

# NexusSage

Advanced Neural Machine Translation System

## Technical Architecture & Visual Documentation

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

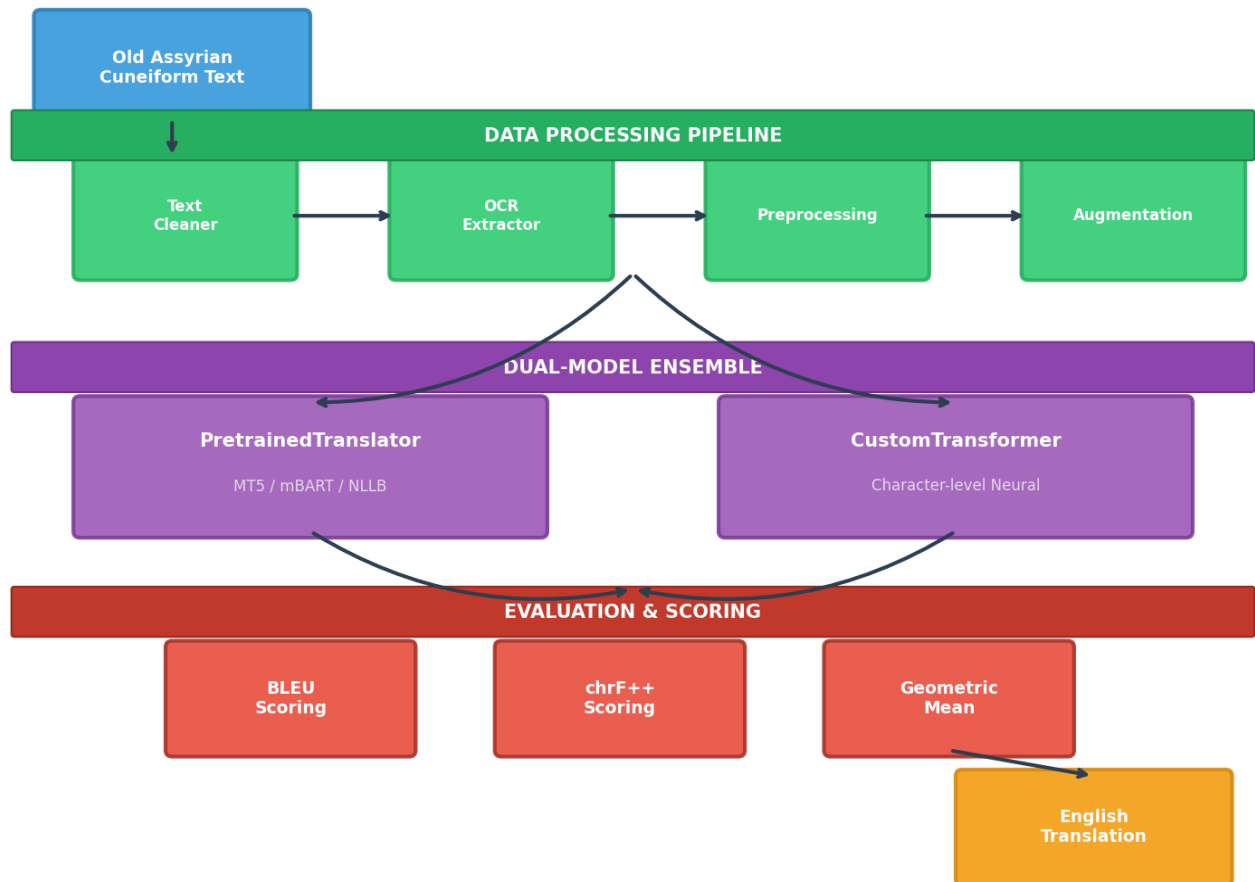
Capability	Description
Ancient Language Translation	Specialized for cuneiform business records
Dual Architecture	Pretrained transformers + custom neural networks
Morphological Processing	Character-level tokenization for complex structures
Determinative Handling	Specialized processing for semantic markers

