

# Nowcasting waterborne commerce: a Bayesian model averaging approach

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# Motivation

## Nowcasting

- Use real-time data to predict contemporaneous values of a variable released with lag

## Why do we care?

- Timely and accurate data helps market participants with operational decisions
- Network supports 600m tons of cargo per year

## This paper

- Nowcast waterborne commerce (WBC) with coincident indicators
- Bayesian model averaging (BMA) averages nowcasts across models with different sets of predictors

# Waterborne commerce (WBC)



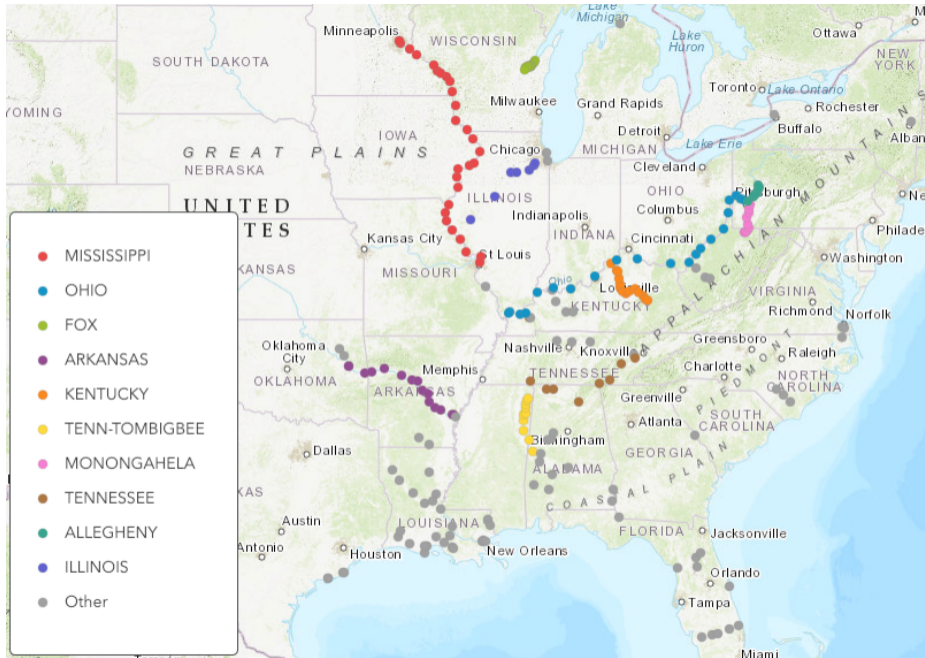


A lock is an elevator for ships

- Helps navigate uneven and inconsistent water levels

Lock Performance Monitoring System (LPMS)

- Measures cargo for each ship inside lock



# Data

## WBC and LPMS example

Load 1 10 tons through locks L1, L2, L3

Load 2 20 tons through locks L2, L3, L4

Load 3 50 tons through locks L3, L4

$$WBC = 10 + 20 + 50 = \mathbf{80}$$

	L1	L2	L3	L4
Load 1	10	10	10	
Load 2		20	20	20
Load 3			50	50
LPMS	<b>10</b>	<b>30</b>	<b>80</b>	<b>70</b>



## Data (2000-2013)

Monthly data for total tonnage and four primary commodities

- Coal
- Petroleum
- Food & farm
- Chemical

Waterborne commerce (WBC) data

- Point to point tonnages moved along the waterway
- Released with more than a year lag

Lock Performance Monitoring System (LPMS) data

- Real-time tonnages passing through particular locks

# Conceptual framework

# Bayesian model averaging (BMA)

Posterior probability that  $M_j$  is the true model

$$Pr(M_j|Y) = \frac{f(Y|M_j) Pr(M_j)}{\sum_{i=1}^J f(Y|M_i) Pr(M_i)} \quad j = 1, \dots, J$$

- $Y$  observed data
- $Pr(M_j)$  prior probability that  $M_j$  is the true model
- $f(Y|M_j)$  marginal likelihood for model  $M_j$

$$f(Y|M_j) = \int f(Y|\theta_j, M_j) p(\theta_j|M_j) d\theta_j$$

# Bayesian model averaging (BMA)

BMA averages posterior inference across alternative models

$$\widehat{WBC}_t = \sum_{j=1}^J \widehat{WBC}_t^j \Pr(M_j|Y)$$

Posterior inclusion probability (PIP)

$$PIP_n = \sum_{j=1}^J \Pr(M_j|Y) I_j(X_n)$$

Empirically

- Markov-chain Monte Carlo Model Composition (MC<sup>3</sup>)
- Rolling window techniques

