

## Overview of Time Series Foundation Models (TSFM)

Model	Open / Closed	Architecture	Pretraining dataset(s)	# Params	Output	Input
<b>TimesFM (v1.0 / v2.0)</b>	Open (Apache-2.0) ( <a href="#">GitHub</a> )	Decoder-only Transformer with patching ( <a href="#">Google Research</a> )	~100B real-world points incl. Google Trends & Wikipedia + synthetic ( <a href="#">Google Research</a> )	200M (v1.0), 500M (v2.0) ( <a href="#">GitHub</a> )	Point forecast; optional <b>experiment</b> al 10-quantile heads (uncalibrated) ( <a href="#">GitHub</a> )	Univariate target; optional past/future covariates via external regressors (zero-shot) ( <a href="#">GitHub</a> )
<b>Chronos</b>	Open (Apache-2.0) ( <a href="#">Hugging Face</a> )	Encoder-decoder <b>T5</b> with value tokenization & de/quantization; autoregressive sampling ( <a href="#">Hugging Face</a> )	Large corpus of public TS + synthetic GP data ( <a href="#">Hugging Face</a> )	~8M / 20M / 46M / 200M / 710M (tiny → large) ( <a href="#">Hugging Face</a> )	Probabilistic (sampled trajectories) ( <a href="#">Hugging Face</a> )	Univariate target (single series per forecast) ( <a href="#">Hugging Face</a> )
<b>Chronos-X</b>	Open (code released)	Lightweight adapter modules: <b>IIB</b> (Input Injection Block) + <b>OIB</b> (Output	Inherits Chronos pretraining; adds covariate benchmark	Uses Chronos backbones (8M–710M); adapter overhead	Probabilistic (inherits Chronos' sampled forecasts; OIB adjusts logits)	<b>Univariate target + covariates</b> (past & future)

		Injection Block) added to pretrained model; supports past & future covariates	synthetic datasets) and tests on 18 real-world datasets	extra params] ( <a href="#">Hugging Face</a> )		
<b>Chronos-2</b>	Open (Apache-2.0) ( <a href="#">Hugging Face</a> )	Encoder-only Transformer (T5-encoder-inspired) with patch embedding s + group attention for in-context learning across series/covariates	Subset of Chronos datasets + GIFT-Eval pretrain + large-scale synthetic multivariate /covariate data	~120M (base size)	Multi-step <b>quantile</b> forecasts (probabilistic)	Uni- or multivariate targets; supports <b>past-only</b> and <b>known-future</b> covariates (real & categorical)
<b>Moirai</b>	Open (Apache-2.0; code & weights) ( <a href="#">GitHub</a> )	Masked <b>encoder</b> Transformer ; “any-variate attention” (handles variable # of variables) ( <a href="#">arXiv</a> )	<b>LOTSA:</b> 27B observations across 9 domains ( <a href="#">arXiv</a> )	14M / 91M / 311M (small/base/large)	Probabilistic (mixture distribution head) ( <a href="#">arXiv</a> )	Any-variate (uni/multi); supports covariates ( <a href="#">arXiv</a> )

<b>Moirai-2</b>	Open (cc-by-nc-4.0) ( <a href="#">Hugging Face</a> )	<b>Decoder-only</b> Transformer ; multi-token generation; quantile loss; patch embedding s with missing-value flag; data filtering for non-forecastable series	GIFT-Eval (pretrain/train) + Chronos mixup (non-leaking subsets) + KernelSynth synthetic + internal Salesforce ops data	~12M params (estimated)	Probabilistic <b>quantile</b> forecasts (multi-step)	supports dynamic covariates via Uni2TS interface; more sizes likely forthcoming ( <a href="#">HF collection</a> )
<b>Lag-Llama</b>	Open (Apache-2.0) ( <a href="#">Hugging Face</a> )	Small decoder-only Transformer ; uses <b>lags as covariates</b>	Diverse public TS corpora (paper)	<b>2.45M</b> ( <a href="#">Hugging Face</a> )	Probabilistic (distribution per step) ( <a href="#">Hugging Face</a> )	Univariate (lags; no exogenous by default)
<b>Tiny Time Mixers (TTM)</b>	Open (research weights; Granite variants Apache) ( <a href="#">Hugging Face</a> )	<b>TSMixer/M</b> <b>LP-Mixer</b> family with Granite variants (Monash + LibCity), I tweaks (AP/DRS/RP T); multi-level channel mixing	Public datasets (starts at ≈1M) ~1B samples (pretrain)	<b>~1M+</b>	Point forecasts (paper focuses on accuracy/speed; no native probabilistic head reported)	<b>Multivariate</b> ; can incorporate <b>exogenous</b> signals during fine-tuning (channel mixing)
<b>TimeGPT-1</b>	<b>Closed</b> (hosted)	Transformer encoder-	<b>&gt;100B</b> data points	NA	Probabilistic via	Uni- or multivariate;

