

Reference Week Adjustment for Employment Insurance Statistics

November 20, 2019

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Delivering insight through data for a better Canada

Agenda

- Overview of Employment Statistics Program
- Reference Week
- Issues Encountered During Seasonal Adjustment (SA)
- Solution 1: Modelling
- Solution 2: Alternative Data Source

Overview of El Statistics Program

- El Statistics Program:
 - Monthly estimates
 - Statistics on number of EI beneficiaries, number of claims, type of benefits, number of disqualifications and disentitlements.
 - Uses administrative data: administered by Service Canada on behalf of Employment and Social Development Canada
 - Data seasonally adjusted (X-12-ARIMA)





- Reference Week (RW): The week containing the 15th day of the month
- Number of beneficiaries obtained by counting number of people who qualified for EI benefits during the reference week (whether or not they have received benefits for other weeks)
- Reference week introduces a calendar effect due to location of the 15th day.
- Impacts month-to-month change in the EI statistical estimates.

| MONTH | | | | | | |
|-------|-----|-----|-----|-----|-----|-----|
| SUN | MON | TUE | WED | THU | FRI | SAT |
| | | | 1 | 2 | 3 | 4 |
| 5 | 6 | 7 | 8 | 9 | 10 | 11 |
| 12 | 13 | 14 | 15 | 16 | 17 | 18 |
| 19 | 20 | 21 | 22 | 23 | 24 | 25 |
| 26 | 27 | 28 | 29 | 30 | 31 | |
| | | | | | | |





- Prior to 2017: beneficiaries wait 2 weeks before receiving EI benefits.
- Adjustment is made relative to where the 15th falls compared to Wednesday; done via a linear regARIMA model:

$$y_t = \sum_i \beta_i x_{it} + z_t, \qquad z_t \sim ARIMA$$

- where:
 - y_t is the dependent time series
 - x_{it} are the regression variables depending on time t
 - β_i are the regression parameters
 - \mathbf{z}_t are the regression residuals



Positive adjustment

| | MONTH | | | | | |
|-----|-------|-----|-----|-----|-----|-----|
| SUN | MON | TUE | WED | THU | FRI | SAT |
| | 1 | 2 | 3 | 4 | 5 | 6 |
| 7 | 8 | 9 | 10 | 11 | 12 | 13 |
| 14 | 15 | 16 | 17 | 18 | 19 | 20 |
| 21 | 22 | 23 | 24 | 25 | 26 | 27 |
| 28 | 29 | 30 | 31 | | | |
| | | | | | | |

Negative adjustment

| MONTH | | | | | | |
|-------|-----|-----|-----|-----|-----|-----|
| SUN | MON | TUE | WED | THU | FRI | SAT |
| | | | | | | 1 |
| 2 | 3 | 4 | 5 | 6 | 7 | 8 |
| 9 | 10 | 11 | 12 | 13 | 14 | 15 |
| 16 | 17 | 18 | 19 | 20 | 21 | 22 |
| 23 | 24 | 25 | 26 | 27 | 28 | 29 |
| 30 | 31 | | | | | |

$$x_{it} = +2$$

$$x_{it} = \delta_{it} - 15$$

where δ_{it} is Wednesday's date during RW

$$x_{it} = -3$$

As of 2017: new legislation changed the waiting period to one week.



Before 2017: 2 week waiting period



As of 2017:
1 week waiting period

■ This introduced a problem for the reference week adjustment we were using.



