In [7]: pip install pandas numpy matplotlib seaborn psutil

Defaulting to user installation because normal site-packages is not writeable Requirement already satisfied: pandas in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (2.2.0)

Requirement already satisfied: numpy in /shared/spack/opt/spack/linux-amzn2-s kylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/e nvs/neuro140/lib/python3.11/site-packages (1.26.3)

Requirement already satisfied: matplotlib in /shared/spack/opt/spack/linux-am zn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmeg d3a/envs/neuro140/lib/python3.11/site-packages (3.8.2)

Requirement already satisfied: seaborn in /shared/spack/opt/spack/linux-amzn2 -skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3 a/envs/neuro140/lib/python3.11/site-packages (0.13.1)

Requirement already satisfied: psutil in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (6.1.1)

Requirement already satisfied: python-dateutil>=2.8.2 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from pandas) (2.9.0.post0)

Requirement already satisfied: pytz>=2020.1 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkm egd3a/envs/neuro140/lib/python3.11/site-packages (from pandas) (2025.1) Requirement already satisfied: tzdata>=2022.7 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from pandas) (2025.1) Requirement already satisfied: contourpy>=1.0.1 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (1.3.1)

Requirement already satisfied: cycler>=0.10 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkm egd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (0.12.1) Requirement already satisfied: fonttools>=4.22.0 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (4.5 5.8)

Requirement already satisfied: kiwisolver>=1.3.1 in /shared/spack/opt/spack/l inux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap 3ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (1.4.7)

Requirement already satisfied: packaging>=20.0 in /shared/spack/opt/spack/lin ux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3y lkmegd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (24.2) Requirement already satisfied: pillow>=8 in /shared/spack/opt/spack/linux-amz n2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd 3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (10.2.0) Requirement already satisfied: pyparsing>=2.3.1 in /shared/spack/opt/spack/linux-amzn2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3 ylkmegd3a/envs/neuro140/lib/python3.11/site-packages (from matplotlib) (3.2.1)

Requirement already satisfied: six>=1.5 in /shared/spack/opt/spack/linux-amzn 2-skylake_avx512/gcc-14.1.0/miniconda3-24.3.0-zxx5jostrj4myhf7bi3oap3ylkmegd3 a/envs/neuro140/lib/python3.11/site-packages (from python-dateutil>=2.8.2->pa ndas) (1.17.0)

