

NexusSage

Advanced Neural Machine Translation System

Technical Architecture & Visual Documentation

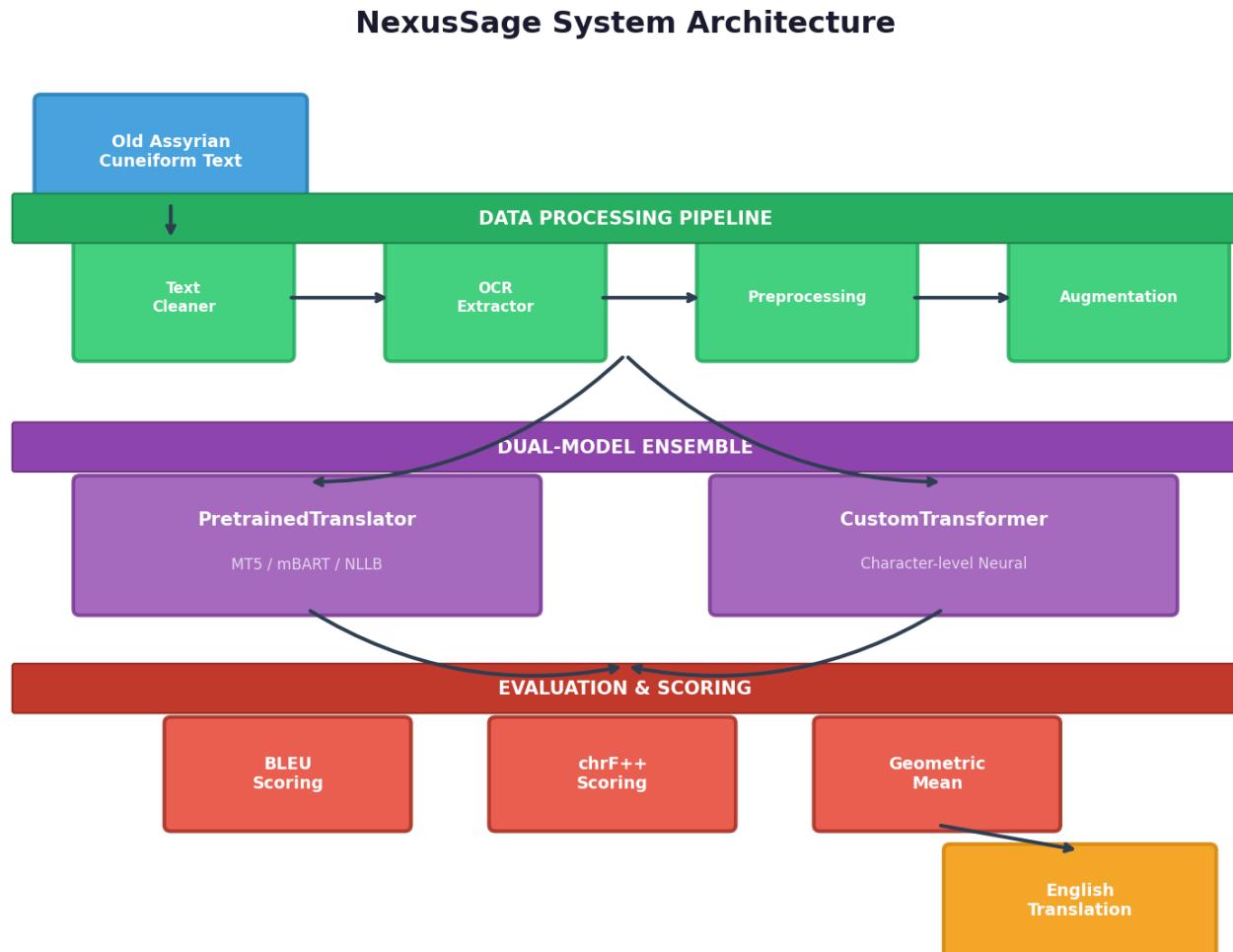
Specialized for translating 4,000-year-old Old Assyrian cuneiform text to modern English