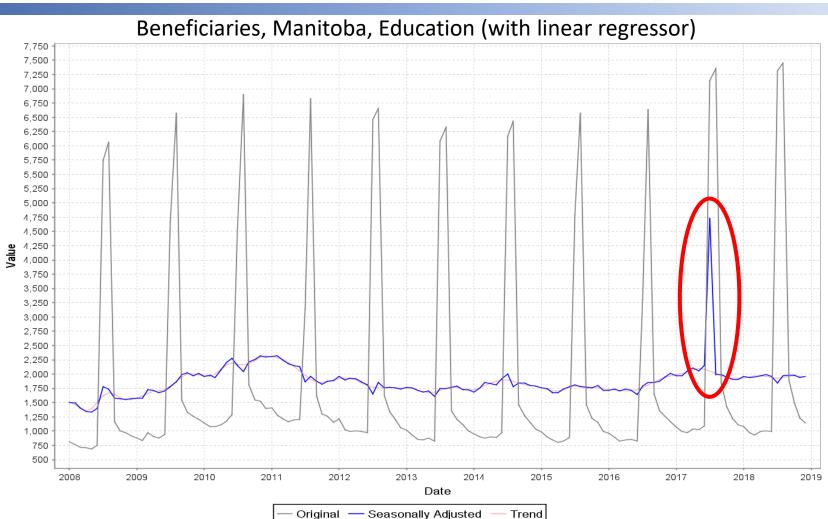
- Most beneficiaries would request EI benefits at end of month:
 - Before 2017: Waiting period often intersects reference week
 - After 2017: Waiting period rarely intersects reference week (big impact on Health and Education sectors)





- As of 2017, reference week adjustment did not seem necessary
- Keeping the linear reference week adjustment: spikes in the data would be introduced post-2017.
- Removing the reference week adjustment: spikes introduced in pre-2017 data.
- Obviously, the linear regressor was not working as we wanted.
- Had to think of a new solution to take care of this problem.





Solution 1: Use Various Non-Linear Models

• Here, we tried various non-linear models and compared to linear model:

$$f(x_{it}) = \sqrt[3]{x_{it}}$$

$$f(x_{it}) = \exp(x_{it})$$

$$f(x_{it}) = x_{it}^2$$

$$f(x_{it}) = x_{it}^2 + x_{it}$$

$$f(x_{it}) = \expit(x_{it})$$

$$\operatorname{expit}(x_{it}) = \operatorname{logit}^{-1}(x_{it}) = \operatorname{log}^{-1}\left(\frac{x_{it}}{1 - x_{it}}\right) = \frac{\exp(x_{it})}{\exp(x_{it}) + 1}$$

■ 1341 series: 6 models, 12 months \rightarrow 96,552 results







Looking at the best model fit for each month based on R²_{adj}, AIC, BIC (16,092 models)

R² adj

% Model Frequency 5.56 Linear Cube root 21.11 Exponential 11.38 37.61 Quadratic Quadratic with 17.83 linear term Expit 6.50 AIC

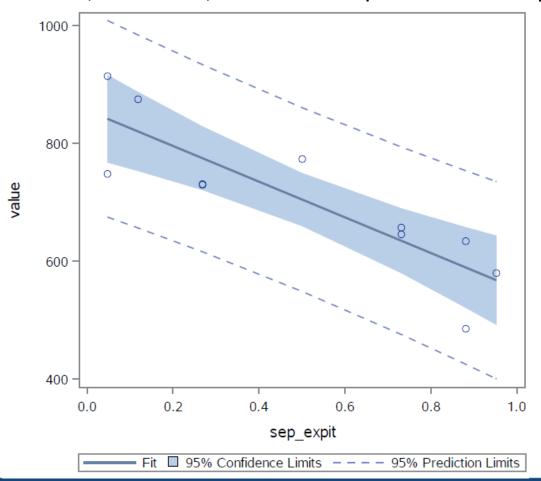
| Model | % Frequency | | |
|----------------------------|----------------|--|--|
| Linear | 6.05 | | |
| Cube root | 22.20 | | |
| Exponential | 12.93 | | |
| Quadratic | 42.09 | | |
| Quadratic with linear term | 9.57 | | |
| Expit | 7.16 | | |

BIC

| Model | % Frequency |
|----------------------------|----------------|
| Linear | 6.33 |
| Cube root | 23.01 |
| Exponential | 13.97 |
| Quadratic | 44.44 |
| Quadratic with linear term | 4.72 |
| Expit | 7.53 |

