

Outlier Review During Concurrent Seasonal Adjustment of CES State and Area Series

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Disclaimer

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Introduction



Introduction

- Current Employment Statistics (CES) State and Area program
- Publish over 2,000 seasonally adjusted series each month
- Covering over 400 subnational geographic areas



Background



Background

Traditional Seasonal Adjustment

- Seasonal factors have traditionally been developed using historical employment data forecasted for one year
- Employ the “Two Step” adjustment method
- Outsized impact of various events can have a significant impact on localized areas.

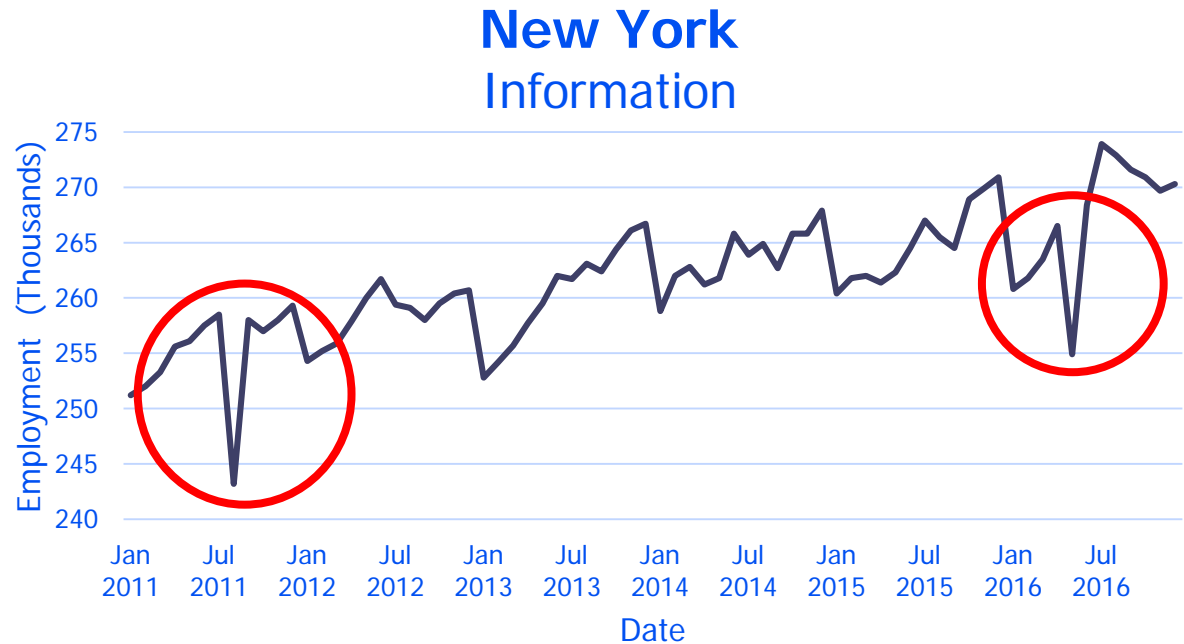
Background

Concurrent Seasonal Adjustment

- CES National has been conducting concurrent adjustment since 2003
- Research was done to investigate the viability of switching to a concurrent seasonal adjustment process in 2015
- Evidence showed that concurrent would make seasonally adjusted data **more accurate** and **less volatile**
 - ▶ *Mance (2015)*

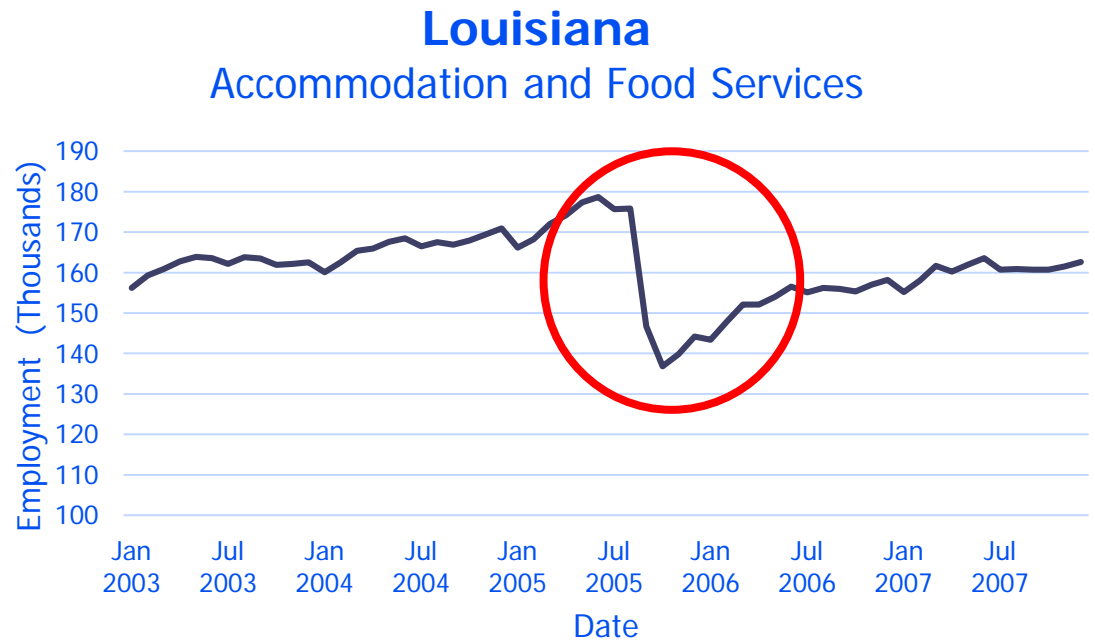
Background

- Additive Outlier
- New York – Information
- Aug. 2011: OTMC -15,300 (-6%)
- May 2016: OTMC -11,600 (-4%)



Background

- Level Shift
- Louisiana – Accom. and Food Service
- OTMC -29,100 (-17%)



Research Motivation



Research Motivation

- Enhance outlier detection capabilities during concurrent seasonal adjustment
 - ▶ McDonald-Johnson and Hood (2001)
- Gains from stricter Critical Value policy

Simulations





ARIMA Models and Parameters

ARIMA Models and Parameters

ARIMA Model	AR	MA	AR12	MA12	d	d12
(0,1,1)(0,1,1)	-	0.23	-	0.68	1	1
(0,2,1)(0,1,1)	-	0.72	-	0.71	2	1
(1,1,0)(0,1,1)	0.43	-	-	0.71	1	1
(0,1,0)(1,1,0)	-	-	-0.54	-	1	1