Capability	Description
Ancient Language Translation	Specialized for cuneiform business records
Dual Architecture	Pretrained transformers + custom neural networks
Morphological Processing	Character-level tokenization for complex structures
Determinative Handling	Specialized processing for semantic markers

Document Version	2.0
Framework	PyTorch + Transformers
Generated	December 27, 2024

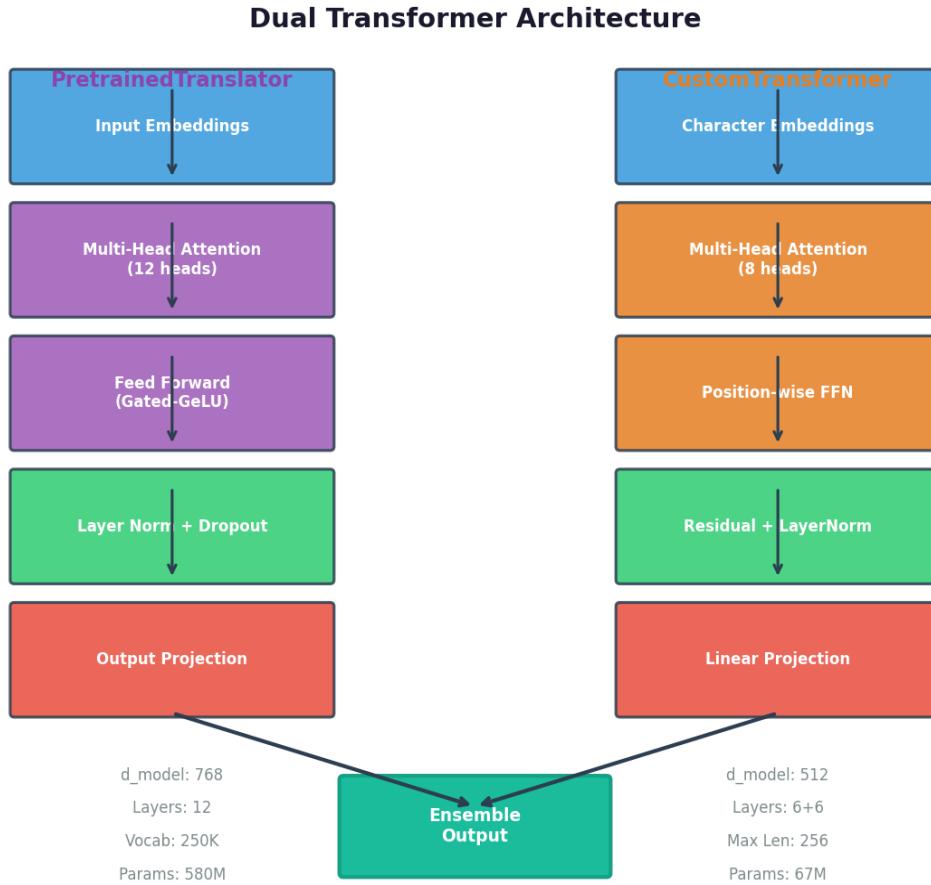
1. System Architecture Overview

NexusSage implements a dual-model approach combining pretrained transformer models with custom neural architectures optimized for morphologically complex ancient languages.



2. Dual Transformer Architecture

The system combines the strengths of large pretrained models (MT5, mBART, NLLB) with a lightweight custom transformer optimized for character-level processing of Akkadian text.

