

# Calendar Effects and Omitted Variables in Employment Time Series

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Seasonal Adjustment Practitioners Workshop  
November 4, 2016



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*Any opinions expressed in this presentation are those of the author and do not constitute policy of the Bureau of Labor Statistics.*

# Overview

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- Very brief explanation of Current Employment Statistics (CES) program
  - ▶ Focus on subnational data
- Four vs. Five week calendar effect
- Omitted variables and the problems they bring
- Screening for problems with alternate runs

# Current Employment Statistics

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- Large monthly survey (>600,000 establishments)
  - ▶ Fed-State cooperative
- Employment, hours, and earnings at National, State, and Area (MSA) level
- Some of the most timely economic indicators
  - ▶ Most interest: employment change
- Benchmarked to admin. data

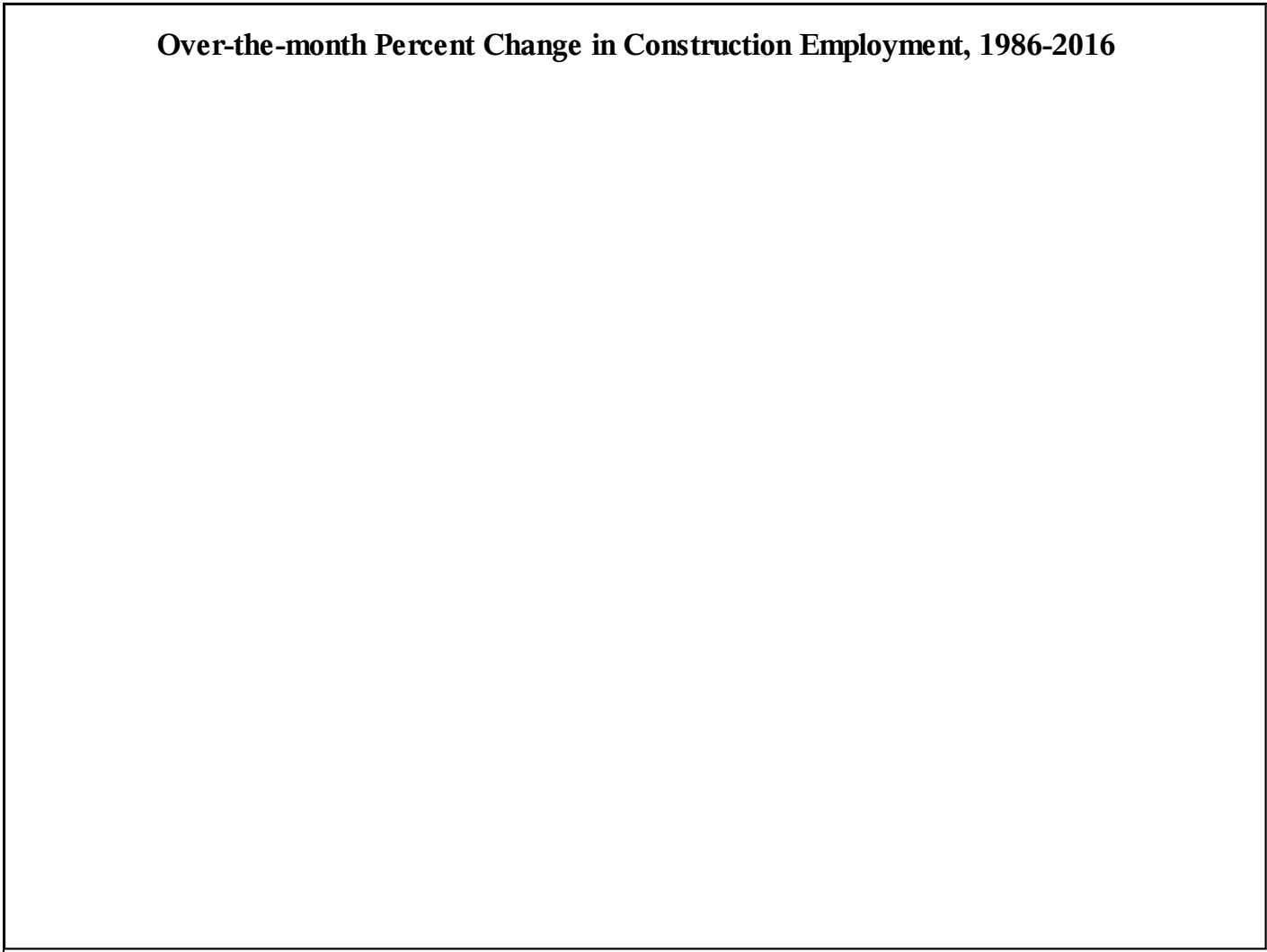
# Current Employment Statistics

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- State/MSA-level seasonal adjustment
  - ▶ “Two-Step” due to benchmark technique
  - ▶ Projected factors
    - Move to concurrent proposed for 2018
  - ▶ Publish 2024 SA series (not incl. 3MMA)
- CES reference period: “Pay period including the 12<sup>th</sup> of the month”
  - ▶ Time between reference weeks is variable
    - Noticeable in highly seasonal months
  - ▶ User-defined “4/5 Week Effect”

# Four-Five Week Effect

Over-the-month Percent Change in Construction Employment, 1986-2016



# Four-Five Week Effect

Date	Weeks	dum1	dum2	dum3	dum4	dum5	dum6	dum7	dum8	dum9	dum10	dum11
JAN2016	4	<b>1.0</b>	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
FEB2016	5	0.0	<b>-0.6</b>	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
MAR2016	5	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
APR2016	4	0.0	0.0	<b>1.0</b>	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
MAY2016	5	0.0	0.0	0.0	<b>-0.6</b>	0.0	0.0	0.0	0.0	0.0	0.0	0.0
JUN2016	4	0.0	0.0	0.0	0.0	<b>1.0</b>	0.0	0.0	0.0	0.0	0.0	0.0
JUL2016	5	0.0	0.0	0.0	0.0	0.0	<b>-0.6</b>	0.0	0.0	0.0	0.0	0.0
AUG2016	5	0.0	0.0	0.0	0.0	0.0	0.0	<b>-0.6</b>	0.0	0.0	0.0	0.0
SEP2016	4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	<b>1.0</b>	0.0	0.0	0.0
OCT2016	5	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	<b>-0.6</b>	0.0	0.0
NOV2016	5	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	<b>-0.6</b>	0.0
DEC2016	4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	<b>1.0</b>