Solution 1: Use Various Non-Linear Models

Beneficiaries, Manitoba, Education: Expit model for September



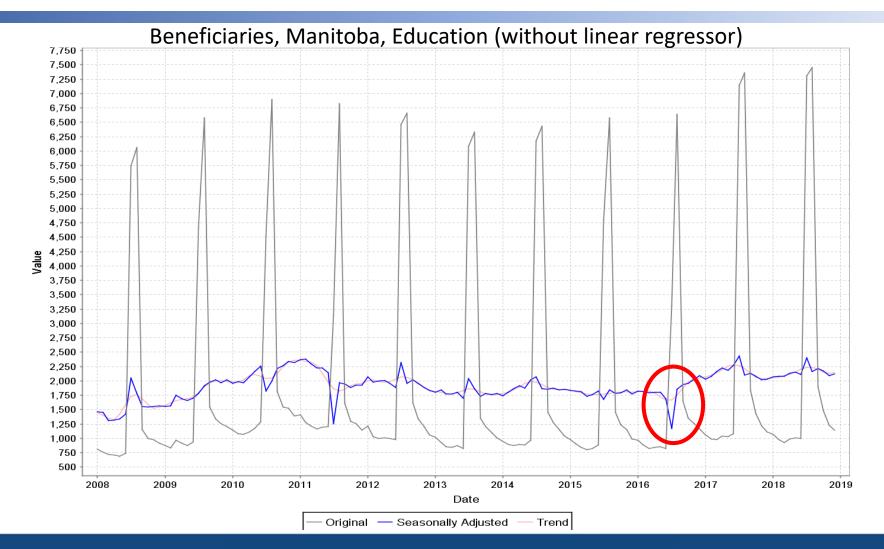


Solution 2: Use Detailed Data

- Pretend beneficiaires prior to 2017 had one-week waiting period (we know the date when people applied for EI)
- Since reference week adjustment works well for up to end of 2016, we would do seasonal adjustment in 2 parts:
 - Prior to 2017: use the linear regARIMA model
 - Use the information from the detailed data where we pretend we have a 1-week period up to end of 2016, and use the actual data in 2017 onward.
 - Results are better and results in smoother graphs with no spikes.



Solution 2: Use Detailed Data

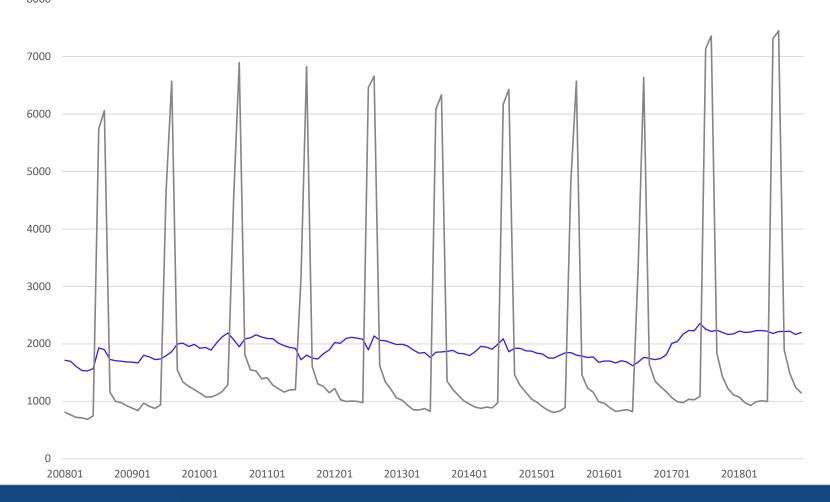




13

Solution 2: Use Detailed Data

Beneficiaries, Manitoba, Education (hybrid approach; in production)







Conclusion

- Discontinuity and spike in our data gave us issues in seasonal adjustment.
- 2 solutions considered: modelling and alternative data source.
 - The hybrid model we chose in the end removed spikes and worked well.
- Maintain the linear models for now
 - There may be potential for considering the other models (perhaps quadratic) and may leave the door open to other models when we get more data.





Questions?

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