

# How Increased Chinese Exports Drive Media Slant in U.S. Local Newspapers?

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## Introduction

- Despite the benefits of globalization, the increased imports from China have resulted in various **adverse effects**:
  - Surging U.S. manufacturing unemployment (Autor et al., 2013; Acemoglu et al., 2016)
  - Deterioration in public health (McManus and Schaur, 2016; Pierce and Scott, 2016)
  - “China-bashing” in election campaign strategies (Ramirez and Rong, 2012)
- Measurement of **media slant** in existing studies relies on “negative keyword detection”, which is biased due to subjectivity and ambiguity.

## Research Questions

- How increased Chinese exports drive media slant in U.S. local newspapers?
- How to perform sentiment analysis on news via deep learning for NLP?
- What is the main topics of negative reports about China?

## Data

- 114,788 pieces of articles from 157 local newspapers: *Newslibrary*
- Newspaper circulation covering 52 states: *Alliance of Audited Media*
- 4-digit Harmonized System international trade: *U.N. Comtrade*
- State-level industrial and demographic structures: *U.S. Census Bureau*

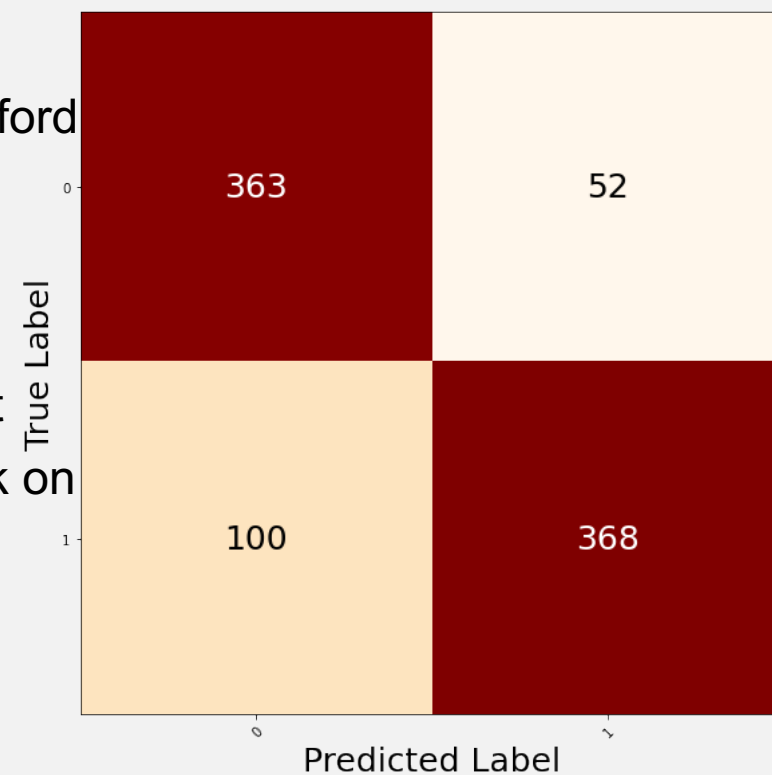
## Sentiment Analysis using LSTM

To identify negative China-related reports, LSTM is adopted. The learning algorithm is as follows:

- Choose movie reviews dataset from Stanford AI Lab as training set
- Tokenize and vectorize the texts by “nltk”, and remove English stop words
- Split the dataset into training & testing set
- Import “keras” to construct neural network on the Tensorflow framework

The trained model reaches **82.79%** accuracy in the testing set.

Then apply the model to selected newspaper articles.