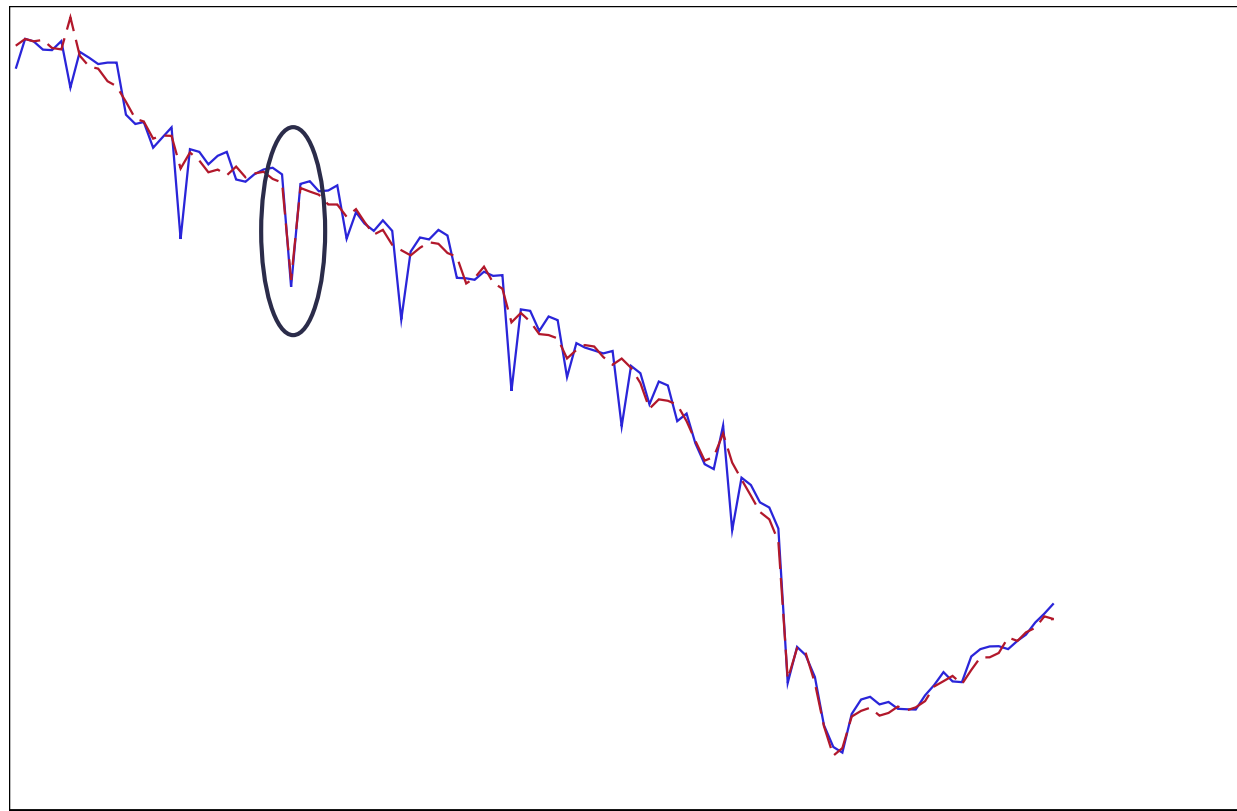
# Four-Five Week Effect

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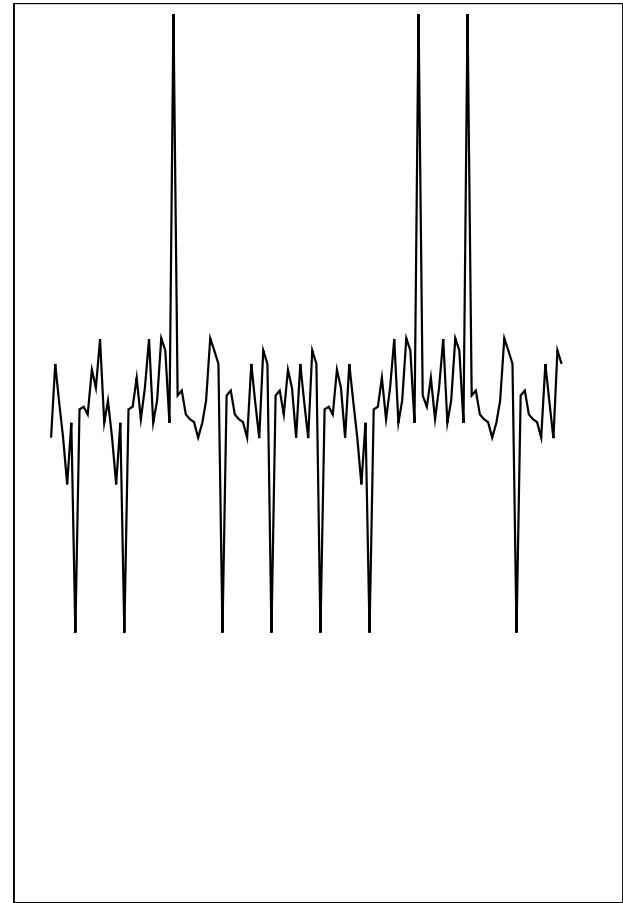
- Important to control for fluctuations in the calendar
- Implemented w/ X-12-ARIMA in May 1996 (Cano et al.)
- Ten-year spans (standard CES input) will have few four (or five) week observation for each month
  - ▶ Potential for over-fitting the data



