Aligning Multilingual News for Stock Return Prediction

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Introduction & Contribution

We propose a method to align sentences in multilingual news articles using optimal transport, identifying semantically similar content across languages. Aligned sentences are sparser, more interpretable, and exhibit higher semantic similarity. Return scores constructed from aligned sentences show stronger correlations with realized stock returns, and long-short trading strategies based on these alignments achieve higher Sharpe ratios than analyzing the full text sample.

Our work:

- Analyzes full news content rather than aggregate sentiment scores
- Employs sentence-level alignment to capture semantic similarity
- Produces sparser, more interpretable cross-lingual alignments
- Applies the method to financial text analysis

Methodology

From Bloomberg, we collect an average of **17,800 stock-days** per year with both **English and Japanese** news for **3,500 Tokyo Stock Exchange stocks**.

Goal: Align English and Japanese articles at the **sentence level** for each stock-day, capturing **semantically similar content** while minimizing misalignments.

Optimal Transport (OT) maps probability mass between two distributions while minimizing transport cost. Kantorovich formulation (discrete):

$$\min_{\gamma} \left\{ \sum_{ij} c_{ij} \gamma_{ij} : \gamma_{ij} \ge 0, \sum_{i} \gamma_{ij} = p_{y_j}, \sum_{j} \gamma_{ij} = p_{x_i} \right\}$$

- x_i : English sentence embedding with empirical distribution p_{x_i}
- y_i : Japanese sentence embedding with empirical distribution p_{y_i}
- c_{ij} : cosine distance between x_i and y_i , scaled to [0,1]

After solving for γ (J \rightarrow E) and γ' (E \rightarrow J), compute a 1-1 sentence mapping:

$$\mathcal{A} = A * B^T * (\xi_{ij} > \xi_{thres})$$

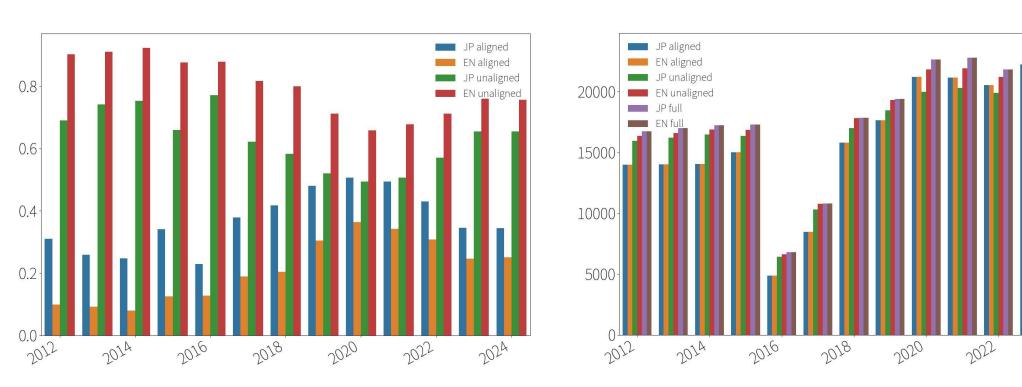
- *A*: J→E alignment, row max and top 5% by column
- B: E→J alignment, row max and top 5% by column
- ξ_{ij} : cosine similarity between x_i and y_i

Construct group-level embeddings by averaging sentence embeddings

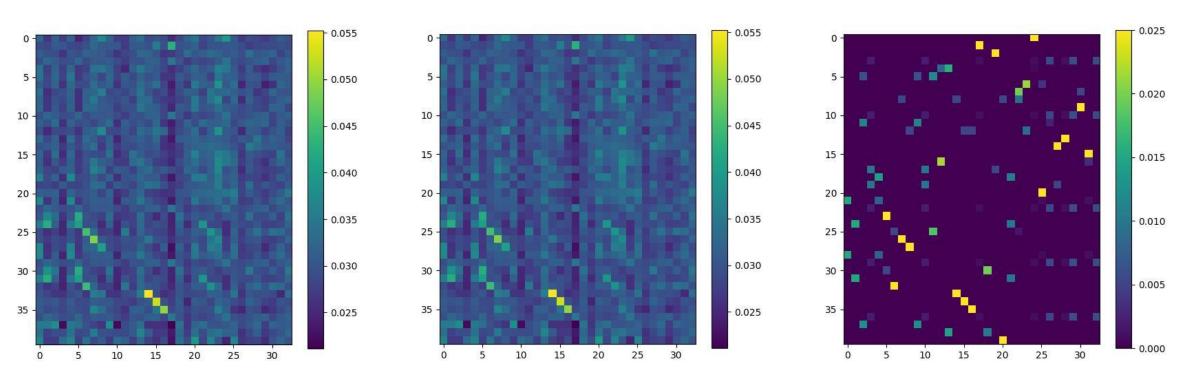
- Unaligned: average over sentences with A = 0
- Aligned: average over sentences with A=1
- Full: average over all sentences in the article