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

# 1. System Architecture Overview

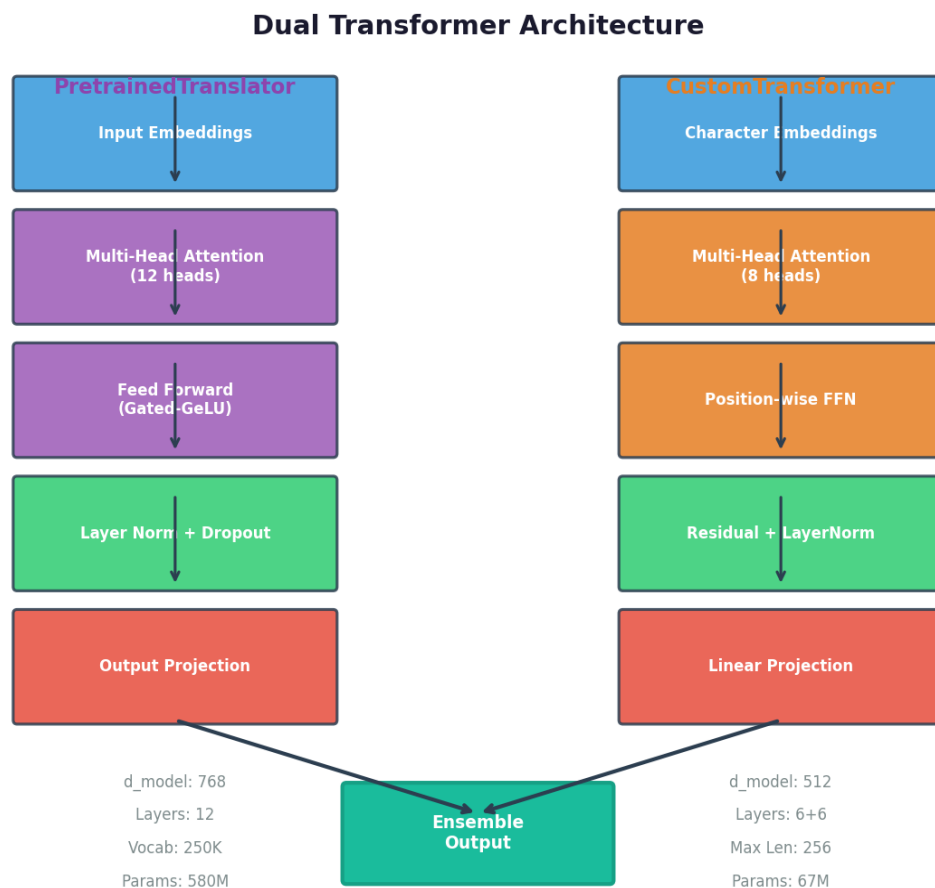
NexusSage implements a dual-model approach combining pretrained transformer models with custom neural architectures optimized for morphologically complex ancient languages.