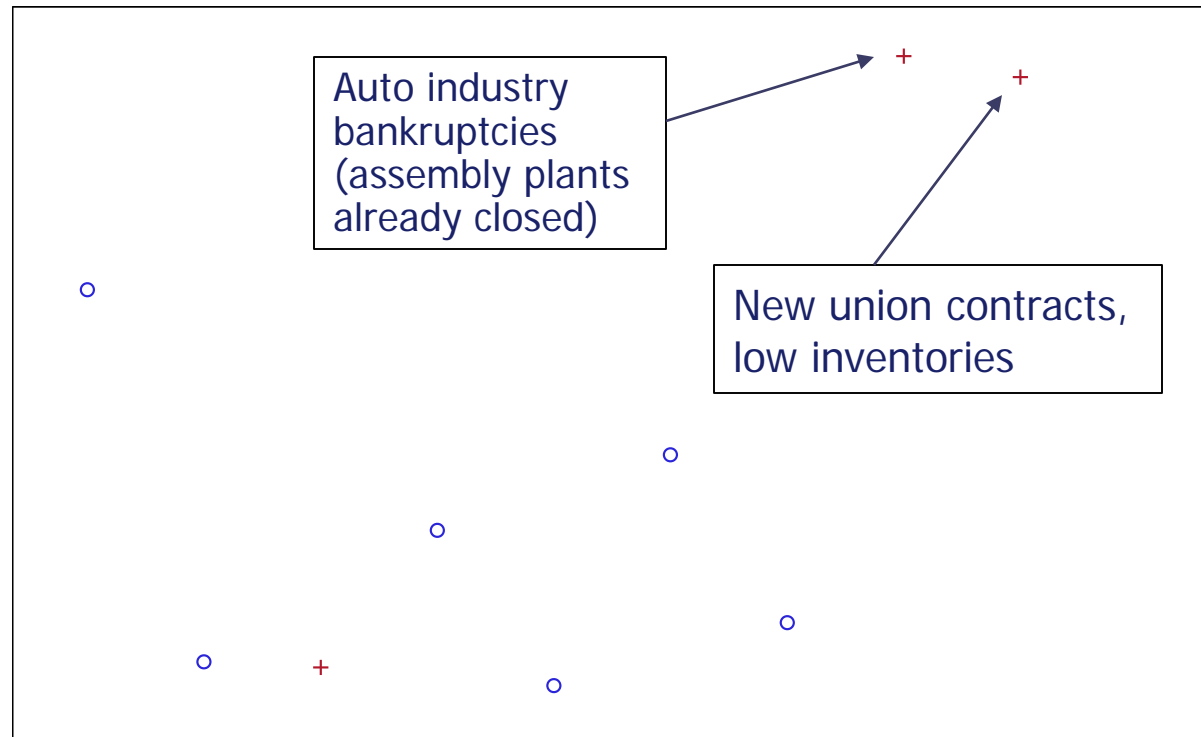
# Ex. Michigan Durable Goods (Historical Problem)



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# Historical Problems

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- Similar issues found in some other series w/ heavy auto-industry concentration
  - ▶ Midwest state-level durable goods mfg.
  - ▶ Some metro area total nonfarm
- Search for other problem series
  - ▶ Decennial Census, hurricanes, etc.
  - ▶ Hard to know what's a problem if not SME
- Bad projected factors: mid-year changes

# Omitted Variables

*Model 1 (“Short regression”):*

$$Y_t = \tilde{\alpha}' M_t + \tilde{\beta}' X_{1,t} + Z_t$$

*Model 2 (“Long regression”):*

$$Y_t = \alpha' M_t + \beta' X_{1,t} + \gamma' X_{2,t} + Z_t$$

$M_t$  = Month variables

$X_{1,t}$  = Additive outliers, level shifts, interventions, &c.