	API) ( <a href="#">Nixtla</a> )	decoder specialized for TS	across many domains (public sources)	<b>conformal</b> prediction intervals; quantiles available via API ( <a href="#">docs.nixtla.io</a> , <a href="#">GitHub</a> )	supports <b>exogenous variables</b> (past & future) via API ( <a href="#">Nixtlaverse</a> )
--	---------------------------------	----------------------------	--------------------------------------	--	--

## Notes:

- **TimesFM**: GitHub README (license, v1/v2 sizes, covariates & quantile heads) and Google Research blog (architecture, 100B pretraining mix). ([GitHub](#), [Google Research](#))
- **Chronos**: HF model card (license, sizes, data mix, sampling, univariate). ([Hugging Face](#))
- **Chronos-X**: AISTATS 2025 paper (adapters, covariates, benchmarks) + Chronos card for backbone sizes. ([Hugging Face](#))
- **Moirai**: ICML 2024 paper + Uni2TS GitHub (LOTSA size, any-variate, params; Apache-2.0). ([arXiv](#), [GitHub](#))
- **Lag-Llama**: HF model card (params, license [? double check], probabilistic output) + paper context. ([Hugging Face](#))
- **Tiny Time Mixers**: arXiv (public data, ~1B samples, exogenous/channel mixing; starts at ~1M params); HF Granite/IBM cards for licensing context. ([Hugging Face](#))
- **TimeGPT-1**: arXiv (architecture, >100B points; conformal intervals; param size unknown) + Nixtla docs (closed API, intervals, exogenous support). ([Nixtla](#), [docs.nixtla.io](#), [Nixtlaverse](#))

## Shortlist for macro + GDELT-style covariates + probabilistic output:

### 1. Moirai (Salesforce)

- **Strengths**: Handles uni/multivariate targets and dynamic covariates; features a **mixture-of-distributions** head for full predictive densities. ([arXiv](#))
- **Scale & data**: Pretrained on **LOTSA** (~27B observations, 9 domains); available in **Small/Base/Large (14M/91M/311M)** sizes for flexibility. ([Salesforce](#), [Hugging Face](#))
- **Notes**: Weights are released under **CC-BY-NC-4.0** (non-commercial); code in Uni2TS. ([Hugging Face](#), [GitHub](#))
- **Notes**: uni2ts can be added to environment to ease use of model

## 2. Chronos (Amazon)

- T5-family **encoder–decoder** LM; values are **tokenized**; forecasts are **probabilistic** via sampling multiple futures. ([Hugging Face](#), [GitHub](#))
- **Caveat: No native covariates**; designed for **univariate targets** only. Participants can add covariate effects through external models, residual forecasting, etc. ([Hugging Face](#))

## 3. Chronos-2 (Amazon)

- Natively ingests covariates and multivariate signals.
- Outputs quantiles directly (no sampling loop).
- Much simpler to deploy in a classical forecasting pipeline (X, y, horizons).

## 4. Moirai-2 (Salesforce)

- Decoder-only TS FM with direct multi-step **\*\*quantile\*\*** outputs, patch embeddings (with missing-value flags), and stronger data hygiene (filters non-forecastable series).
- Trains on GIFT-Eval + curated Chronos-style mixes + KernelSynth synthetic; released under **\*\*CC-BY-NC-4.0\*\*** with a small checkpoint prioritized for efficiency
- Good fit for lightweight deployment with native uncertainty