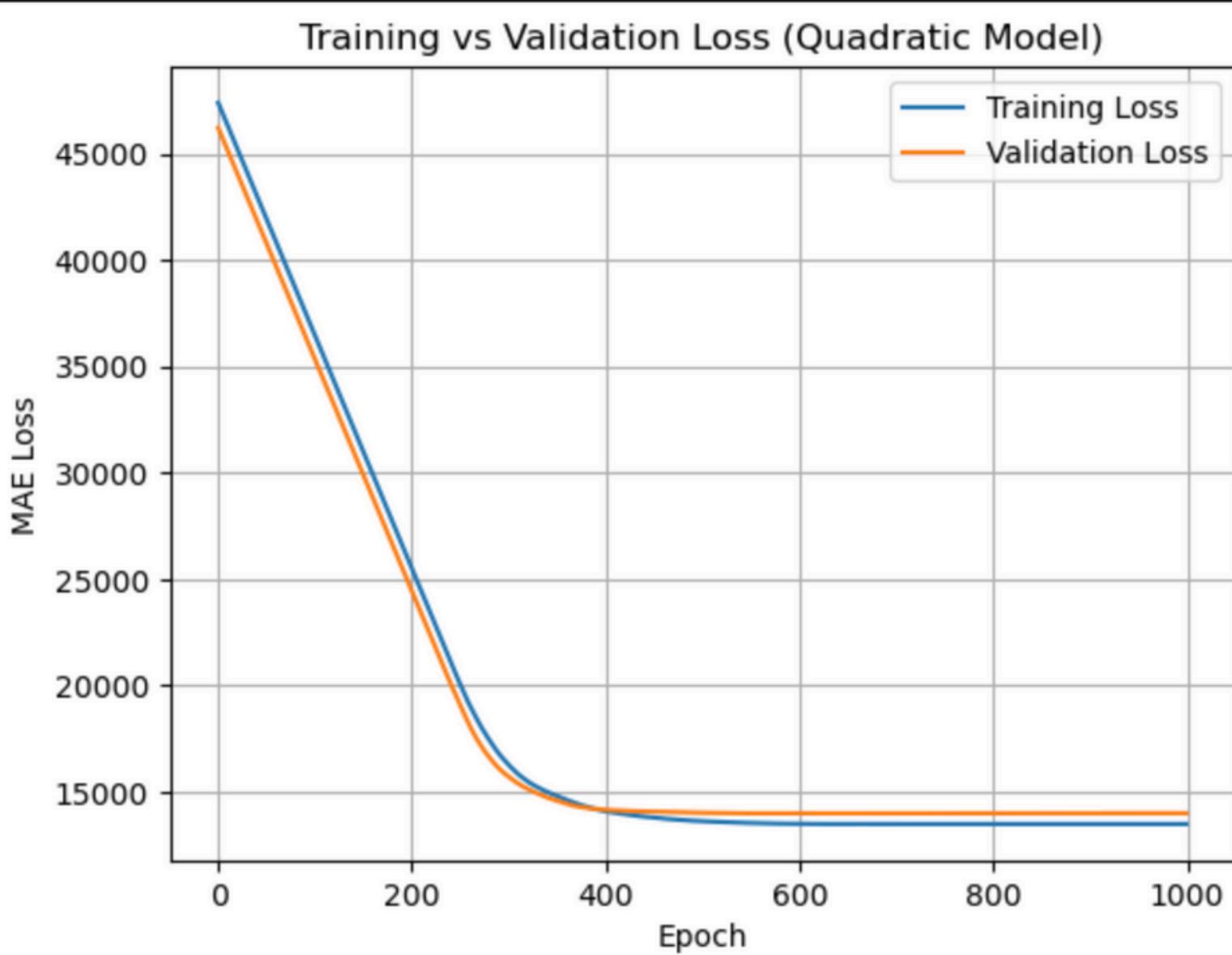