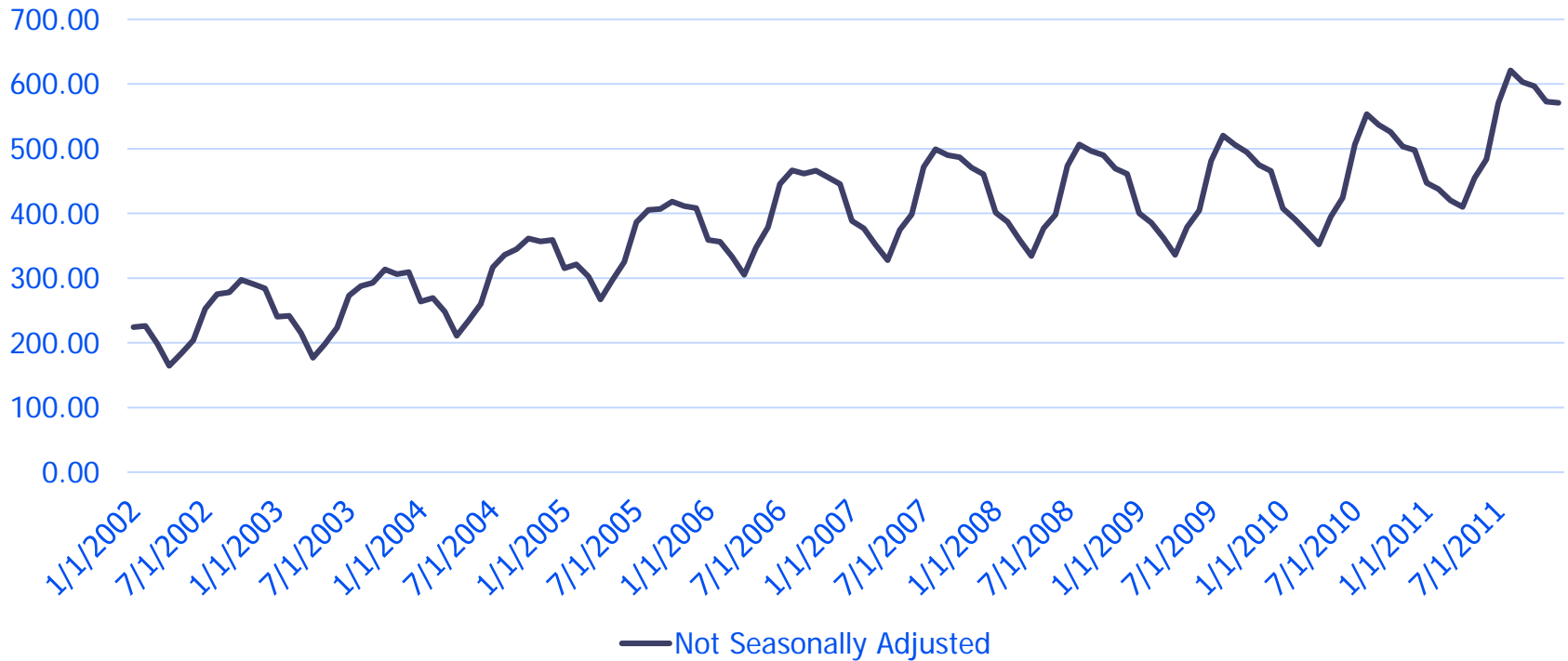
Exogenous Events

Exogenous Events				
Event	Description	Scenario	Model	Graphic
Additive Outlier (AO)	Series is impacted at a single point in time (t)	Labor Strikes	1 for $t = t_0$ 0 for $t \neq t_0$	
Level Shift (LS)	Series is impacted and continues at new level	Hurricane Katrina	-1 for $t < t_0$ 0 for $t \geq t_0$	