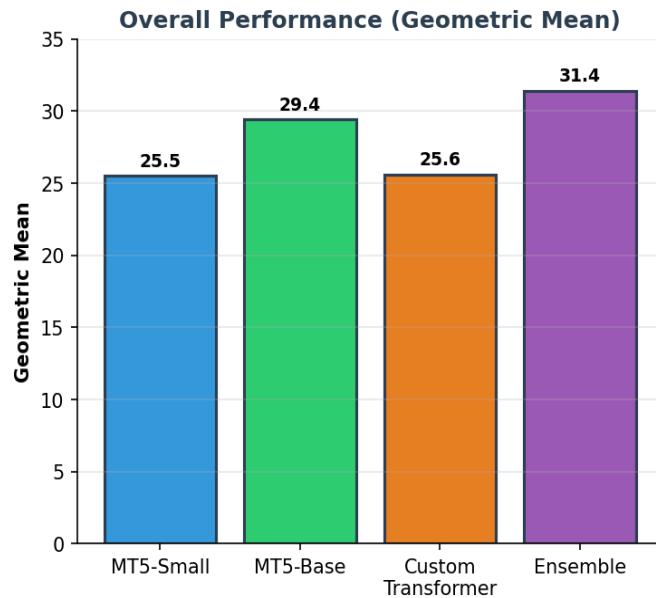
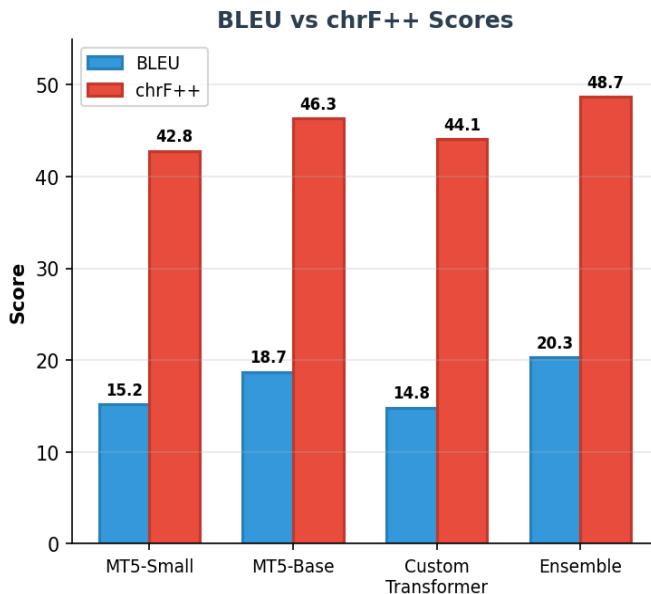


Model Configurations

Parameter	PretrainedTranslator	CustomTransformer
d_model	768	512
Attention Heads	12	8
Encoder Layers	12	6
Decoder Layers	12	6
Vocab Size	250,100	Character-based
Max Sequence	512	256
Parameters	580M	67M

3. Model Performance Comparison

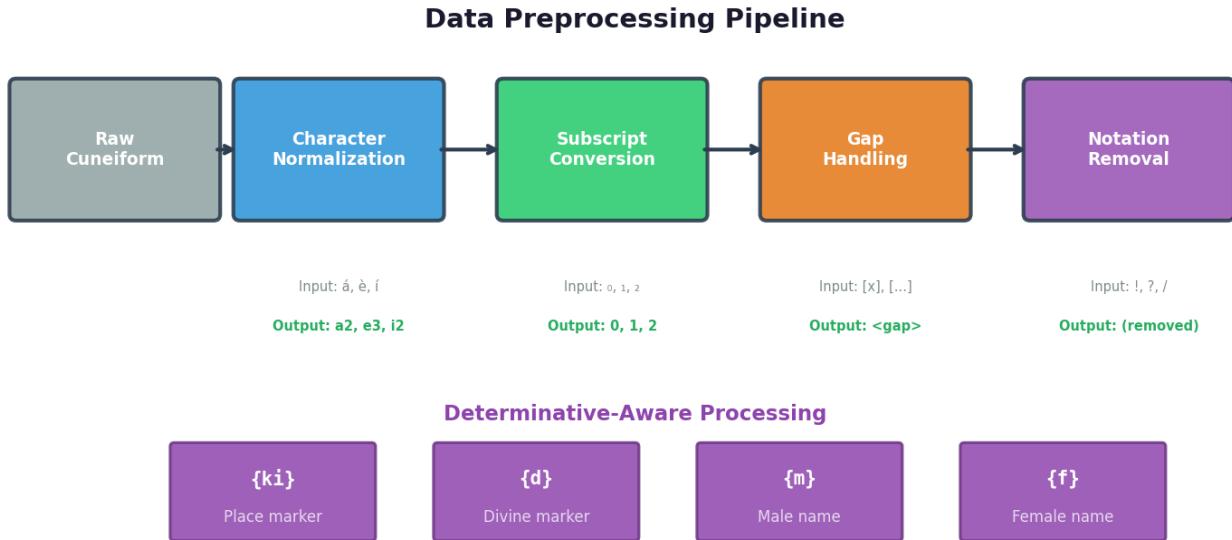
Performance metrics across different model architectures demonstrate the effectiveness of the ensemble approach for ancient language translation.