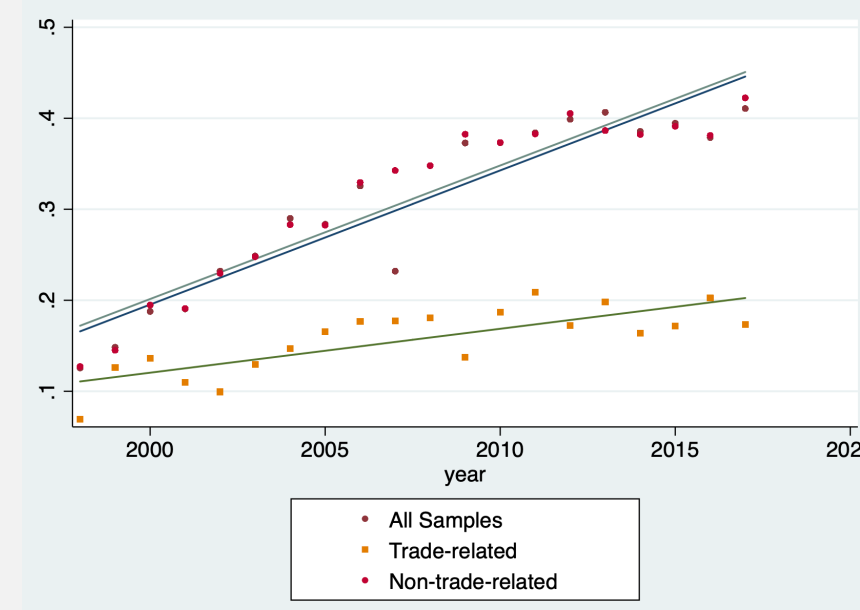


## Key Variables

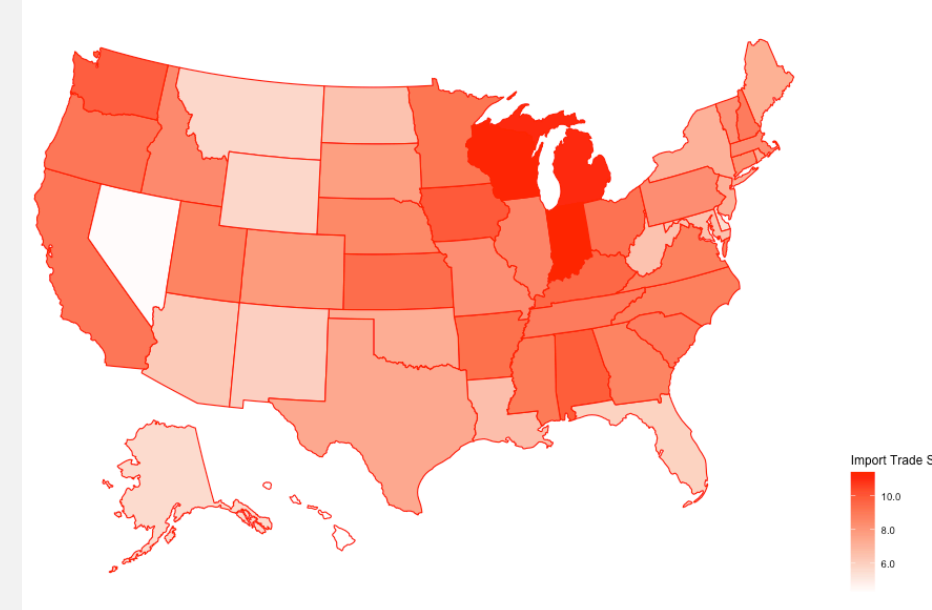
### Media Slant

$$NegRatio_{i,t} = \frac{\# Negative_{i,t}^j}{\# China_{i,t}}$$

where  $j$  denotes all, trade-related and non-trade-related reporting.



Annual Average Media Slant in All and Subgroups, 1998-2017



U.S. Exposure to Chinese Imports across States, 1998-2018  
Calculated according to Autor et al. (2013). Darker color indicates greater increase in Chinese import competition.

### Imported Trade Shocks from China

$$\Delta Import_i^{China} = \sum_s \frac{Circulation_{i,s}^{1998}}{Circulation_i^{1998}} \sum_s \frac{L_{s,j}^{1998}}{L_s^{1998}} \frac{\Delta M_j^{China}}{L_j^{1998}}$$

where  $i$  denotes newspaper,  $j$  denotes industry and  $s$  denotes state;

$Circulation$  is weekly circulation,  $L$  is the employment and  $\Delta M_j^{China}$  denotes is the change in Chinese imports over 1998 to 2017.

## Model Specification

$$\Delta NegRatio_{i,t} = \alpha + \beta \Delta Imports_{i,t}^{China} + \gamma X_{i,1998} + \Delta \varepsilon_{i,t}$$

- $\Delta NegRatio_{i,t} \equiv NegRatio_{i,t} - NegRatio_{i,1998}$  captures the change in media slant against China by newspaper  $i$  from 1998 to year  $i$ ;
- $\Delta Imports_{i,t}^{China}$  denotes the newspaper-level change in Chinese imports;
- $X_{i,1998}$  is a host of circulation-weighted shares of readership attributes: female population, Asian population, population with a bachelor’s degree and median income level.

## Empirical Results

### Baseline

	(1)	(2)	(3)	(4)	(5)	(6)
$\Delta NegRatio$	All Sample	All Sample	Trade	Trade	Non-trade	Non-trade
$\Delta Imports$	0.0283*** (17.80)	0.0281*** (17.28)	0.00999*** (6.13)	0.00974*** (5.80)	0.0286*** (17.89)	0.0284*** (17.32)
Controls	No	Yes	No	Yes	No	Yes
Constant	0.157*** (16.38)	0.119* (2.10)	0.104*** (11.60)	0.00173 (0.23)	0.161*** (16.68)	0.116* (2.03)
$N$	3140	3140	3140	3140	3140	3140

### Robustness Checks with TextBlob Sentiment Analysis

	(1)	(2)	(3)	(4)	(5)	(6)
$\Delta NegRatio$	All Sample	All Sample	Trade	Trade	Non-trade	Non-trade
$\Delta Imports$	0.0457*** (22.18)	0.0466*** (22.21)	0.0178*** (7.62)	0.0173*** (7.19)	0.0464*** (22.42)	0.0474*** (22.48)
Controls	No	Yes	No	Yes	No	Yes
Constant	0.242*** (18.32)	0.304** (3.09)	0.177*** (13.38)	0.00442 (0.43)	0.244*** (18.50)	0.302** (3.07)
$N$	3140	3140	3140	3140	3140	3140

“TextBlob” is a python package for sentiment analysis that uses simple dictionary-based algorithm.

## Conclusions

- Newspapers whose circulation states face greater exposure to Chinese import shocks report more negative news about China.
- The source of negative descriptions more stems from non-trade-related topics rather than trade-related ones.
- The increase of female and Asian population shares restrain the rise of negative trade-related coverage (not shown above).

## Contact Me

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