Note: you may need to restart the kernel to use updated packages.

```
In [8]: ### IMPORTS

    from resource_metrics_and_visualizations import (
        add_metrics_to_predictor,
        ResourceMetrics
)
    from paper_visualizations import PaperVisualizations
    from integration_example import run_tte_prediction_with_metrics

In [9]: # Run the enhanced pipeline
    predictor, model_comparison, figures = run_tte_prediction_with_metrics(
        './2241data/train_operational_readouts.csv',
        './2241data/train_tte.csv',
        config_path=None # Or provide a path to your config file
)

# Generate paper summary
    from integration_example import generate_paper_summary
    summary_file = generate_paper_summary(predictor, "tte_prediction_paper_findir)
```

```
Running on CPU (GPU disabled in config)
--- LOADING DATA ---
Loaded data with 39716 rows and 500 vehicles
--- CLEANING DATA ---
Data cleaned: 39716 rows remaining with 105 features
--- ENGINEERING FEATURES ---
Sample counts per variability level:
variability
hiah
         13504
low
         13106
medium
         13106
Name: count, dtype: int64
Feature engineering complete. Total features: 124
--- CLEANING DATA ---
Data cleaned: 39716 rows remaining with 124 features
--- ENGINEERING FEATURES ---
Sample counts per variability level:
variability
high
         13504
low
         13106
         13106
medium
Name: count, dtype: int64
Feature engineering complete. Total features: 124
--- ANALYZING SCENARIOS ---
Analyzing low variability scenario...
Processing 13106 samples for low variability
Created 8283 sequences for evaluation
Processing fold 1 of 2
Training with 2761 sequences, testing with 2761 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (2761, 10, 124)
Epoch 1/15
2025-05-09 18:25:30.627441: E external/local_xla/xla/stream_executor/cuda/cud
a_driver.cc:274] failed call to cuInit: CUDA_ERROR_NO_DEVICE: no CUDA-capable
device is detected
2025-05-09 18:25:30.627477: I external/local xla/xla/stream executor/cuda/cud
a_diagnostics.cc:129] retrieving CUDA diagnostic information for host: gpu-pa
rallel-dy-gpu-parallel-cr-9
2025-05-09 18:25:30.627483: I external/local_xla/xla/stream_executor/cuda/cud
a_diagnostics.cc:136] hostname: gpu-parallel-dy-gpu-parallel-cr-9
2025-05-09 18:25:30.627604: I external/local_xla/xla/stream_executor/cuda/cud
a diagnostics.cc:159] libcuda reported version is: 535.216.1
2025-05-09 18:25:30.627623: I external/local xla/xla/stream executor/cuda/cud
a diagnostics.cc:163] kernel reported version is: 535.216.1
2025-05-09 18:25:30.627627: I external/local_xla/xla/stream_executor/cuda/cud
a_diagnostics.cc:241] kernel version seems to match DSO: 535.216.1
```

```
69/69 [========================== ] - 2s 7ms/step - loss: 0.1313 - val_los
s: 0.0613
Epoch 2/15
69/69 [=================== ] - 0s 4ms/step - loss: 0.0383 - val_los
s: 0.0473
Epoch 3/15
69/69 [==================== ] - 0s 4ms/step - loss: 0.0267 - val los
s: 0.0404
Epoch 4/15
69/69 [================== ] - 0s 4ms/step - loss: 0.0214 - val los
s: 0.0392
Epoch 5/15
69/69 [================== ] - 0s 4ms/step - loss: 0.0181 - val los
s: 0.0384
Epoch 6/15
69/69 [================ ] - 0s 3ms/step - loss: 0.0165 - val_los
s: 0.0385
Epoch 7/15
69/69 [=================== ] - 0s 4ms/step - loss: 0.0155 - val_los
s: 0.0309
Epoch 8/15
69/69 [=============== ] - 0s 3ms/step - loss: 0.0143 - val_los
s: 0.0314
Epoch 9/15
69/69 [=============== ] - 0s 3ms/step - loss: 0.0141 - val los
s: 0.0325
Epoch 10/15
69/69 [================ ] - 0s 3ms/step - loss: 0.0136 - val_los
s: 0.0350
87/87 [========= ] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 10.0
Ridge Regression successful
Best ES smoothing level: 0.9
Exponential Smoothing successful
Ensemble model created successfully
Found 4 extreme prediction errors. Capping for metric calculation.
Found 4 extreme prediction errors. Capping for metric calculation.
Processing fold 2 of 2
Limiting training sequences from 5522 to 3000
Training with 3000 sequences, testing with 2761 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (3000, 10, 124)
Epoch 1/15
s: 0.0715
Epoch 2/15
s: 0.0506
Epoch 3/15
s: 0.0353
Epoch 4/15
s: 0.0243
Epoch 5/15
```

```
s: 0.0175
Epoch 6/15
s: 0.0149
Epoch 7/15
75/75 [=========================== ] - 0s 4ms/step - loss: 0.0185 - val los
s: 0.0135
Epoch 8/15
s: 0.0117
Epoch 9/15
s: 0.0113
Epoch 10/15
s: 0.0101
Epoch 11/15
s: 0.0095
Epoch 12/15
75/75 [=============== ] - 0s 3ms/step - loss: 0.0136 - val_los
s: 0.0098
Epoch 13/15
s: 0.0087
Epoch 14/15
s: 0.0086
Epoch 15/15
75/75 [=============== ] - 0s 3ms/step - loss: 0.0129 - val los
s: 0.0087