# Results

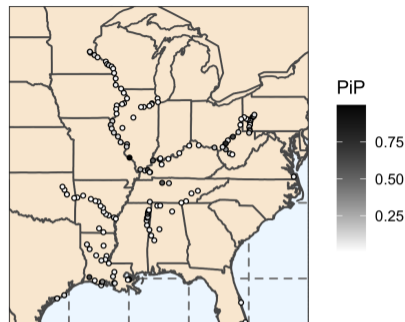
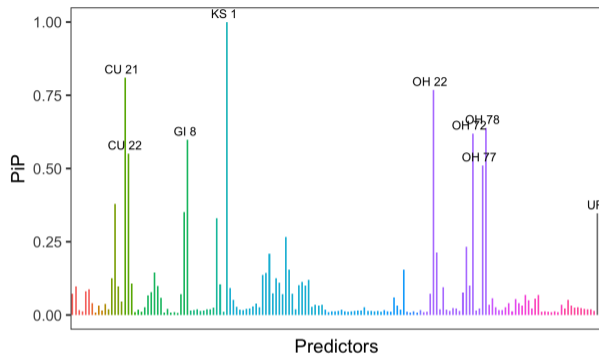
## Top models

Posterior Model Probabilities for Top 10 Models:  $\Pr(M_j|Y)$

	Total	Coal	Farm	Petro	Chem
1	1.47	1.72	1.31	1.61	1.57
2	1.42	1.28	1.20	1.49	1.28
3	1.12	1.17	1.17	1.23	1.17
4	1.11	0.96	1.15	1.09	1.01
5	0.95	0.95	0.99	1.05	0.98
6	0.82	0.82	0.93	0.96	0.94
7	0.81	0.83	0.92	0.73	0.86
8	0.80	0.80	0.86	0.65	0.82
9	0.77	0.70	0.75	0.63	0.69
10	0.75	0.67	0.72	0.58	0.68

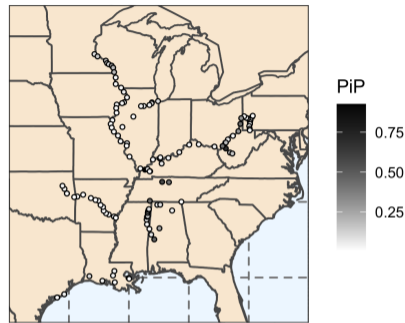
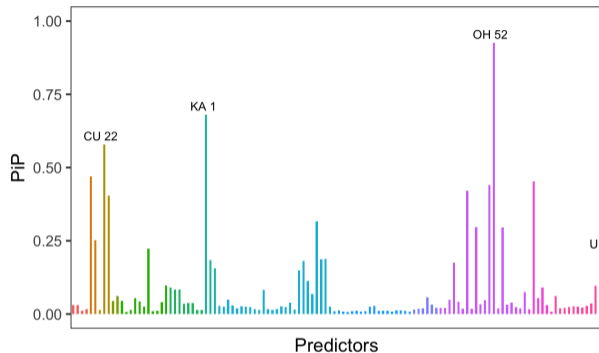
*Note:* All table entries should be multiplied by  $10^{-7}$

# Total



Predictor	PiP	River	State
Kaskaskia River Navigation Lock	0.999	Kaskaskia	IL
Barkley Lock	0.809	Cumberland	KY
Racine Lock and Dam	0.767	Ohio	OH
Smithland Lock and Dam	0.638	Ohio	KY
Willow Island Locks and Dam	0.618	Ohio	OH