- 1,000 series per model

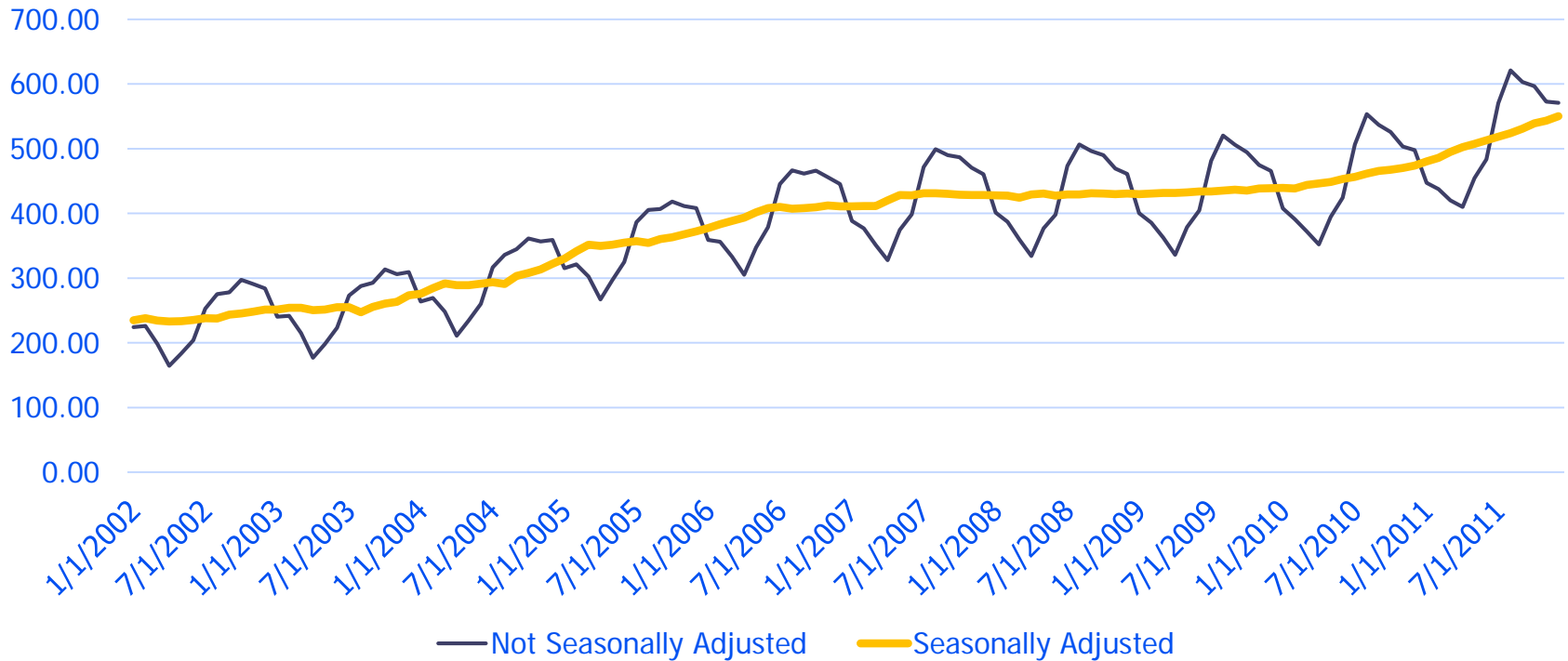
Simulations

Not Seasonally Adjusted



Simulations

Seasonally Adjusted



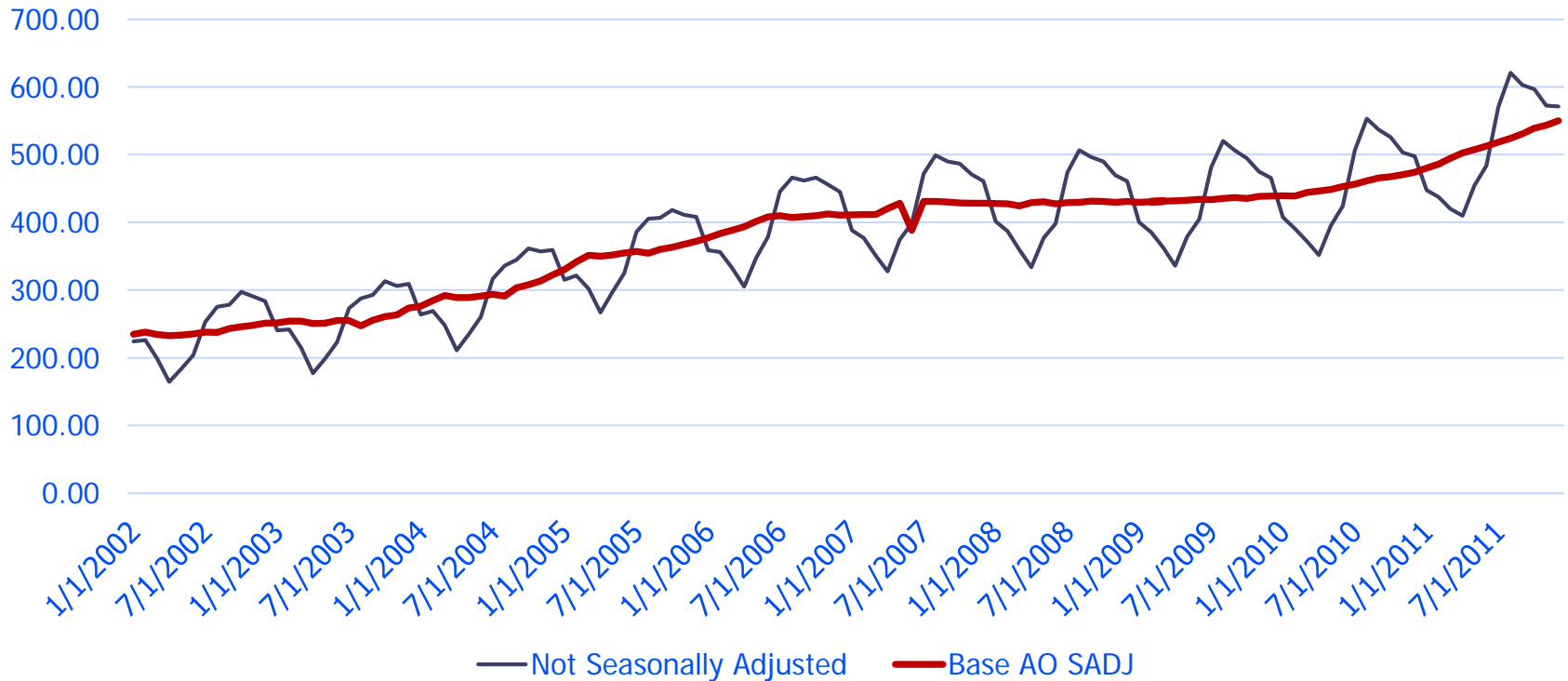