87/87 [======== ] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 1.0
Ridge Regression successful
Best ES smoothing level: 0.3
Exponential Smoothing successful
Ensemble model created successfully
Analyzing medium variability scenario...
Processing 13106 samples for medium variability
Created 8942 sequences for evaluation
Processing fold 1 of 2
Training with 2982 sequences, testing with 2980 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (2982, 10, 124)
Epoch 1/15
75/75 [===========] - 2s 7ms/step - loss: 0.0674 - val_los
s: 0.0375
Epoch 2/15
s: 0.0213
Epoch 3/15
s: 0.0200
```

```
Epoch 4/15
s: 0.0193
Epoch 5/15
s: 0.0148
Epoch 6/15
s: 0.0142
Epoch 7/15
s: 0.0102
Epoch 8/15
s: 0.0111
Epoch 9/15
75/75 [=========================== ] - 0s 4ms/step - loss: 0.0075 - val los
s: 0.0160
Epoch 10/15
s: 0.0115
94/94 [=======] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 0.01
Ridge Regression successful
Best ES smoothing level: 0.9
Exponential Smoothing successful
Ensemble model created successfully
Processing fold 2 of 2
Limiting training sequences from 5962 to 3000
Training with 3000 sequences, testing with 2980 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (3000, 10, 124)
Epoch 1/15
s: 0.0264
Epoch 2/15
s: 0.0125
Epoch 3/15
s: 0.0065
Epoch 4/15
s: 0.0062
Epoch 5/15
s: 0.0041
Epoch 6/15
s: 0.0044
Epoch 7/15
s: 0.0035
Epoch 8/15
```

```
s: 0.0032
Epoch 9/15
75/75 [=========================== ] - 0s 3ms/step - loss: 0.0064 - val los
s: 0.0035
Epoch 10/15
75/75 [=========== ] - 0s 4ms/step - loss: 0.0063 - val_los
s: 0.0025
Epoch 11/15
75/75 [=============== ] - 0s 3ms/step - loss: 0.0064 - val_los
s: 0.0027
Epoch 12/15
75/75 [=============== ] - 0s 3ms/step - loss: 0.0064 - val_los
s: 0.0030
Epoch 13/15
s: 0.0025
Epoch 14/15
75/75 [========================== ] - 0s 4ms/step - loss: 0.0060 - val los
s: 0.0023
Epoch 15/15
s: 0.0023
94/94 [======] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 0.01
Ridge Regression successful
Best ES smoothing level: 0.3
Exponential Smoothing successful
Ensemble model created successfully
Analyzing high variability scenario...
Processing 13504 samples for high variability
Created 9848 sequences for evaluation
Processing fold 1 of 2
Limiting training sequences from 3284 to 3000
Limiting test sequences from 3282 to 3000
Training with 3000 sequences, testing with 3000 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (3000, 10, 124)
Epoch 1/15
s: 0.0354
Epoch 2/15
s: 0.0170
Epoch 3/15
s: 0.0067
Epoch 4/15
s: 0.0048
Epoch 5/15
75/75 [=============== ] - 0s 4ms/step - loss: 0.0091 - val los
s: 0.0039
Epoch 6/15
```

```
s: 0.0035
Epoch 7/15
75/75 [=================== ] - 0s 3ms/step - loss: 0.0075 - val los
s: 0.0045
Epoch 8/15
s: 0.0029
Epoch 9/15
s: 0.0027
Epoch 10/15
s: 0.0038
Epoch 11/15
s: 0.0022
Epoch 12/15
s: 0.0026
Epoch 13/15
75/75 [============= ] - 0s 4ms/step - loss: 0.0055 - val_los
s: 0.0022
Epoch 14/15
s: 0.0021
Epoch 15/15
s: 0.0022
94/94 [=======] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 0.01
Ridge Regression successful
Best ES smoothing level: 0.1
Exponential Smoothing successful
Ensemble model created successfully
Processing fold 2 of 2
Limiting training sequences from 6566 to 3000
Limiting test sequences from 3282 to 3000
Training with 3000 sequences, testing with 3000 sequences
Selected 20 features for Ridge Regression
LSTM input shape: (3000, 10, 124)
Epoch 1/15
s: 0.0353
Epoch 2/15
s: 0.0178
Epoch 3/15
s: 0.0080
Epoch 4/15
s: 0.0049
Epoch 5/15
s: 0.0038
```

```
Epoch 6/15
s: 0.0034
Epoch 7/15
75/75 [=================== ] - 0s 4ms/step - loss: 0.0077 - val los
s: 0.0030
Epoch 8/15
s: 0.0026
Epoch 9/15
s: 0.0024
Epoch 10/15
s: 0.0025
Epoch 11/15
75/75 [=========================== ] - 0s 3ms/step - loss: 0.0062 - val los
s: 0.0033
Epoch 12/15
s: 0.0026
94/94 [=======] - 0s 1ms/step
LSTM model training and prediction successful
Best Ridge alpha: 0.01
Ridge Regression successful
Best ES smoothing level: 0.1
Exponential Smoothing successful
Ensemble model created successfully
--- DETAILED PERFORMANCE METRICS ---
LOW VARIABILITY SCENARIO:
 LSTM: MAE=0.1356, RMSE=0.1647, R<sup>2</sup>=0.5005, Median AE=0.1241, Explained Var=
 Ridge Regression: MAE=0.1547, RMSE=0.2084, R<sup>2</sup>=0.2602, Median AE=0.1347, Exp
lained Var=0.2850
 Exponential Smoothing: MAE=0.0320, RMSE=0.0847, R^2=0.8757, Median AE=0.007
5, Explained Var=0.8757
 Ensemble: MAE=0.1041, RMSE=0.1255, R<sup>2</sup>=0.7328, Median AE=0.0969, Explained V
ar=0.7353
 Top 5 Features (Ridge Regression):
   time_step_normalized: 0.3126
   171_0_roll_mean_5: 0.1488
   171_0: 0.1408
   171_0_roll_mean_10: 0.1200
   397 24: 0.0967
```