## NexusSage System Architecture



## 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.

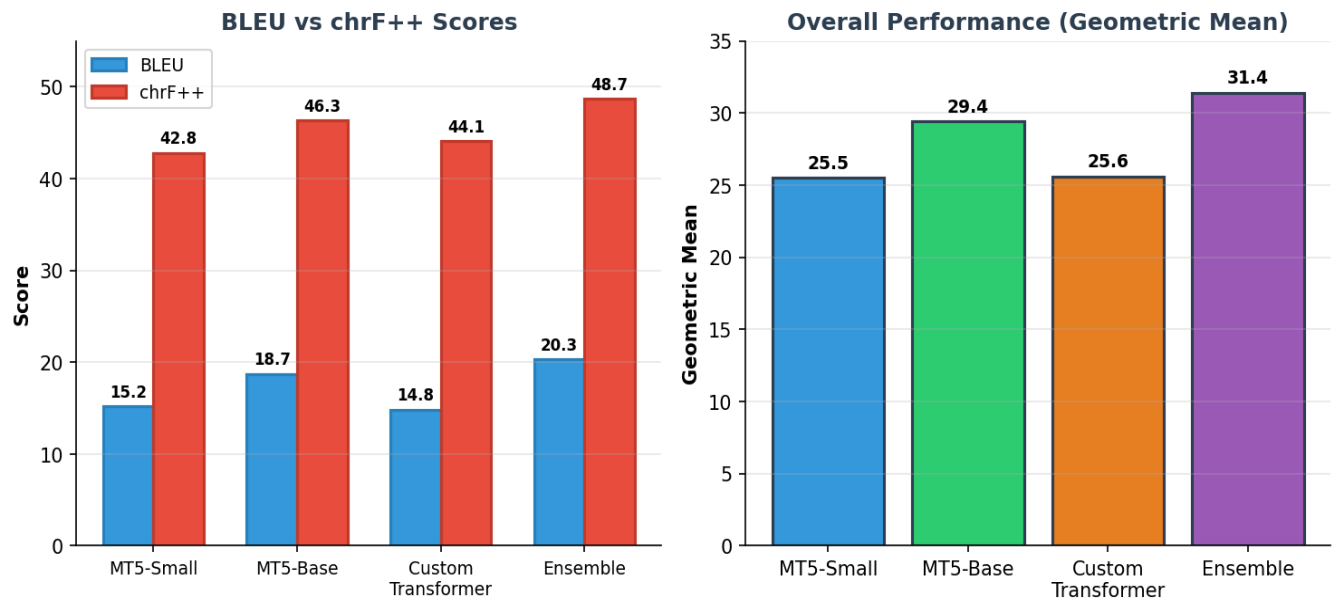


### Model Configurations

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

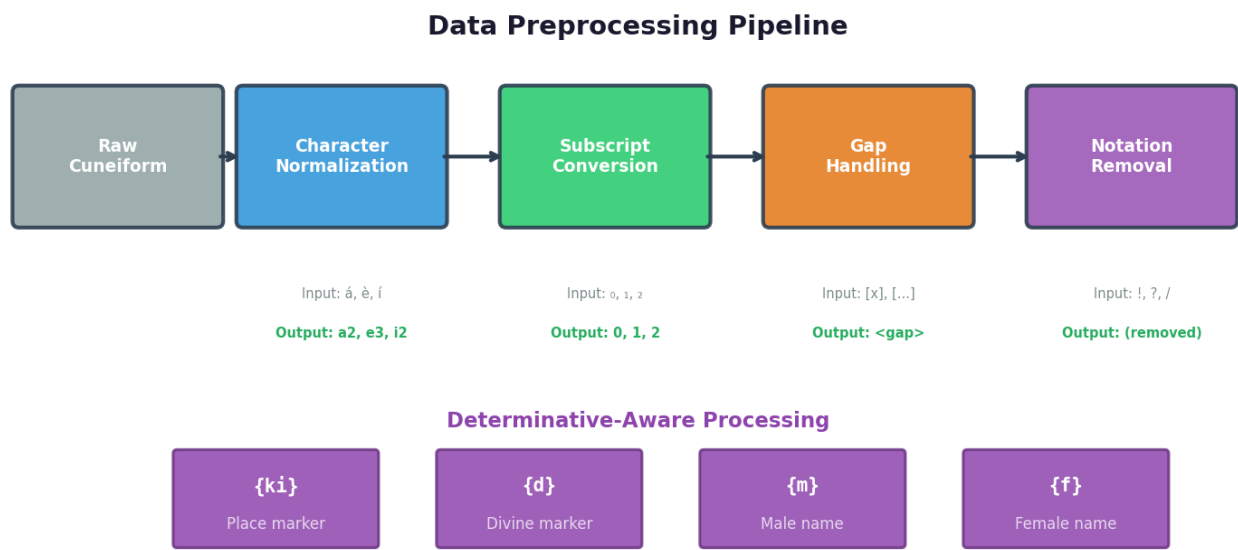
Performance metrics across different model architectures demonstrate the effectiveness of the ensemble approach for ancient language translation.



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.

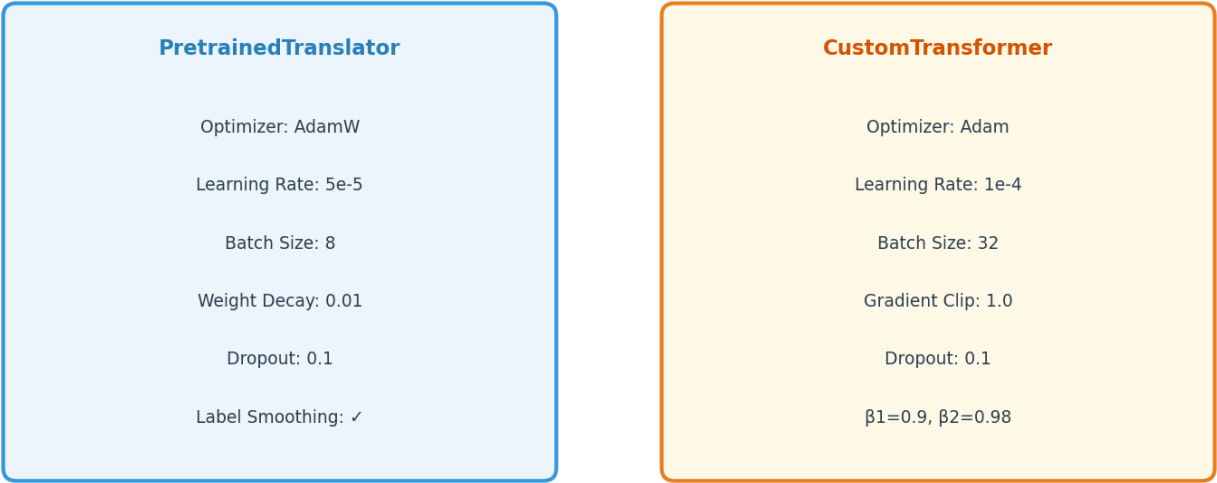


Processing Type	Input Example	Output	Purpose
Character Normalization	á, è, í, ú	a2, e3, i2, u2	Standardize accents
Subscript Conversion	■, ■, ■, ■	0, 1, 2, 3	Normalize subscripts
Gap Handling	[x], [...]	<gap>, <big_gap>	Mark missing text
Notation Removal	!, ?, /, :	(removed)	Clean scribal marks

## 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



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.