Seasonally Adjusted:

where Z_t = ARIMA model noise component



Simulations

Baseline Adjustment with Additive Outlier



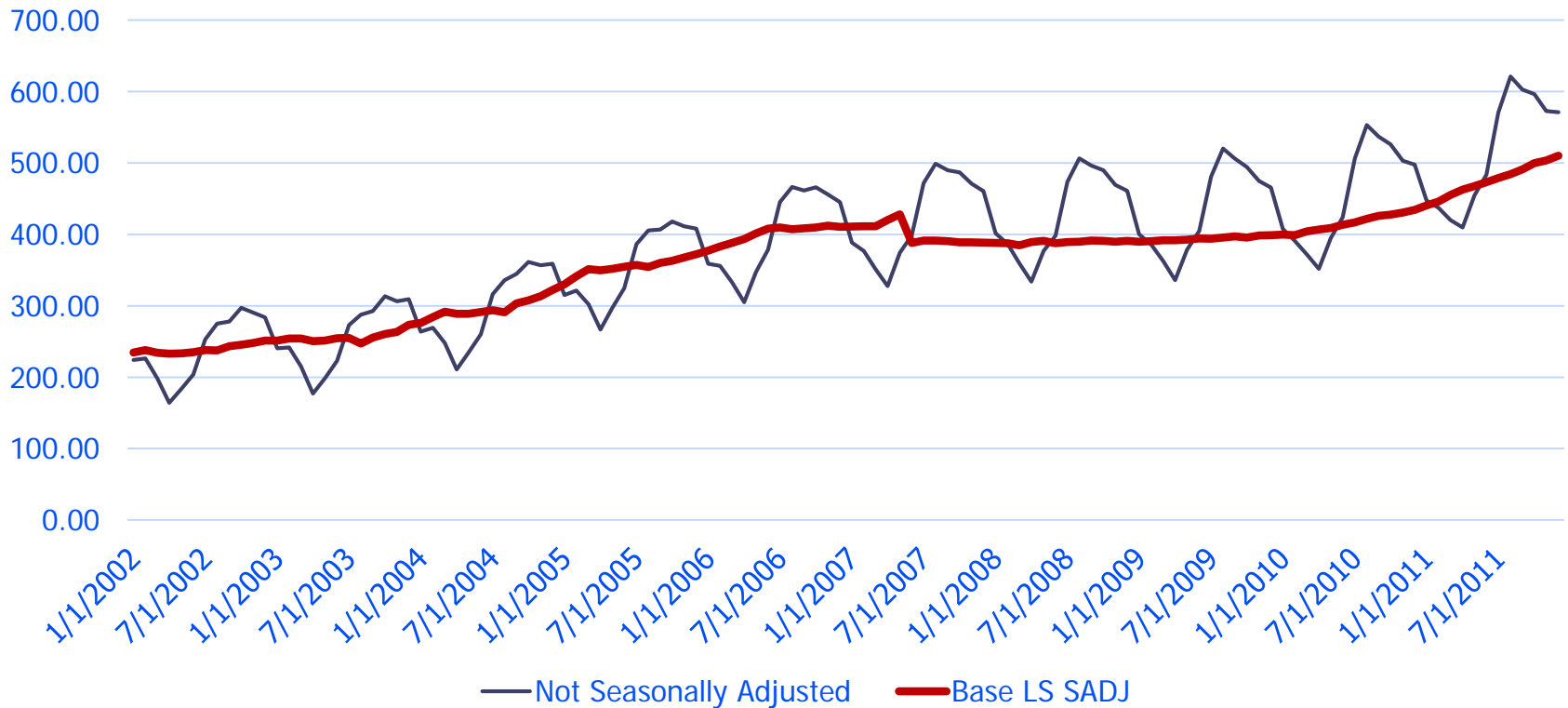
Seasonally Adjusted:

where Z_t = ARIMA model noise component



Simulations

Baseline Adjustment with Level Shift



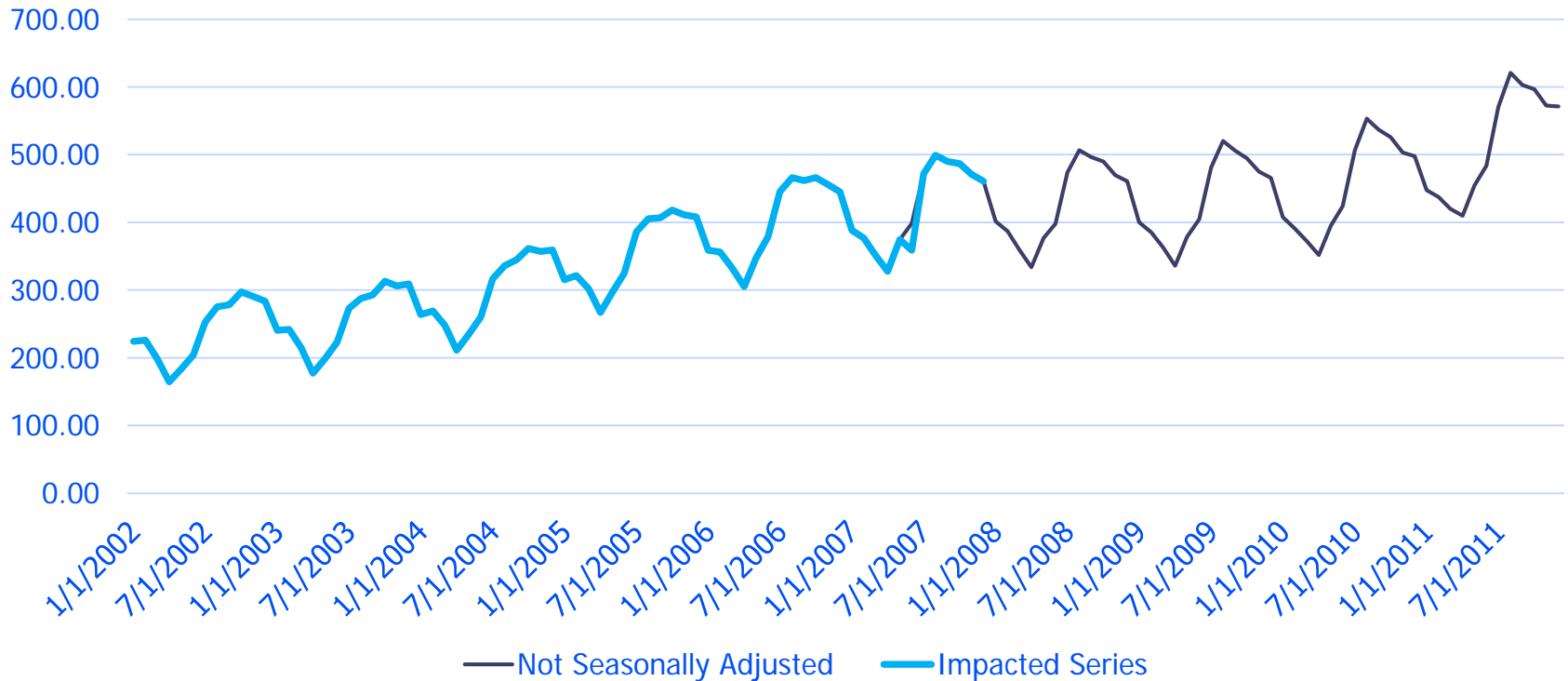
Seasonally Adjusted:

where Z_t = ARIMA model noise component



Simulations

Baseline Adjustment with Impacted Series

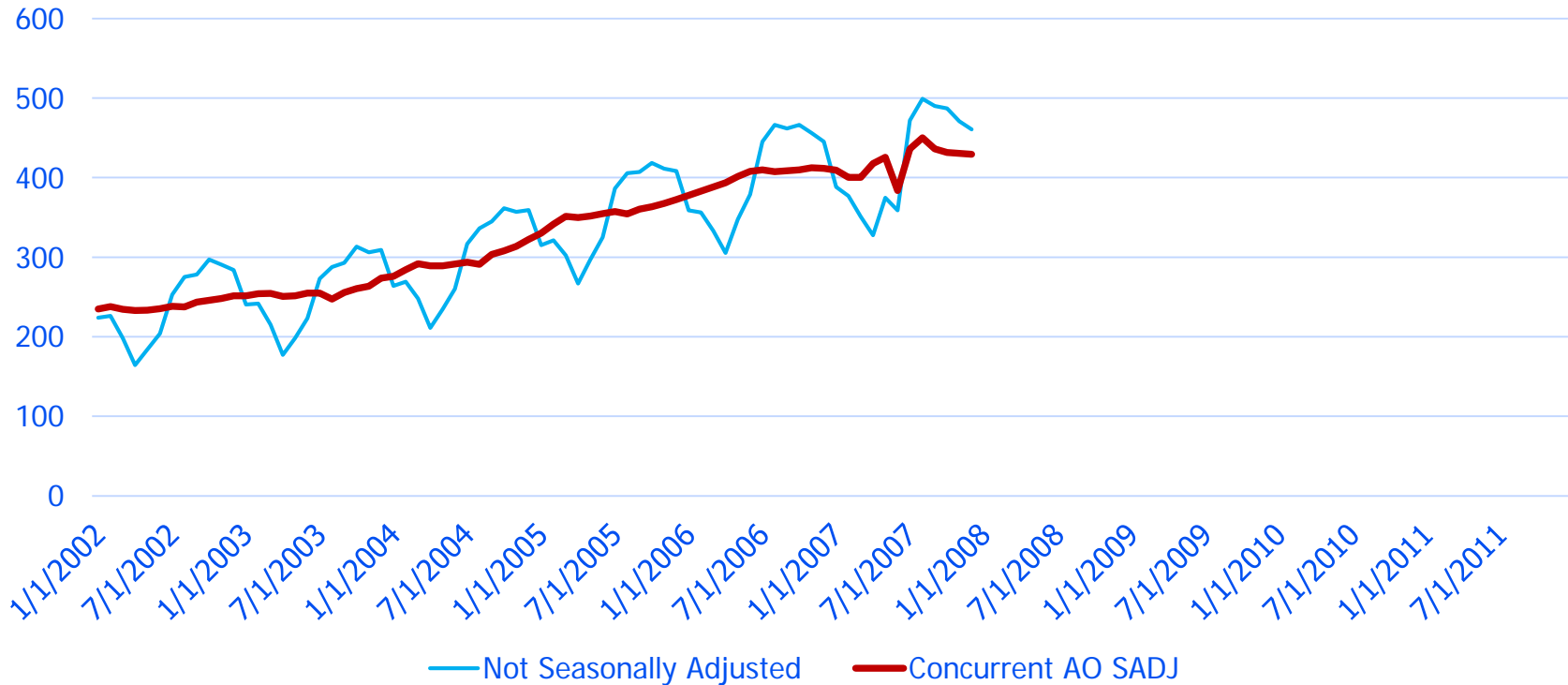


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Impacted Series with Concurrent Additive Outlier