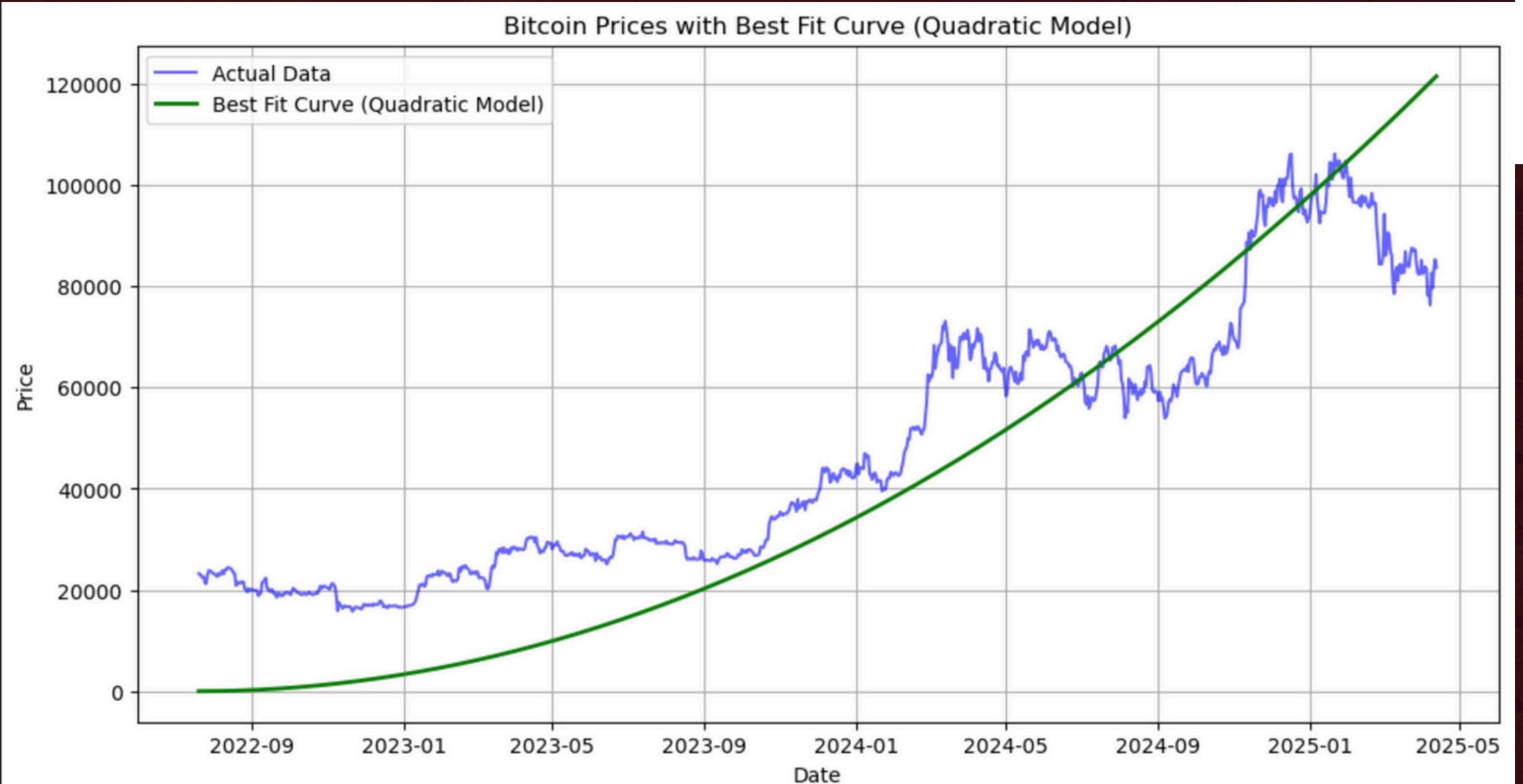
$$\hat{y} = a \cdot e^{bx} + c$$