MEDIUM VARIABILITY SCENARIO:

LSTM: MAE=0.0724, RMSE=0.1002, R^2 =0.8486, Median AE=0.0524, Explained Var= 0.8501

Ridge Regression: MAE=0.1211, RMSE=0.5838, R^2 =-8.5355, Median AE=0.0513, Explained Var=-8.3739

Exponential Smoothing: MAE=0.0463, RMSE=0.1117, R^2 =0.8081, Median AE=0.020 1, Explained Var=0.8081

Ensemble: MAE=0.0732, RMSE=0.2517, R²=-0.5450, Median AE=0.0363, Explained Var=-0.5155

Top 5 Features (Ridge Regression):

427_0_roll_mean_20: 0.5150 time_step_normalized: 0.4947 171_0_roll_mean_10: 0.4256

427 0: 0.3714

427_0_roll_mean_10: 0.3183

HIGH VARIABILITY SCENARIO:

LSTM: MAE=0.0508, RMSE=0.0667, R²=0.8946, Median AE=0.0409, Explained Var= 0.8978

Ridge Regression: MAE=0.0248, RMSE=0.0355, R^2 =0.9698, Median AE=0.0181, Exp lained Var=0.9715

Exponential Smoothing: MAE=0.1203, RMSE=0.1607, R^2 =0.3877, Median AE=0.095 1, Explained Var=0.3877

Ensemble: MAE=0.0384, RMSE=0.0502, R^2 =0.9402, Median AE=0.0314, Explained V ar=0.9416

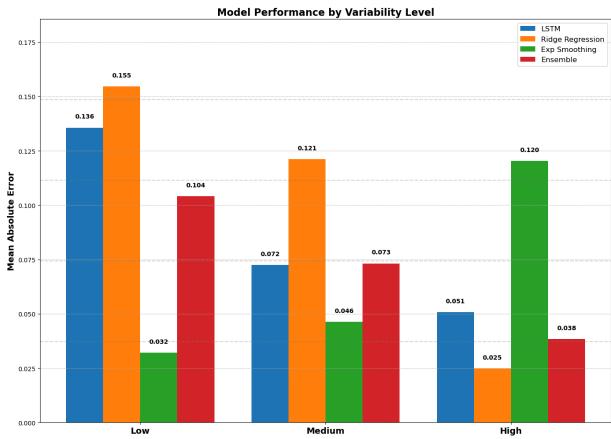
Top 5 Features (Ridge Regression):

171_0: 0.4264

time_step_normalized: 0.4041 427_0_roll_mean_10: 0.3836 171_0_roll_mean_10: 0.3370