$X_{2,t}$  = Other outliers and interventions not in Model 1

# Two-Stage Runs

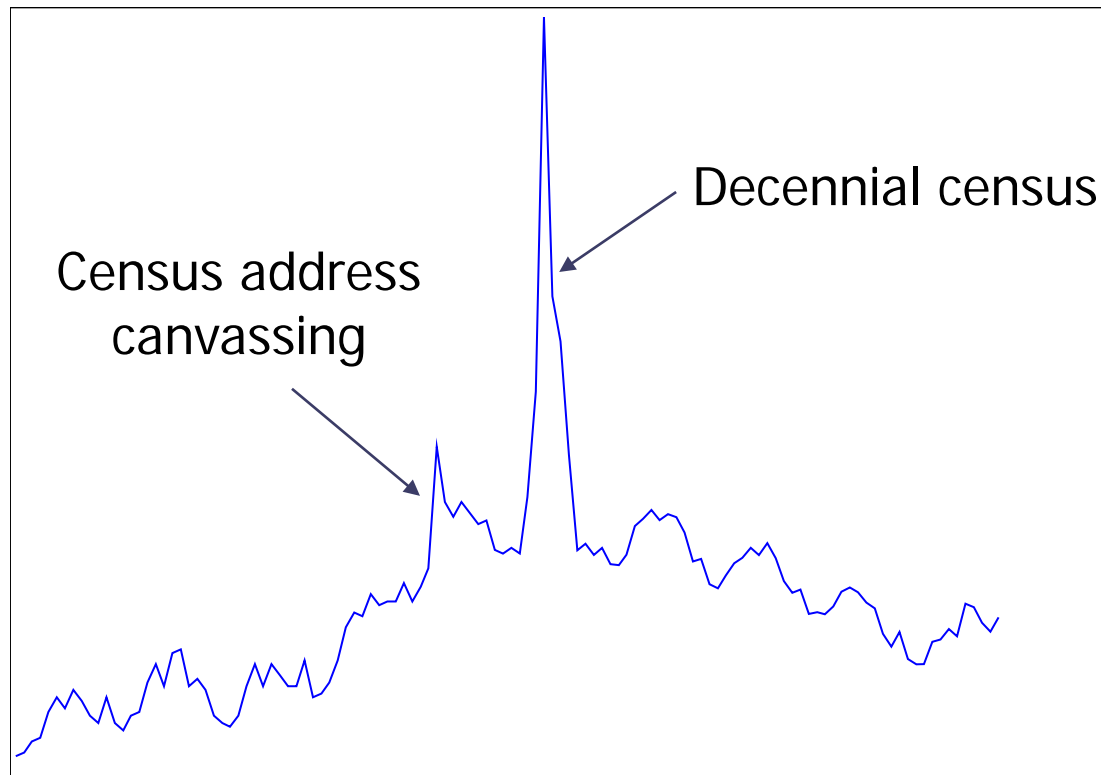
- Potentially biased estimates of  $\alpha$  when omitting  $\gamma'X_2$ 
  - ▶ If calendar and omitted vars correlated
- Proposal: do two runs
  - ▶ First run without  $\alpha'M_t$ 
    - Run auto outlier detection
  - ▶ Second run include  $\alpha'M_t$ 
    - Use outliers from first run in regression spec
- Compare out-of-sample forecast