# Linear Model Results



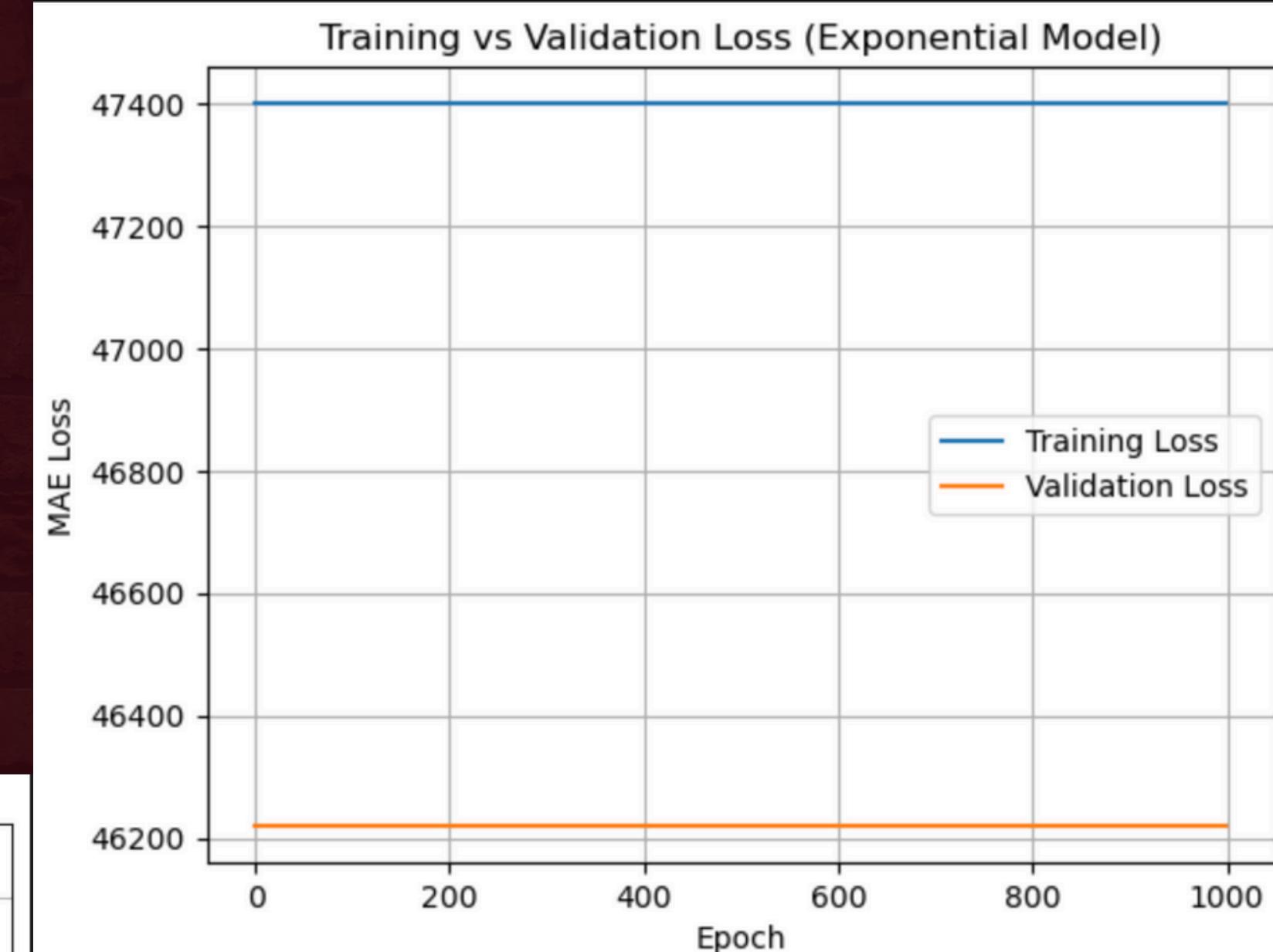
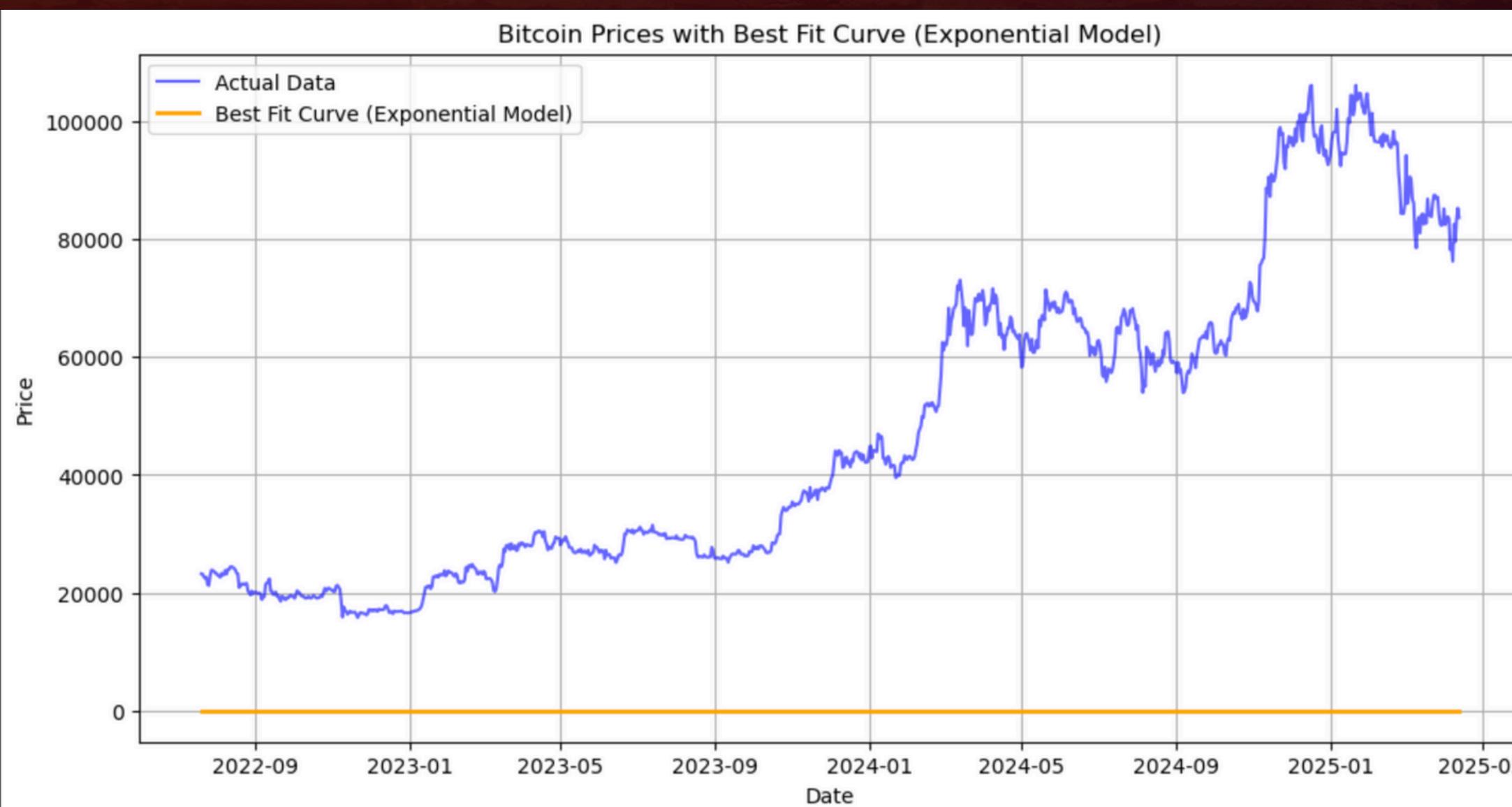
Test MAE: 8,025.69  
Captures general trend but underfits volatility  
Train & validation losses closely aligned