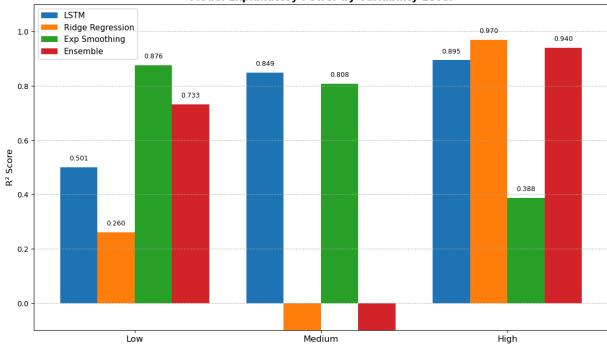
427_0: 0.2840

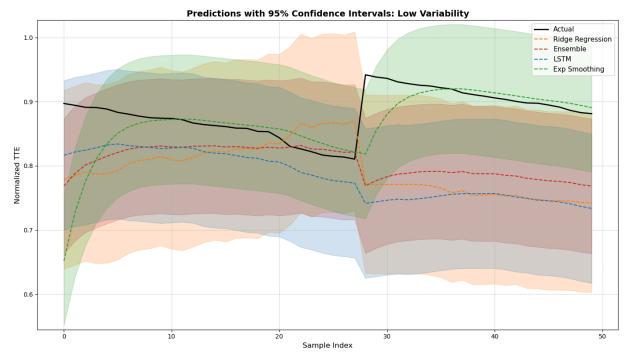
--- CREATING VISUALIZATIONS ---



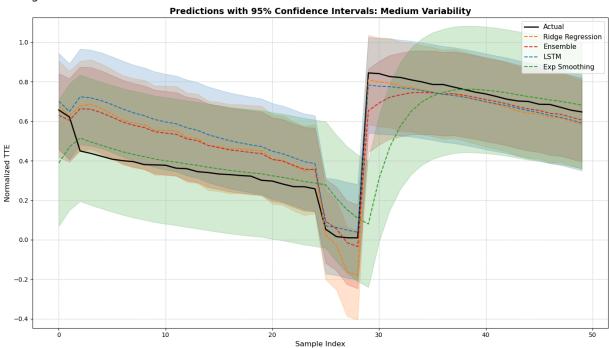
Zoomed MAE comparison plot saved to ./results/mae_comparison_zoomed_20250509_ 182410.png

Model Explanatory Power by Variability Level

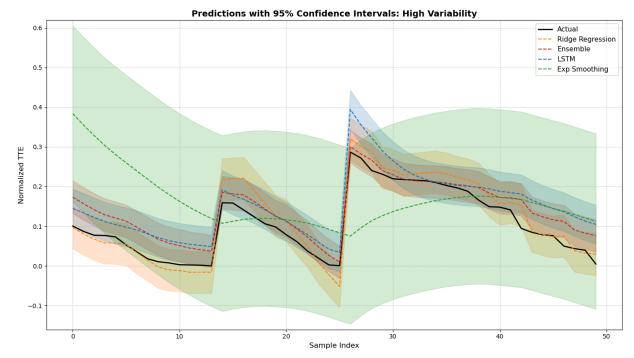




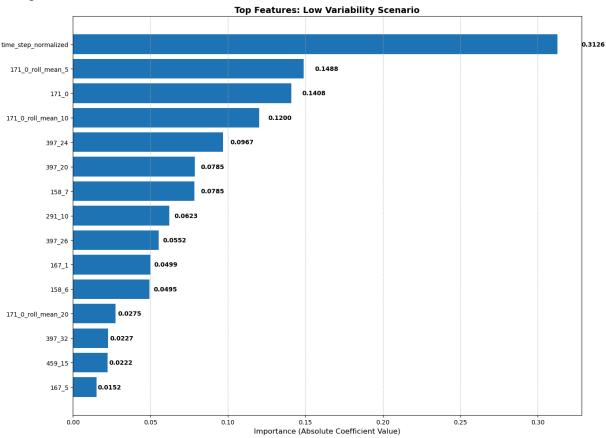
No prediction data available for error plot in low scenario <Figure size 1400x800 with 0 Axes>