  - ▶ At series level and aggregate level

# Overall Results

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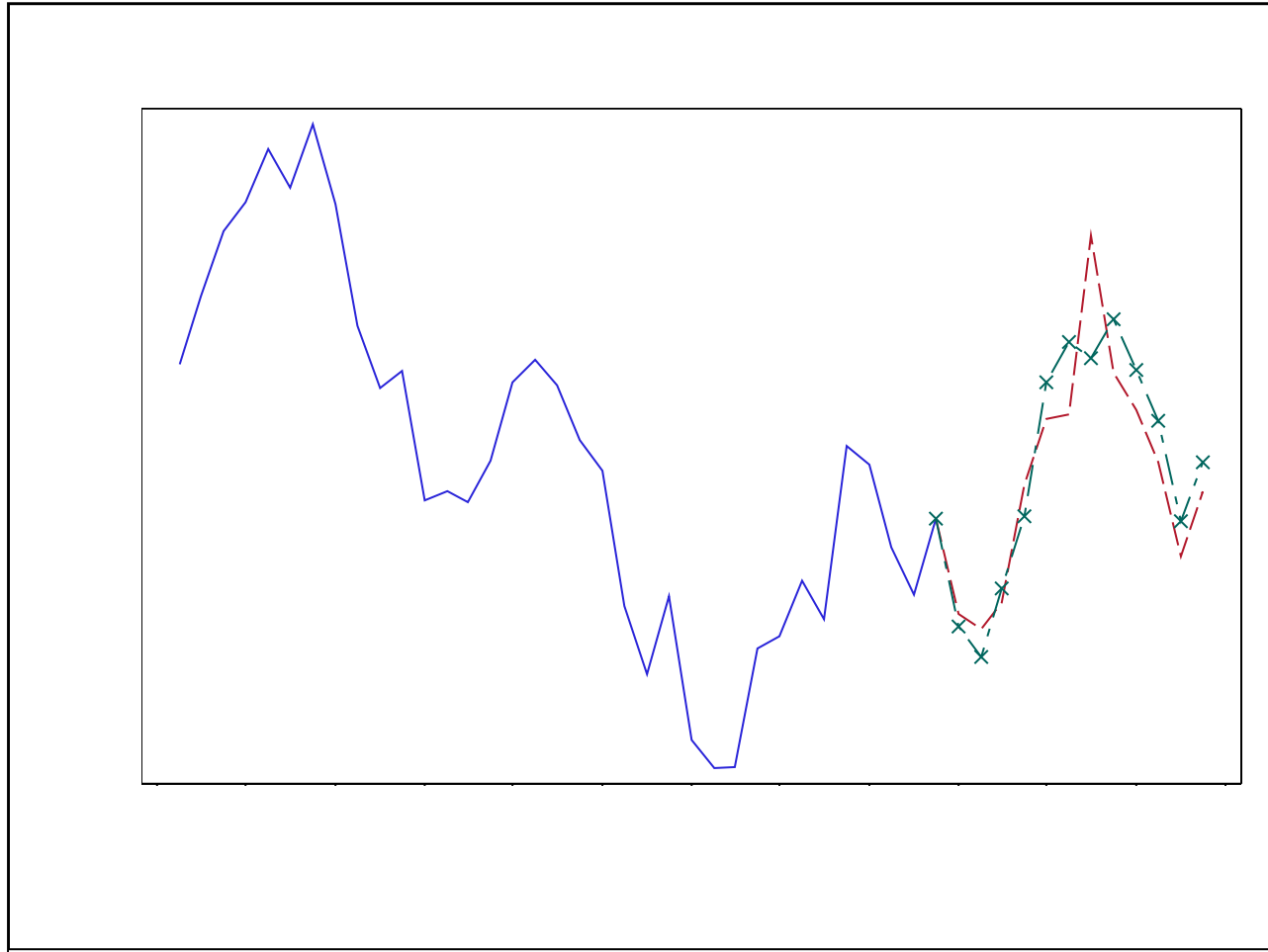
- Slight improvement in forecasts overall
- Noticeable improvement in known decennial census effect
- When BIG differences: alternate run usually better
- Two-stage modeling doesn't have more outliers in model
  - ▶ Perhaps better ones
- A variant where series was prior-adjusted using outliers from first run produced very similar results