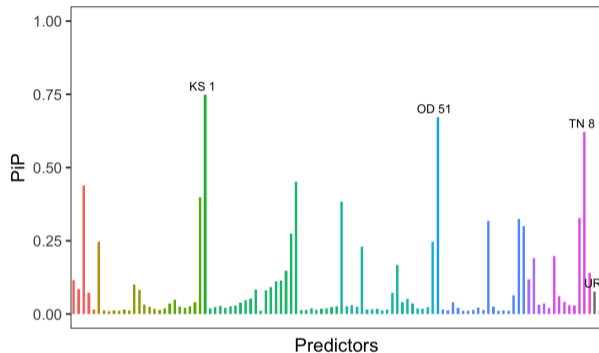
# Coal



Predictor	PiP	River	State
Lock and Dam 52	0.926	Ohio	IL
Winfield Locks and Dam Main 1	0.680	Kanawha	WV
Cheatham Lock	0.578	Cumberland	TN

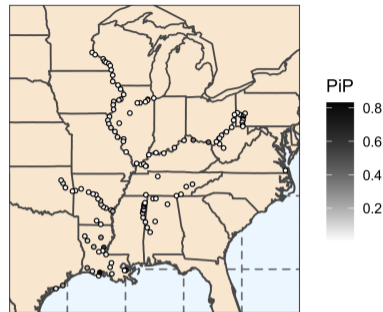
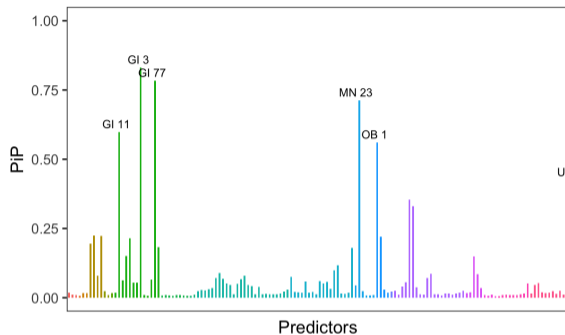


# Food & farm



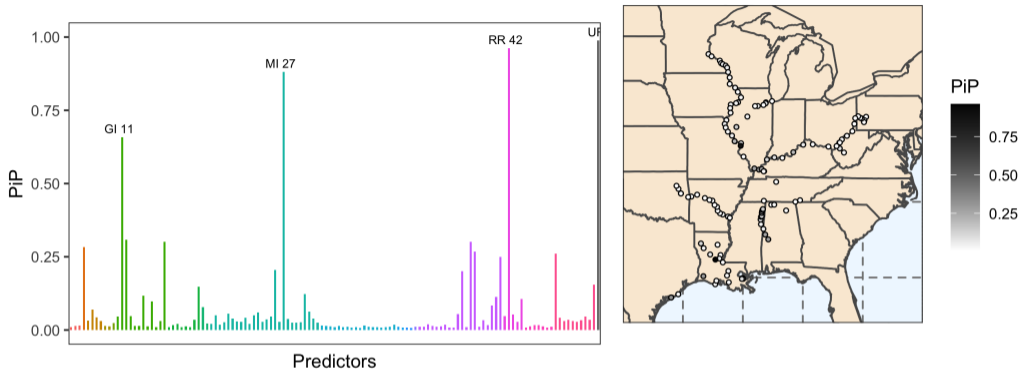
Predictor	PiP	River	State
Kaskaskia River Navigation Lock	0.748	Kaskaskia	IL
Old River Lock	0.672	Old	LA
Watts Bar Lock	0.622	Tennessee	TN

