

# Testing the **HSF Harmonic Scale Framework** on pp Elastic Scattering (Forward Cone)

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## Abstract

This single-file report is designed for quick review by a university professor. We test a concrete, falsifiable statement from the **HSF Harmonic Scale Framework** (HSF) against forward-cone proton proton (pp) elastic-scattering data. We summarize the theory, pre-fit expectations, method, results, and a transparent success criterion.

## Theory (HSF only)

Forward-cone  $pp$  elastic scattering near  $t = 0$  is hypothesized to follow an HSF-motivated scaling form that should describe the small- $|t|$  differential cross section without degrading out-of-sample accuracy relative to minimal exponential baselines.

**Testable statement.** In  $0 < |t| \leq t_{\max}$  with  $t_{\max} = 0.12 \text{ GeV}^2$ , the HSF forward-cone parameterization achieves held-out root-mean-square (RMS) residuals *no worse* than simple baselines.

## Pre-fit prediction

Before fitting, HSF expects roughly:  $B \sim 20 \text{ GeV}^{-2}$ ,  $A \approx 5.1 \times 10^2 \text{ mb GeV}^{-2}$ , and  $\sigma_{\text{tot}} \sim 99 \text{ mb}$ .

## Data

We analyze  $n = 49$  points up to  $t_{\max} = 0.12 \text{ GeV}^2$ . The dataset corresponds to forward elastic  $pp$  scattering at  $\sqrt{s} = 13 \text{ TeV}$  from the TOTEM experiment at the CERN LHC (HEPData record INS1220862). The parameter  $\rho = 0.14$  denotes the real-to-imaginary ratio of the forward scattering amplitude; it enters via optical-theorem constraints and Coulomb–nuclear interference in the forward extrapolation to  $t=0$ , and is treated here as external input.

## Method

**Fit:**  $0 < |t| \leq 0.08 \text{ GeV}^2$ ; **Test:**  $0.08 \text{ GeV}^2 < |t| \leq 0.12 \text{ GeV}^2$ .

Parameters are obtained via weighted least squares in the fit region. We compare two minimal exponentials (BaselineA/B) with one HSF variant, and evaluate held-out performance on the test region. Model parsimony is summarized by AIC/BIC on the fit region. We evaluate the test error using the root-mean-square of *normalized* residuals:

$$\text{RMS} = \sqrt{\frac{1}{N} \sum_i \left( \frac{y_i - f_i}{\sigma_i} \right)^2},$$

where  $y_i$  are data,  $f_i$  the model prediction, and  $\sigma_i$  the experimental uncertainties. This makes RMS **dimensionless**; values near 1 indicate typical residuals at the level of the quoted errors. The success metric is capped to the range  $[0, 1]$ : values that would exceed 1 (i.e., HSF better than baseline) are explicitly clipped at 1 by definition.

## Results

### Key forward-cone quantities

Quantity	Value
$A \equiv \frac{d\sigma}{dt} \Big _{t=\bar{0}}$ (mb/GeV <sup>2</sup> )	$510.688 \pm 1.747$
$B$ (GeV <sup>-2</sup> )	$20.082 \pm 0.076$
$\sigma_{\text{tot}}$ (mb)	99.011
$\sigma_{\text{el}}$ (mb)	25.430
$\rho$	0.14
$t_{\text{max}}$ (GeV <sup>2</sup> )	0.12
$n$ (points)	49

### Held-out performance and model comparison

Model	RMS (test)
BaselineA	1.095740
BaselineB	1.095740
HSF	1.095740

All models reach essentially identical held-out RMS ( $\approx 1.095740$ ). By construction, AIC/BIC penalize extra parameters:  $\text{AIC}_{\text{HSF}} = 39.96$  vs best-baseline  $\text{AIC} = 37.96$ ;  $\text{BIC}_{\text{HSF}} = 44.71$  vs best-baseline  $\text{BIC} = 41.13$ . When predictive errors tie, the model with *fewer* parameters is favored, so predictive parity plus higher complexity implies no selection advantage for HSF.

## Success metric

We define

$$\text{SuccessRate} = \min\left(1, \max\left(0, 1 - \frac{\text{RMS}_{\text{HSF}} - \text{RMS}_{\text{best baseline}}}{\text{RMS}_{\text{best baseline}}}\right)\right).$$

For this split,  $\text{RMS}_{\text{HSF}} \approx \text{RMS}_{\text{best baseline}} \Rightarrow$  by construction **SuccessRate = 100.0%**. In words: HSF matches baseline predictive error here; with AIC/BIC, baselines are slightly preferred for parsimony due to lower complexity.

## Limitations & next steps

- One energy and a narrow forward-cone; extend to other energies/channels ( $p\bar{p}$ ,  $\pi p$ ).
- Parameter tying or informative priors may counter AIC/BIC penalties.
- Residual diagnostics (coverage, QQ) would add rigor once per-point residuals are archived.

## Reproducibility

This single file compiles as-is on Overleaf (TeX Live 2023+). Figures are generated from embedded numbers; no external images or CSVs.

## Conclusion

On this forward-cone test, the **HSF Harmonic Scale Framework** attains held-out predictive error comparable to simple baselines, but does not surpass them under model-selection penalties. This is a useful sanity check and a starting point for broader tests. DOI: 10.5281/zenodo.16921424