# Ex. Nevada, Federal Govt.

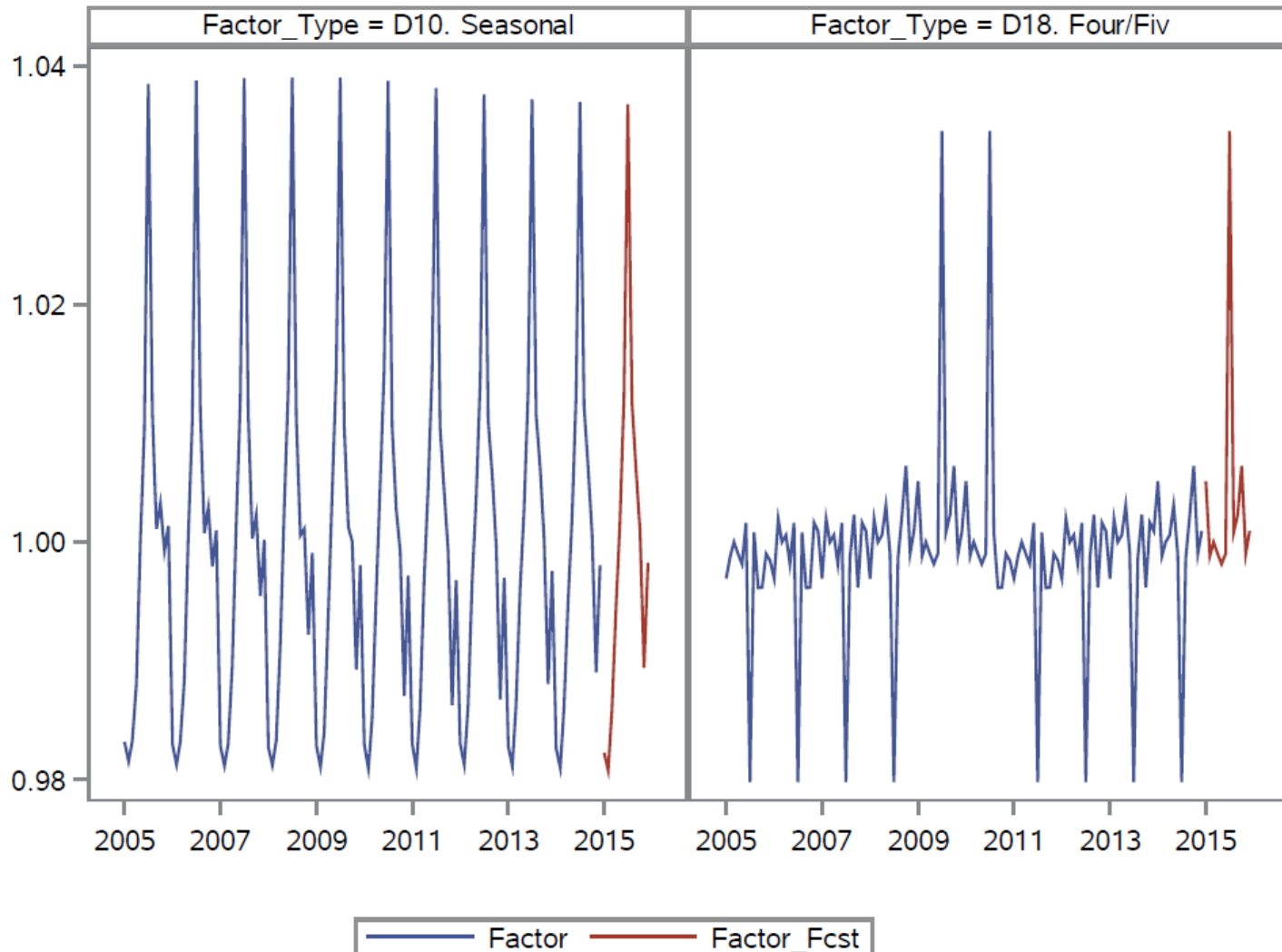