# QUADRATIC MODEL



- Test MAE: 13,596.98
- Models curvature better than linear
- Overshoots prices in later stages (late 2024)

# Exponential Model (Unnormalized)



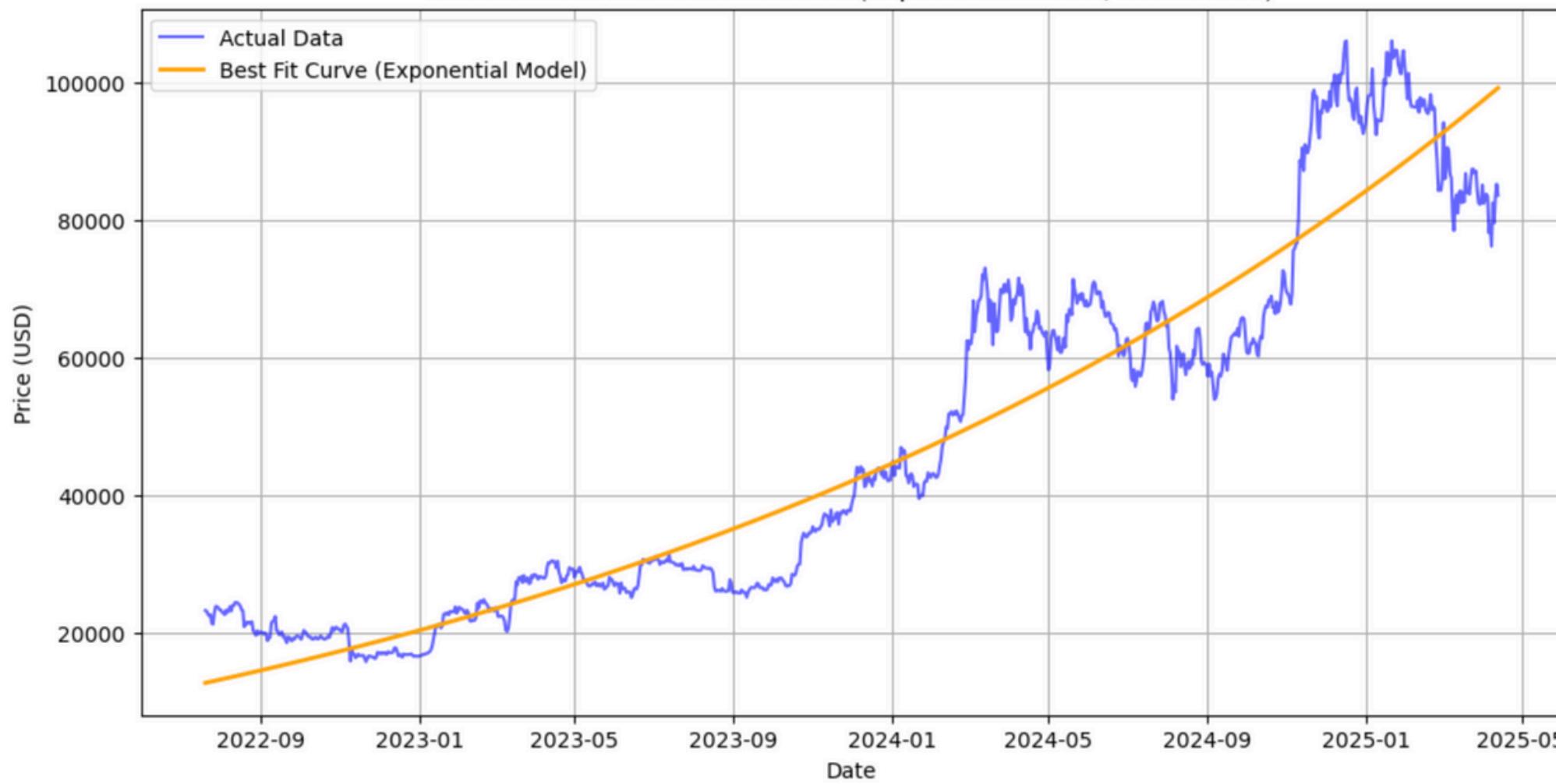
Test MAE: 48,540.17

Failed due to gradient explosion  
(large input magnitudes)