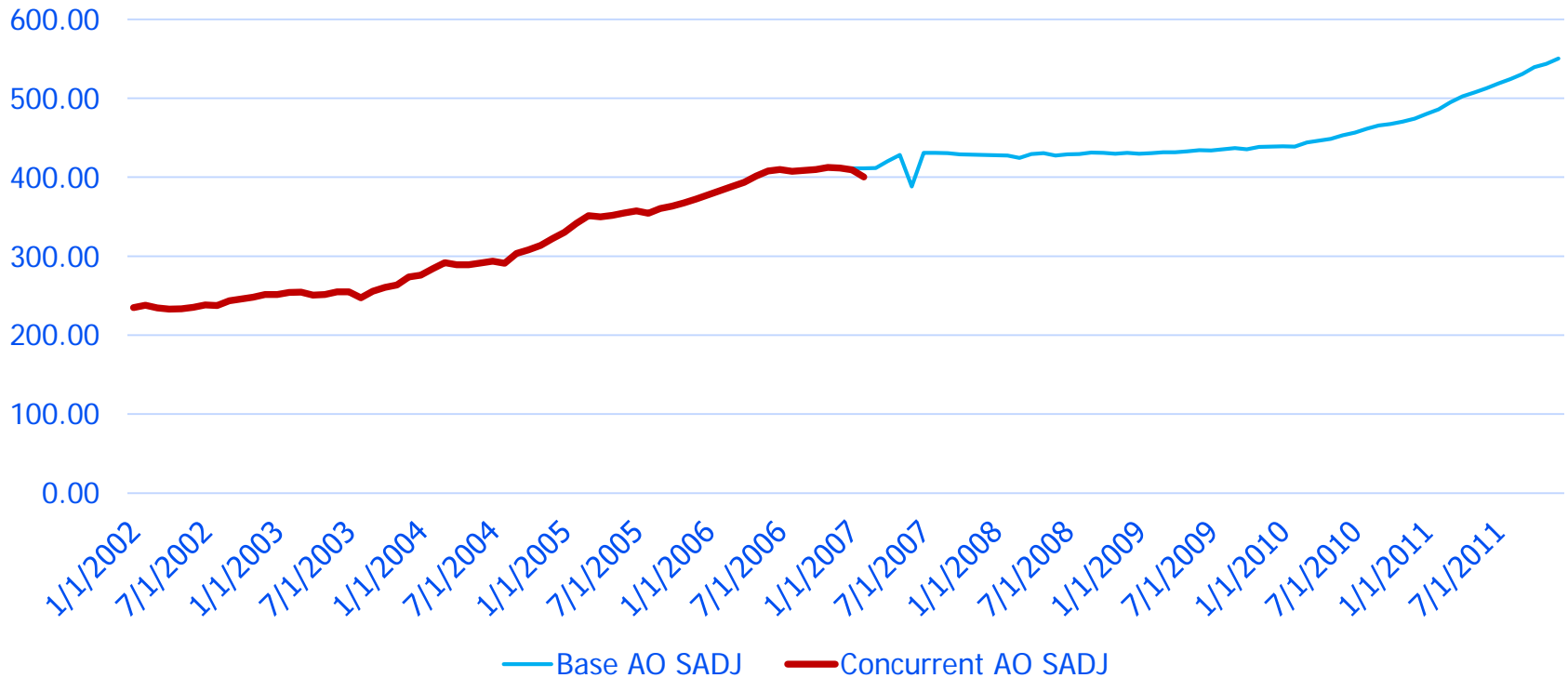


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier

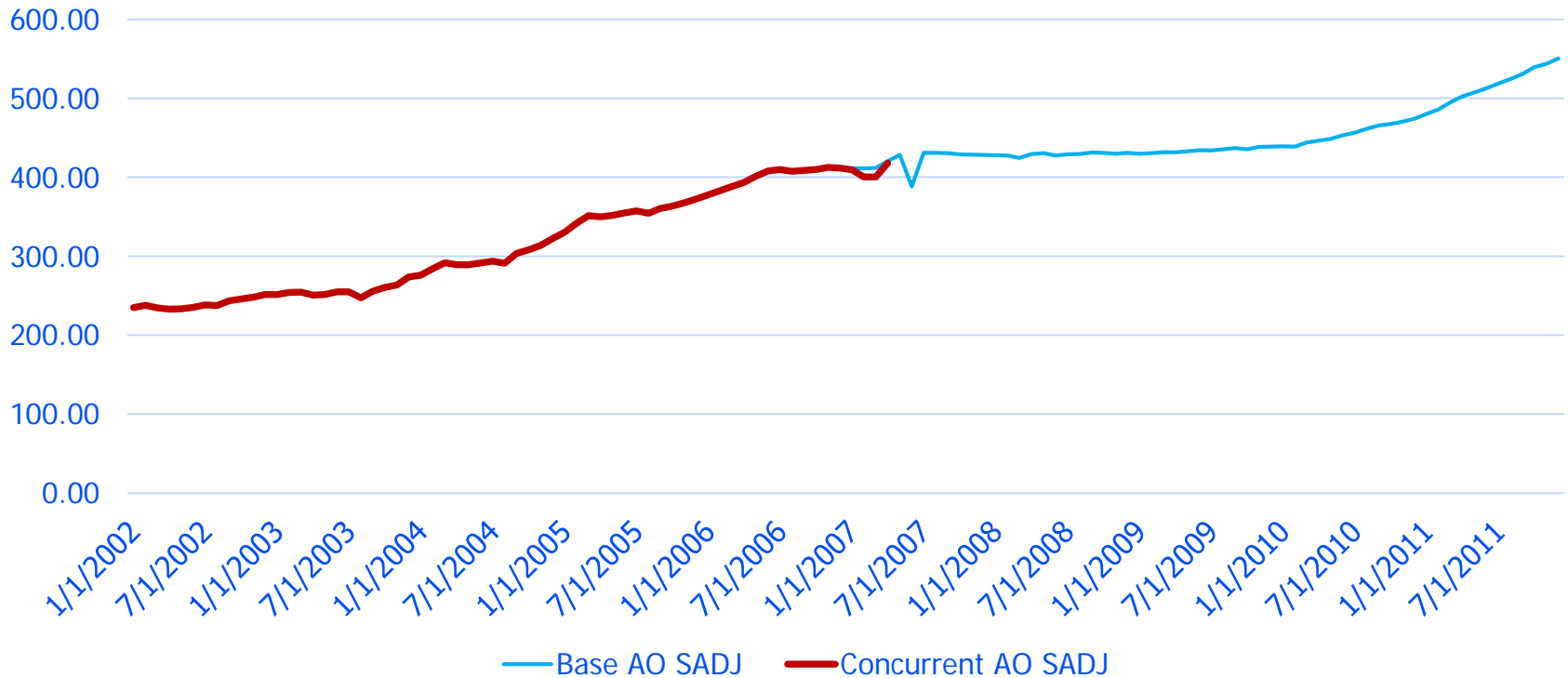


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier

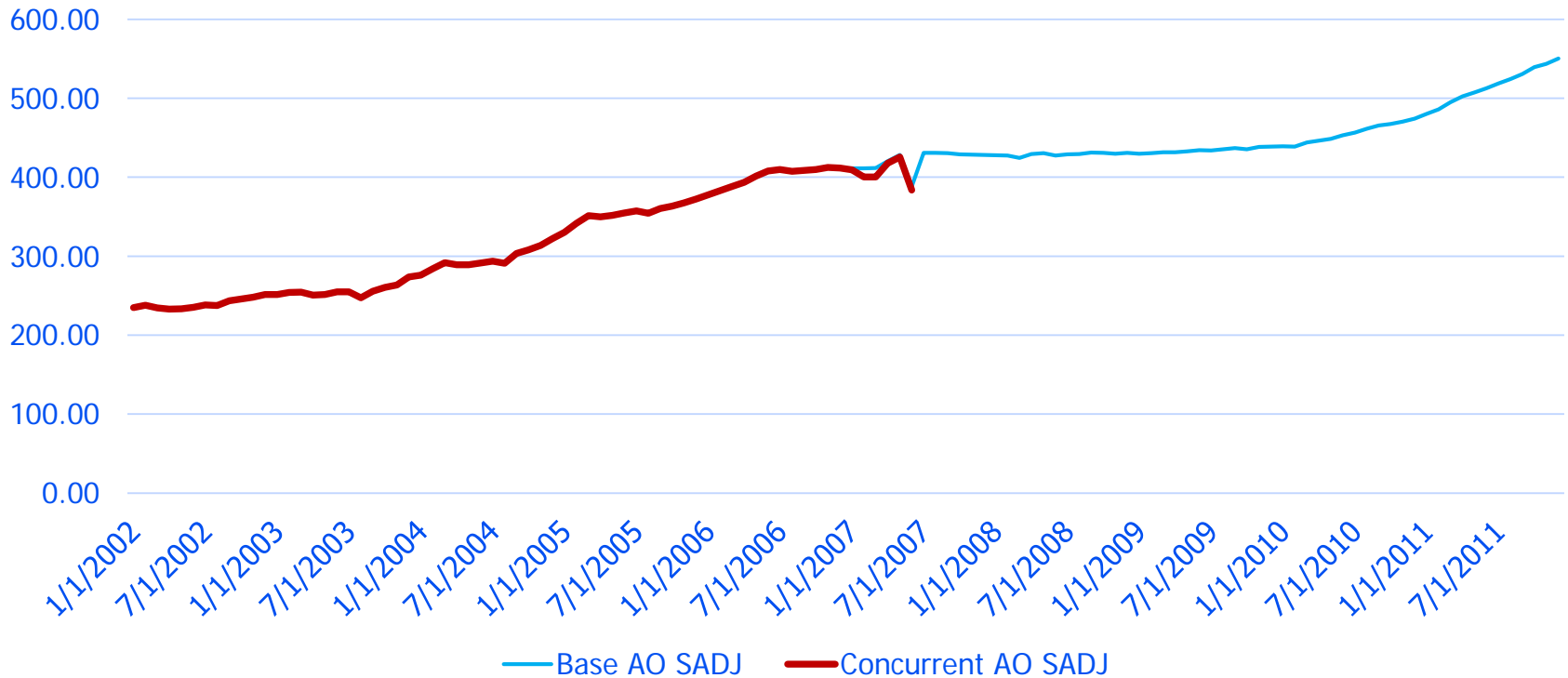


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier

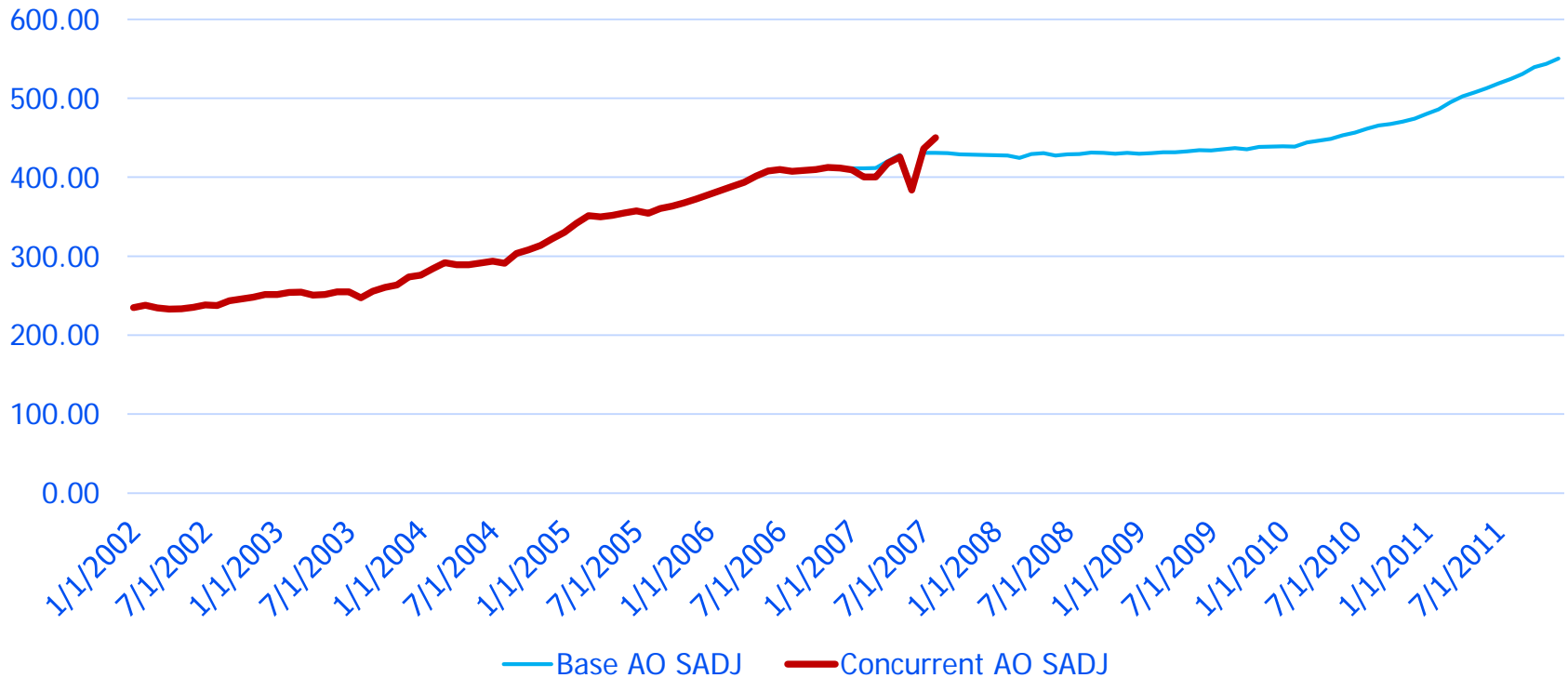


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier

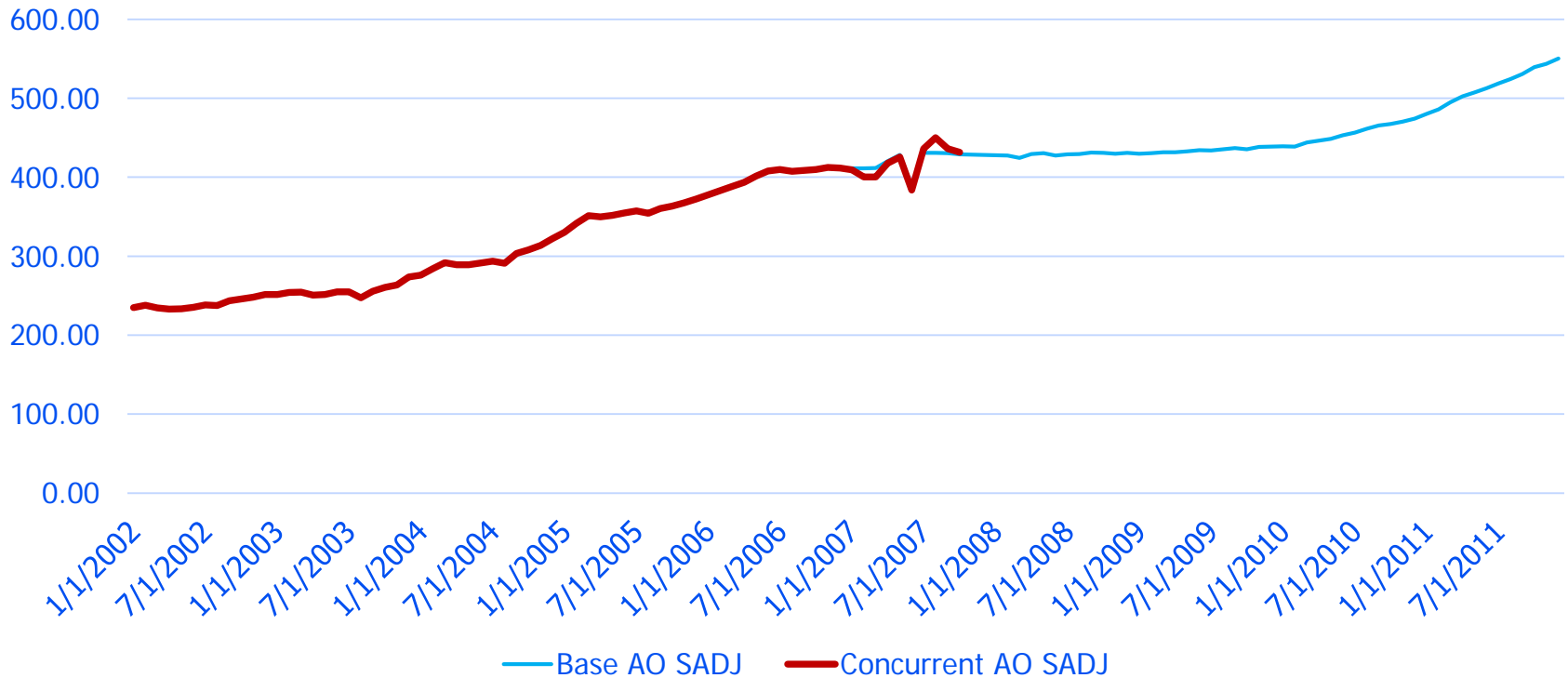


where Z_t = ARIMA model noise component
 AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier



$$y_t = AO_t\beta + Z_t$$

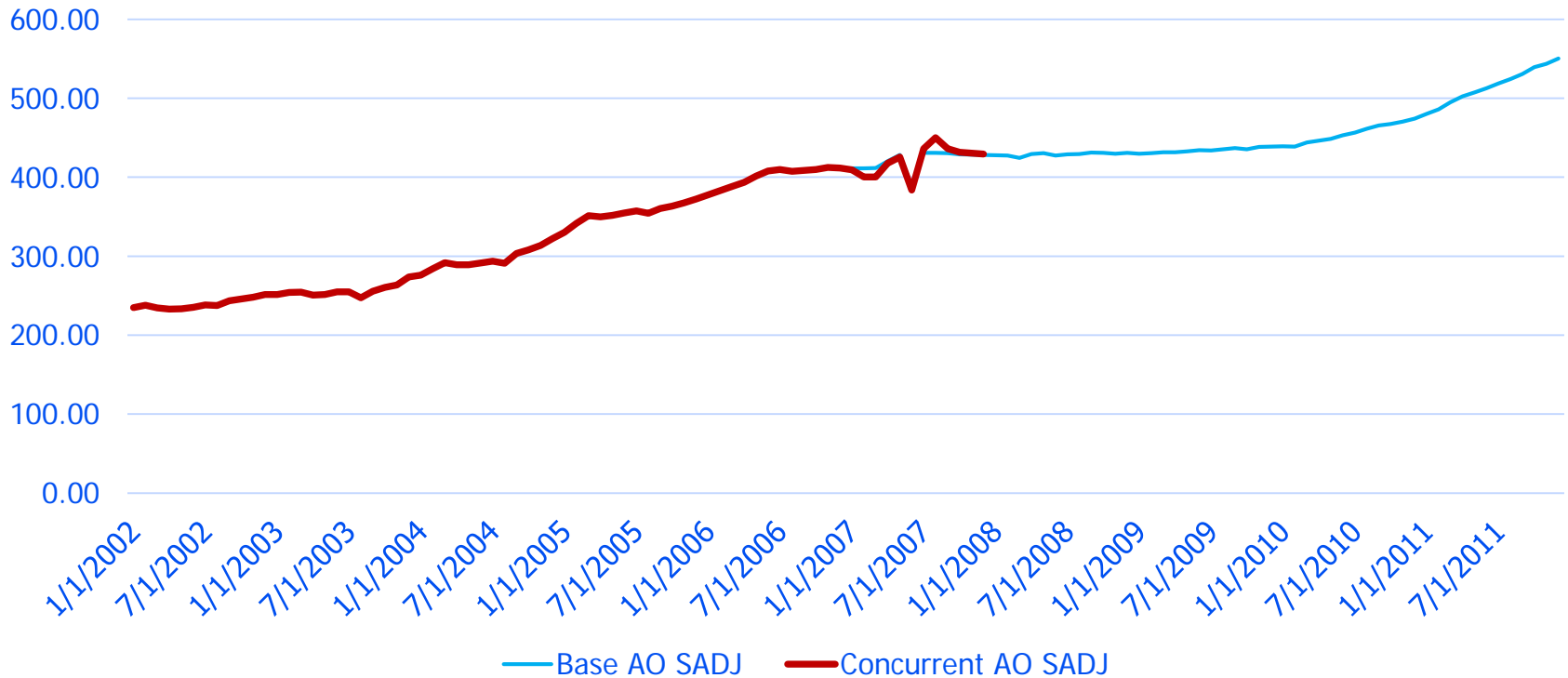
where Z_t = ARIMA model noise component

AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Additive Outlier



$$y_t = AO_t\beta + Z_t$$

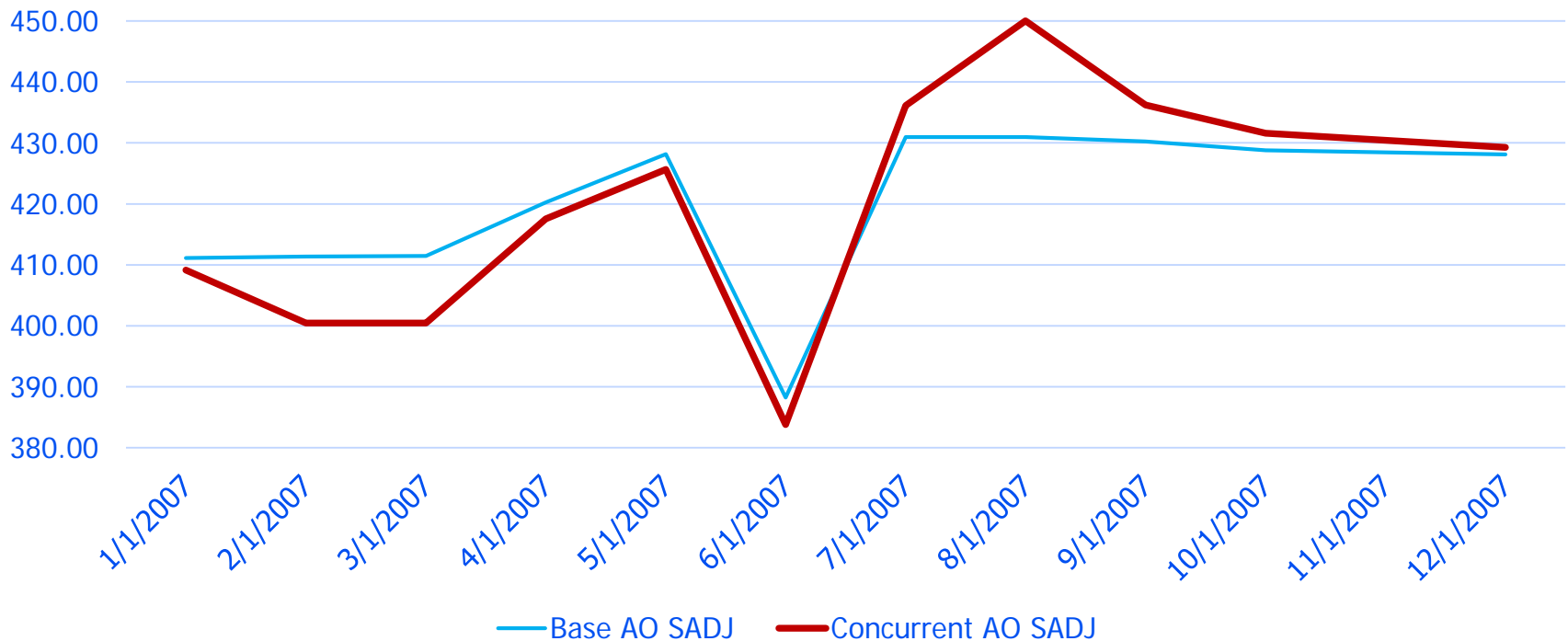
where Z_t = ARIMA model noise component

AO_t = Additive outlier detection parameter



Simulations

Baseline Adjustment with Concurrent Adjustment



$$y_t = AO_t\beta + Z_t$$

where Z_t = ARIMA model noise component

AO_t = Additive outlier detection parameter



Simulations

Event / Model		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85						
	5.00						
	8.00						
	10.00						
	No Detection						



Simulations

Event / Model		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85		X				
	5.00			X			
	8.00				X		
	10.00					X	
	No Detection						

Simulations

Event / Model		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85	X					
	5.00	X					
	8.00	X					
	10.00	X					
	No Detection	X					

Simulations

Event / Model		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85						
	5.00						
	8.00						
	10.00						
	No Detection	X	X	X	X	X	X

Simulations

Event / Model		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85						
	5.00						
	8.00						
	10.00	National CES					
	No Detection						

Results



Results

Gains from stricter Critical Value policy

AO (0,1,1)(0,1,1)		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85	4.386	4.579	4.480	4.429	4.430	4.430
	5.00	4.383	4.686	4.577	4.432	4.429	4.430
	8.00	4.383	4.857	4.939	4.671	4.480	4.430
	10.00	4.383	4.856	4.968	4.999	4.787	4.430
	No Detection	4.383	4.856	4.967	5.105	5.172	14.895



Results

Gains from stricter Critical Value policy

$$RMSE\ ratio_{e,m}(r) = \frac{RMSE(r_{e,m}^{sv})}{RMSE(r_{e,m}^{nv})}$$

$$RMSE\ ratio_{e,m}(r) = 1 \quad \text{“No Value”}$$

Numerators = RMSE of stricter CV for event e and model m

Denominator = RMSE of no CV for event e and model m

Results

Gains from stricter Critical Value policy

AO (0,1,1)(0,1,1)		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85	1.001	0.943	0.902	0.868	0.857	0.297
	5.00	1.000	0.965	0.922	0.868	0.856	0.297
	8.00	1.000	1.000	0.994	0.915	0.866	0.297
	10.00	1.000	1.000	1.000	0.979	0.926	0.297
	No Detection	1.000	1.000	1.000	1.000	1.000	1.000



Results

Gains from stricter Critical Value policy

AO (0,1,1)(0,1,1)		Impact (In terms of equivalent t-value)					
		0.00	3.85	5.00	8.00	10.00	99.00
Critical Value	3.85	1.001	0.943	0.902	0.868	0.857	0.297
	5.00	1.000	0.965	0.922	0.868	0.856	0.297
	8.00	1.000	1.000	0.994	0.915	0.866	0.297
	10.00	1.000	1.000	1.000	0.979	0.926	0.297
	No Detection	1.000	1.000	1.000	1.000	1.000	1.000



Conclusion



Conclusion

- Transition to Concurrent Seasonal Adjustment
- Marginal gain from conducting outlier detection
- Model differences
 - ▶ No significant differences in terms of measure of gain