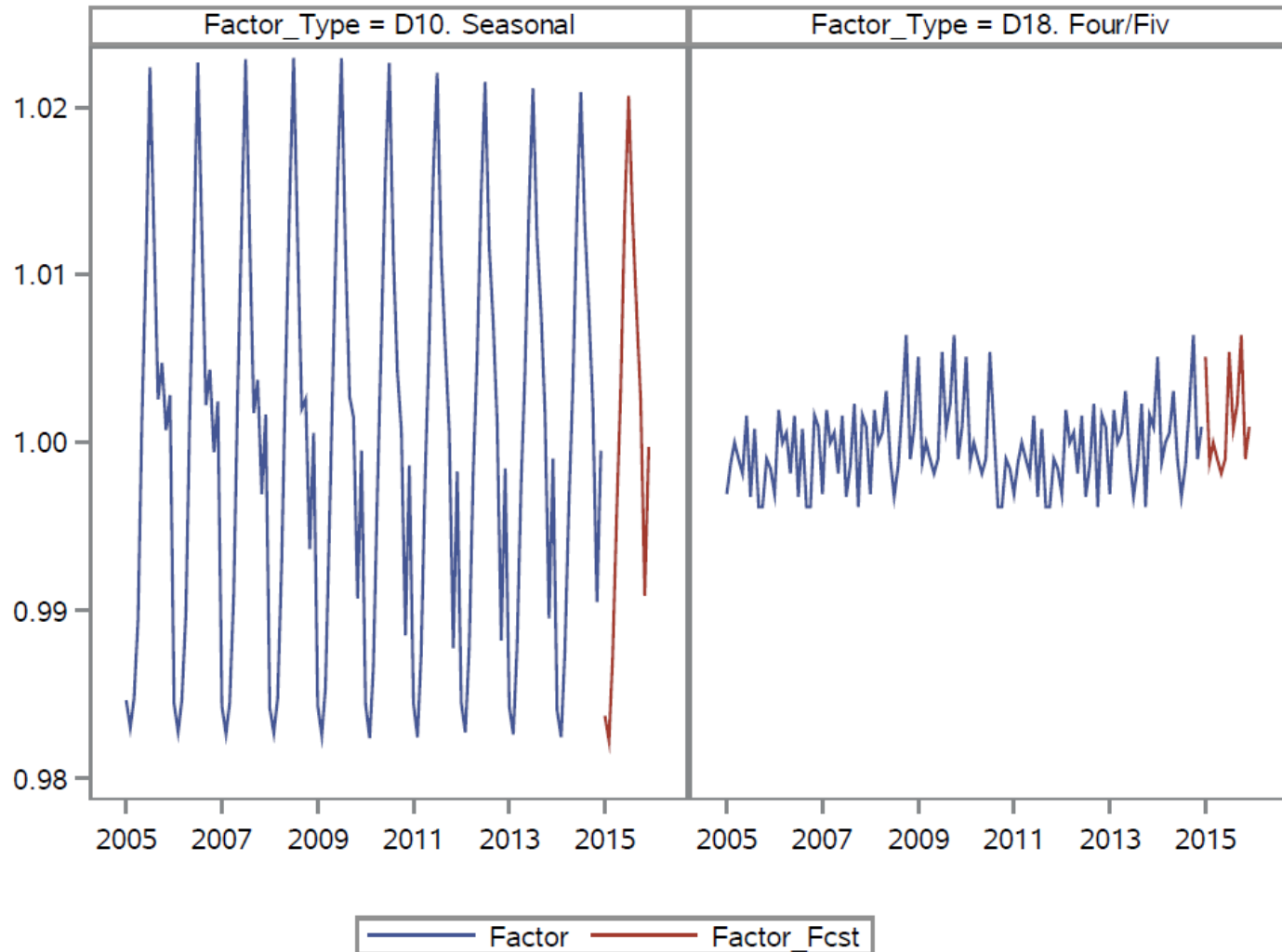
# Ex. Nevada, Federal Govt.