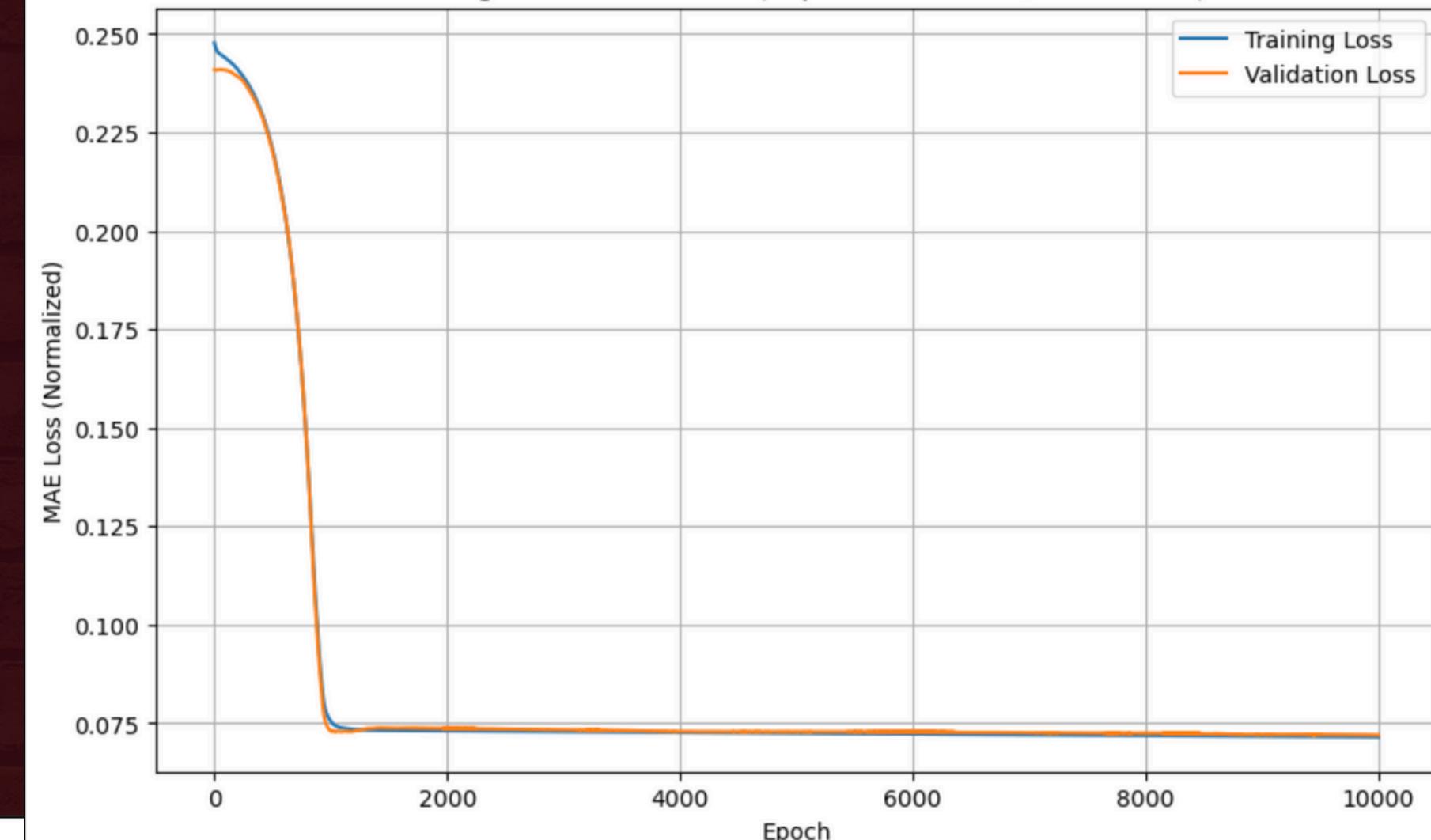
Predictions were flat or unstable

# Exponential Model (Normalized)

Bitcoin Prices with Best Fit Curve (Exponential Model, Normalized)



Training vs Validation Loss (Exponential Model, Normalized)



Test MAE: 6,234.13 (Best)  
After normalization, model converged well  
Accurately modeled compounding growth and  
plateau

# Model Comparison

Model	Train MAE	Val MAE	Test MAE		Notes
Linear	8109.81	8419.15	8025.69		Underfit, stable
Quadratic	13485.80	13993.05	13596.98		Overshoots late prices
Exponential (Unnormalized)	47400.18	46220.12	48540.17		Failed to learn
Exponential (Normalized)	~0.07 (scaled)	~0.07 (scaled)	6234.13		Best model

# CONCLUSION

This project provided a valuable opportunity to apply regression models to a real-world dataset and evaluate their effectiveness in capturing Bitcoin price trends. I implemented linear, quadratic, and exponential models from scratch using custom gradient descent, gaining a deeper understanding of training dynamics and model behavior. The results showed that while the linear model provided a stable baseline and the quadratic model offered flexibility, both struggled to generalize accurately. The unnormalized exponential model failed due to instability, but after applying normalization, the exponential model achieved the lowest test MAE (6234.13) and best overall fit capturing both the early stability and late-stage growth in prices.