# Petroleum



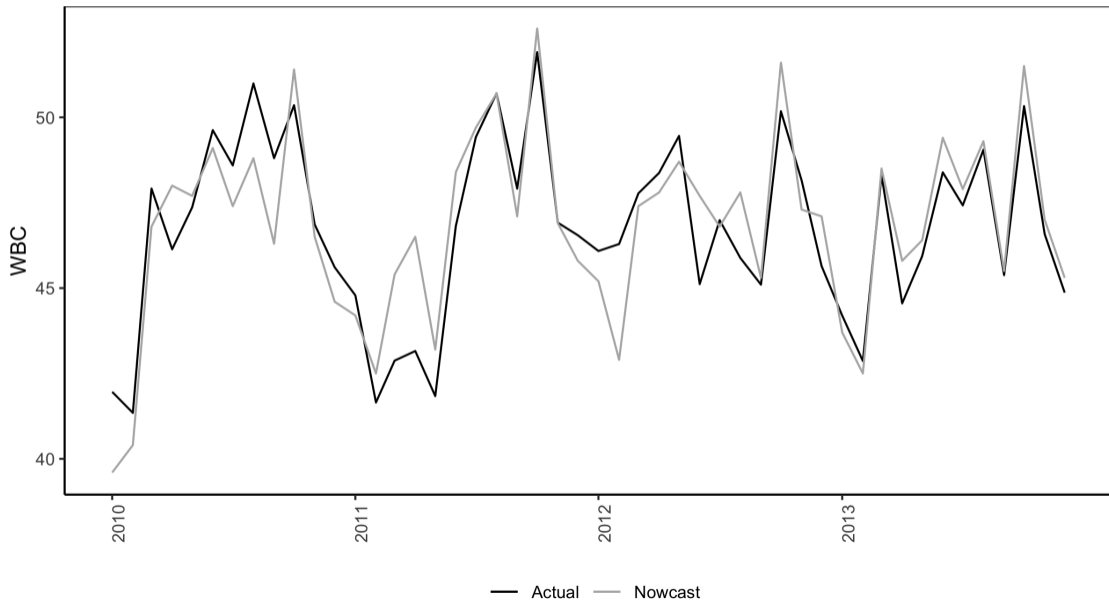
Predictor	PiP	River	State
Inner Harbor Navigation Canal Lock	0.831	Gulf	LA
Leland Bowman Lock	0.783	Gulf	LA
Lock and Dam 3	0.712	Monongahela	PA
Colorado River East Lock	0.598	Gulf	TX
Jonesville Lock and Dam	0.560	Ouachita	LA

# Chemical

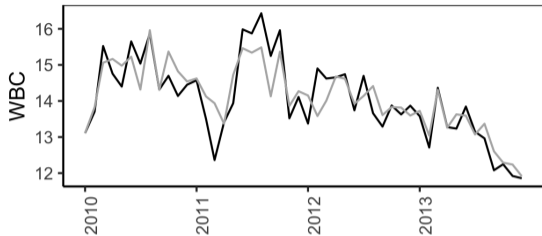


Predictor	PiP	River	State
Unemployment rate (two-month lag)	0.988		
John H. Overton	0.961	Red	LA
Chain of Rocks Lock and Dam 27	0.881	Mississippi	MO
Colorado River East Lock	0.658	Gulf	TX

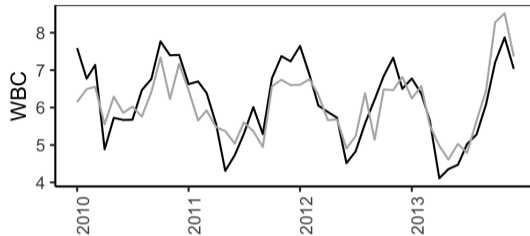
# Total



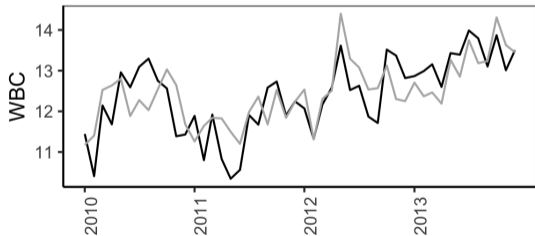
### Coal



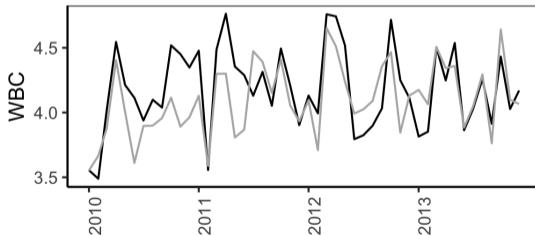
### Food & farm



### Petroleum



### Chemical



— Actual — Nowcast

## Nowcast evaluation metrics

### Average Percentage Nowcast Error

Year	Total	Coal	Farm	Petroleum	Chemical
2010	1.98	-0.65	3.23	-0.94	4.75
2011	-2.31	-0.27	2.29	-2.13	2.95
2012	-0.34	0.28	1.08	-1.44	1.02
2013	-0.96	-1.23	-5.69	1.45	-1.27

# Summary

# Summary

Bayesian model averaging (BMA) incorporates model uncertainty into nowcasts

- Average nowcasts across models with different sets of predictors

Results can be operationalized

- Help firms make decisions
- Negotiate rates
- Gauge equipment needs



Thank you