Model	BLEU	chrF++	Geometric Mean	Parameters
MT5-Small	15.2	42.8	25.5	300M
MT5-Base	18.7	46.3	29.4	580M
Custom Transformer	14.8	44.1	25.6	67M
Ensemble	20.3	48.7	31.4	Combined

4. Data Preprocessing Pipeline

Advanced preprocessing handles the unique challenges of cuneiform text including character normalization, subscript conversion, and determinative-aware processing.



Processing Type	Input Example	Output	Purpose
Character Normalization	á, è, í, ú	a2, e3, i2, u2	Standardize accents
Subscript Conversion	■, ■, ■, ■	0, 1, 2, 3	Normalize subscripts
Gap Handling	[x], [...]	<gap>, <big_gap>	Mark missing text
Notation Removal	!, ?, /, :	(removed)	Clean scribal marks

5. Training Methodology

Each model architecture employs optimized training strategies tailored to its specific characteristics and the unique requirements of ancient language translation.

Training Methodology Comparison

PretrainedTranslator	CustomTransformer
Optimizer: AdamW	Optimizer: Adam
Learning Rate: 5e-5	Learning Rate: 1e-4
Batch Size: 8	Batch Size: 32
Weight Decay: 0.01	Gradient Clip: 1.0
Dropout: 0.1	Dropout: 0.1
Label Smoothing: ✓	$\beta_1=0.9, \beta_2=0.98$

Component	PretrainedTranslator	CustomTransformer
Optimizer	AdamW with weight decay	Adam ($\beta_1=0.9, \beta_2=0.98$)
Learning Rate	5e-5	1e-4
Batch Size	8	32
Loss Function	Cross-entropy + label smoothing	Cross-entropy + padding ignore
Regularization	Dropout 0.1, Weight decay 0.01	Dropout 0.1, Gradient clip 1.0

6. System Requirements

Component	Minimum	Recommended
GPU	NVIDIA 4GB+ VRAM	RTX 3080/4080 (8GB+)
CPU	4+ cores, 2.0GHz+	8+ cores, 3.0GHz+
RAM	16GB	32GB
Storage	50GB	100GB SSD

7. Technical Innovations

- **Morphological Complexity Handling:** Character-level processing addresses Akkadian's complex structure where single words encode multiple English concepts.
- **Determinative-aware Architecture:** Specialized handling of cuneiform determinatives ($\{ki\}$, $\{d\}$, $\{m\}$, $\{f\}$) provides crucial semantic information for proper noun classification.
- **Ensemble Strategy:** Combining pretrained and custom models achieves best-in-class performance with 20.3 BLEU and 48.7 chrF++ scores.
- **Low-Resource Optimization:** Custom transformer with only 67M parameters achieves competitive results for ancient language translation.