Alignment Results

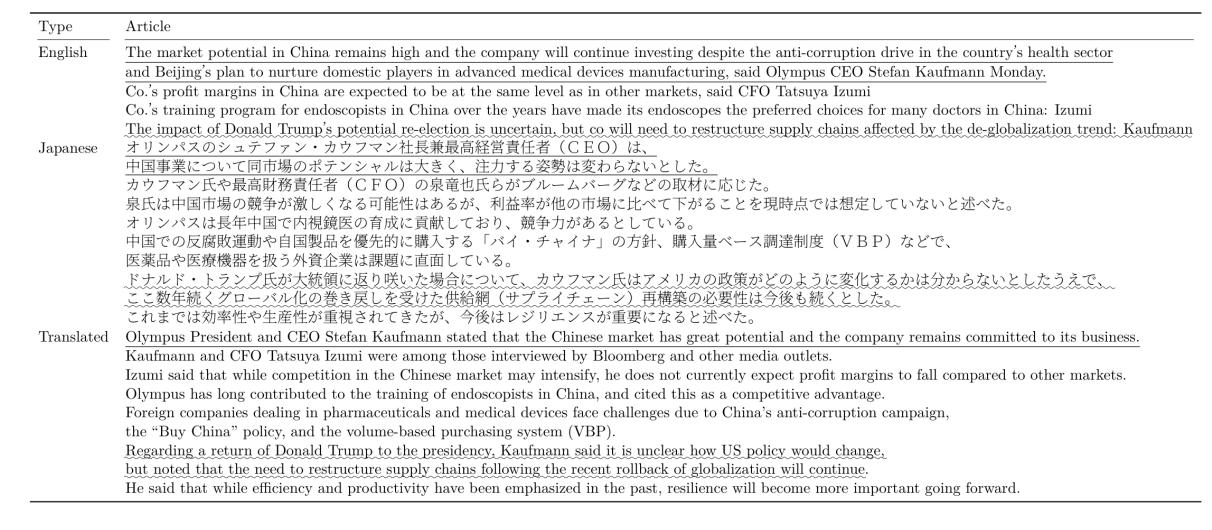
- Most stock-day articles contain a mix of aligned and unaligned sentences
- In recent years, about 40–60% of sentences are mutually aligned across languages
- Produces 1–1 sentence alignments based on semantic similarity



Proportion of aligned sentences and article counts with aligned/unaligned sentences



Sparsity of alignments computed by Softmax (left), Entmax (middle), Optimal Transport (right) on news articles discussing the Bank of Japan (8301.T) on 2023-01-04. Sample alignments for this article can be found in the paper.



Sentence alignments between English and Japanese news for Olympus (7733.T) on 2024-07-23. Translation of the Japanese news is provided by Google Translate.

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Trading Strategy Performance

Apply Ridge regression to link stock returns with text embeddings, generating return scores $Soft^{l,k}$:

$$\arg \min_{w} ||X_{c,t}^{l,k}w - \text{Ret}_{c,t}^{OC}||^2 + \lambda ||w||^2$$

- $X_{c,t}^{l,k}$: embedding in language l and alignment type k, using articles published between 8:30 am day t-1 and 8:30 am day t for each firm c
- Ret^{OC}_{c,t}: open-to-close return from Compustat Global
- λ : regularization parameter chosen by cross-validation
- Rolling window: 6-year training, 1-year out-of-sample prediction
- Evaluation Period: 2018-2024
- Trading strategy: long top 25% and short bottom 25% of stocks by return score (days with \geq 20 traded stocks), **equally weighted**

	Ret	$\mathrm{Soft}^{EN,A}$	$\mathrm{Soft}^{EN,UA}$	Soft^{EN}	Soft JP,A	$\mathrm{Soft}^{JP,UA}$	Soft^{JP}
Ret	1.00	0.02	0.03	0.03	0.02	0.01	0.03
$\mathrm{Soft}^{EN,A}$	0.02	1.00	0.41	0.67	0.67	0.38	0.57
$Soft^{EN,UA}$	0.03	0.41	1.00	0.76	0.49	0.47	0.50
$Soft^{EN}$	0.03	0.67	0.76	1.00	0.62	0.41	0.62
$\mathrm{Soft}^{JP,A}$	0.02	0.67	0.49	0.62	1.00	0.44	0.72
Soft JP,UA	0.01	0.38	0.47	0.41	0.44	1.00	0.71
Soft JP	0.03	0.57	0.50	0.62	0.72	0.71	1.00

Return Score Correlations

Alignment	Lang	Geo Mean	Mean	Std	5%	50%	95%	Sharpe	Ann. Sharpe
Aligned	JP	0.35%	0.36%	1.32%	-1.76%	0.39%	2.31%	0.27	4.36
	EN	0.28%	0.29%	1.34%	-1.86%	0.23%	2.46%	0.22	3.42
Unaligned	JP	0.17%	0.18%	1.34%	-1.98%	0.16%	2.29%	0.13	2.12
	EN	0.23%	0.24%	1.29%	-1.66%	0.16%	2.45%	0.18	2.91
Full	JP	0.30%	0.31%	1.23%	-1.59%	0.27%	2.38%	0.25	3.98
	EN	0.24%	0.25%	1.16%	-1.65%	0.17%	2.20%	0.21	3.40

Strategy Summary

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