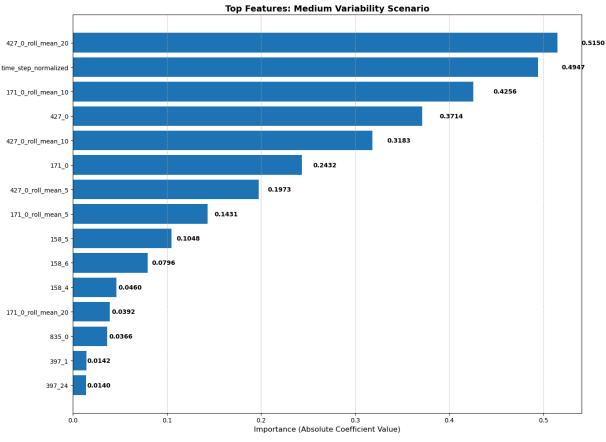


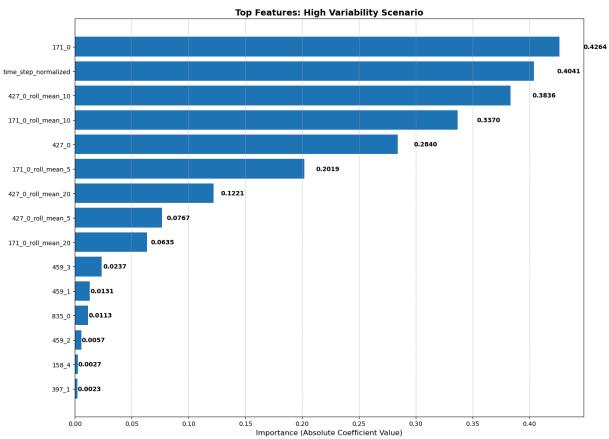
No prediction data available for error plot in medium scenario <Figure size 1400x800 with 0 Axes>



No prediction data available for error plot in high scenario <Figure size 1400x800 with 0 Axes>







```
Performance Summary:
Scenario LSTM MAE LSTM R<sup>2</sup> Ridge MAE Ridge R<sup>2</sup> ES MAE ES R<sup>2</sup> Ensemble
MAE Ensemble R<sup>2</sup>
     Low 0.135634 0.500508
                              0.154691 0.260234 0.032016 0.875681
                                                                         0.104
112
       0.732812
 Medium 0.072427 0.848557
                              0.121147 -8.535544 0.046316 0.808058
                                                                         0.073
    -0.544965
   High 0.050757 0.894553
                              0.024803 0.969821 0.120273 0.387681
                                                                         0.038
422
        0.940210
Performance summary saved to ./results/performance_summary_20250509_182410.cs
--- MODEL SELECTION GUIDELINES ---
For Low Variability Scenario:
  Recommended Model: Exponential Smoothing
 MAE: 0.0320
For Medium Variability Scenario:
  Recommended Model: Exponential Smoothing
 MAE: 0.0463
For High Variability Scenario:
  Recommended Model: Ridge Regression
 MAE: 0.0248
 Top 3 Predictive Features:
   - 171_0: 0.4264
   - time_step_normalized: 0.4041
    - 427_0_roll_mean_10: 0.3836
Saved comparison table to ./metrics 20250509 182410/model comparison report/m
odel comparison table.csv
Saved mae performance plot to ./metrics_20250509_182410/model_comparison_repo
rt/performance_mae.png
Saved rmse performance plot to ./metrics_20250509_182410/model_comparison_rep
ort/performance rmse.png
Saved r2 performance plot to ./metrics_20250509_182410/model_comparison_repor
t/performance r2.png
Saved resource metrics plot to ./metrics_20250509_182410/model_comparison_rep
ort/resource_metrics.png
Saved inference metrics plot to ./metrics 20250509 182410/model comparison re
port/inference metrics.png
Saved efficiency frontier plot to ./metrics_20250509_182410/model_comparison_
report/efficiency_frontier.png
Saved model complexity plot to ./metrics_20250509_182410/model_comparison_rep
ort/model_complexity.png
Saved radar chart to ./metrics_20250509_182410/model_comparison_report/radar_
chart.png
Comprehensive report generated in ./metrics_20250509_182410/model_comparison_
report
posx and posy should be finite values
```

posx and posy should be finite values posx and posy should be finite values

```
Missing required metrics: ['Memory_increase']
```

posx and posy should be finite values

==== RESOURCE USAGE SUMMARY =====

total_execution:

Duration: 202.9792 seconds Memory increase: 761.68 MB Final memory: 1678.70 MB

data_loading_preprocessing: Duration: 2.3104 seconds Memory increase: 50.59 MB

Final memory: 967.61 MB

cleaning_preprocessing:

Duration: 31.4229 seconds Memory increase: 123.50 MB Final memory: 1091.11 MB

feature_engineering_preprocessing:

Duration: 3.8700 seconds Memory increase: 0.12 MB Final memory: 1091.23 MB

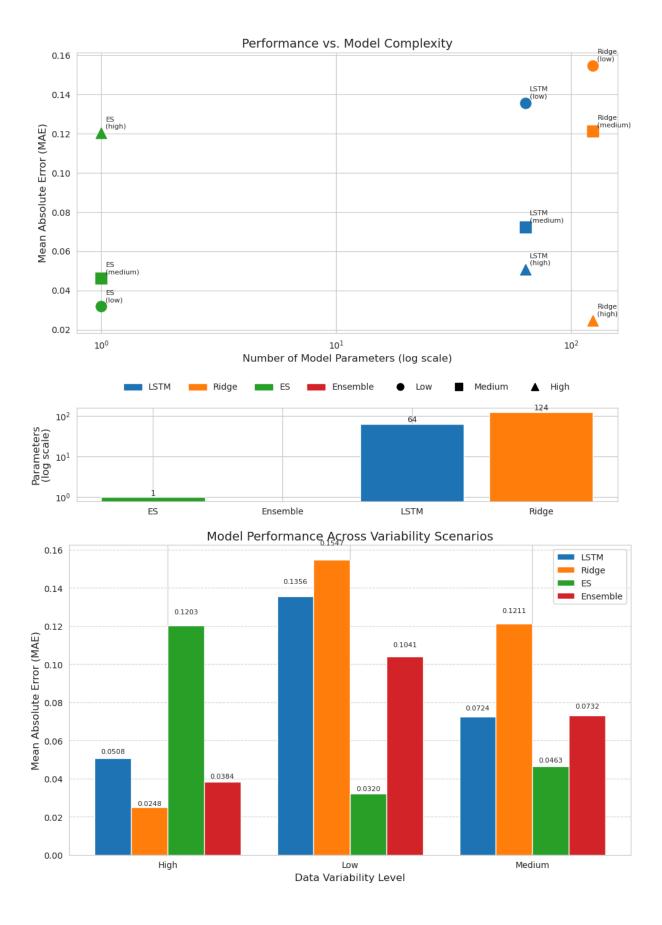
Analysis complete.

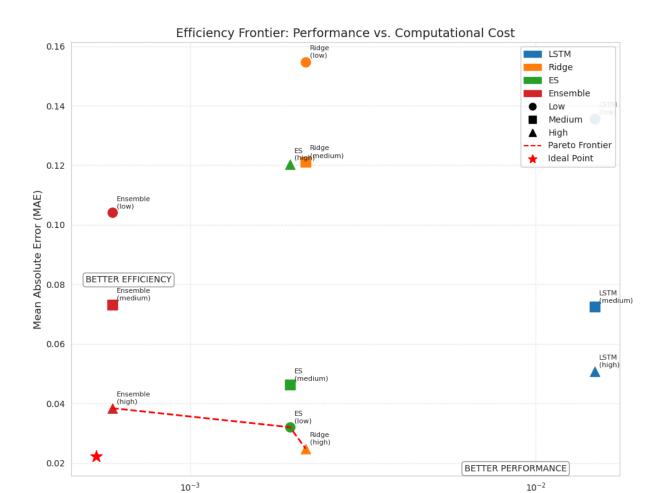
Results saved to: ./results

Metrics saved to: ./metrics_20250509_182410

Paper figures saved to: ./paper_figures_20250509_182410 Paper summary generated: tte_prediction_paper_findings.md

posx and posy should be finite values





Inference Time (seconds, log scale)

Multi-Metric Model Comparison (Medium Variability)