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# Ex. Nevada, Federal Govt.

## Standard Run

Variable	Parameter Estimate	Standard Error	t-value
User-defined			
dum1	0.0016	0.00173	0.91
dum2	-0.0007	0.00153	-0.45
dum3	0.0018	0.00186	0.97
dum4	-0.0021	0.00184	-1.14
dum5	0.0033	0.00170	1.91
dum6	0.0118	0.00196	6.01
dum7	0.0004	0.00168	0.26
dum8	0.0021	0.00185	1.14
dum9	0.0029	0.00168	1.71
dum10	0.0010	0.00155	0.65
dum11	0.0004	0.00168	0.22

## Automatically Identified Outliers

AO2009.Apr	0.0414	0.00435	9.52
LS2010.Mar	0.0228	0.00543	4.19
AO2010.Apr	0.0375	0.00554	6.77
TC2010.May	0.1838	0.00693	26.51
LS2010.Jun	-0.0587	0.00599	-9.80
AO2010.Sep	-0.0261	0.00415	-6.28

## Two-Stage Run

Variable	Parameter Estimate	Standard Error	t-value
User-defined			
dum1	0.0013	0.00207	0.64
dum2	0.0009	0.00181	0.49
dum3	-0.0008	0.00216	-0.37
dum4	-0.0014	0.00216	-0.63
dum5	0.0018	0.00199	0.90
dum6	0.0021	0.00259	0.80
dum7	0.0002	0.00201	0.08
dum8	0.0017	0.00213	0.82
dum9	0.0019	0.00204	0.96
dum10	0.0017	0.00189	0.88
dum11	-0.0003	0.00203	-0.16

## Outliers Identified from Prior Run

AO2009.Apr	0.0451	0.00518	8.72
LS2010.Apr	0.0417	0.00692	6.02
AO2010.May	0.1296	0.00495	26.17
LS2010.Aug	-0.0548	0.00742	-7.38
TC2010.Sep	-0.0401	0.00636	-6.30

# RMSE Ratios

- $RMSE\,R < 1$  shows gain

$$RMSE\,R = \frac{RMSE(r_t^B)}{RMSE(r_t^A)}$$

r=over-the-month growth rates

A=standard run

B=two-stage run

# RMSE Ratios

	2013	2014	2015
All series	0.97	0.89	0.97
Federal only	0.91	0.84	0.89
Sum-of-States	1.08	0.99	1.06

# Takeaways

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- Be careful when adjusting for calendar effects
  - ▶ Correlation of effect and other events causes OVB
- Alternate runs can help to screen for problems
  - ▶ Two-stage not a default for production
  - ▶ Visual screening may be effective
- Subjective prior adjustments or SME-chosen outliers should be considered

# Questions/Comments?

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# Contact Information

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