

# FDA Rule Analysis Quarto

## Intro

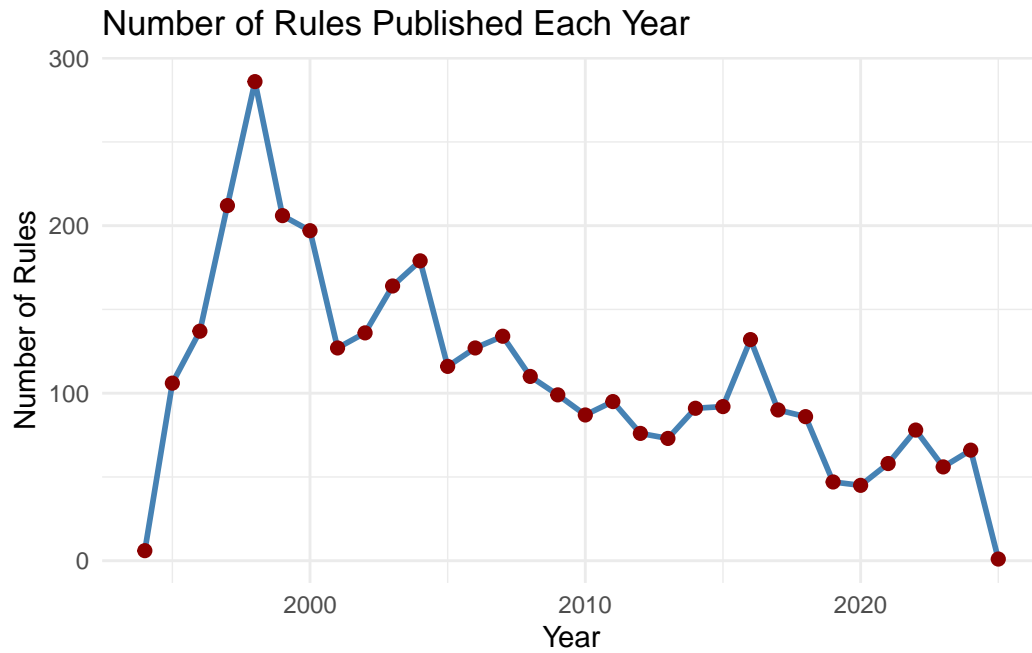
In the summer of 2024, the Supreme Court issued two landmark rulings – *Ohio v. EPA* on June 27, 2024, and *Loper Bright Enterprises v. Raimondo* on June 28, 2024 – that transformed the regulatory landscape for federal agencies. In *Ohio v. EPA*, the Court held that an agency’s failure to adequately respond to significant public comments during the notice-and-comment process renders its rules arbitrary and capricious. The following day, in *Loper Bright v. Raimondo*, the Court overruled Chevron deference, directing lower courts to interpret statutory ambiguities independently rather than deferring to agency interpretations. In doing so, the Court substantially reduced agency latitude in interpreting legislative “gray areas”.

We hypothesize that these decision created an environment where agencies are less likely to publish formal rules due to the increased logistical burden of addressing public comments post-*Ohio v. EPA* and the heightened risk of litigation over statutory interpretations post-*Loper Bright* likely discourage rulemaking. In addition, we propose a broader hypothesis: the overall proportion of formal rules has been decreasing since 2000. This decline is likely driven by a combination of factors. First, evolving expectations for transparency and adaptability have encouraged agencies to favor more flexible, informal governance mechanisms. Second, the cumulative impact of judicial constraints—exemplified by recent rulings—further disincentivizes the proliferation of formal rulemaking.

To investigate these hypotheses, we employ quantitative methods—specifically, regression-discontinuity and interrupted time series analyses—using the period following the release of *Loper Bright Enterprises v. Raimondo* (starting June 29, 2024) as a cutoff, complemented by a longitudinal analysis of rulemaking trends dating back to 2000. This study aims to empirically assess how these Supreme Court rulings, in tandem with broader administrative shifts, have impacted agency rulemaking behavior.

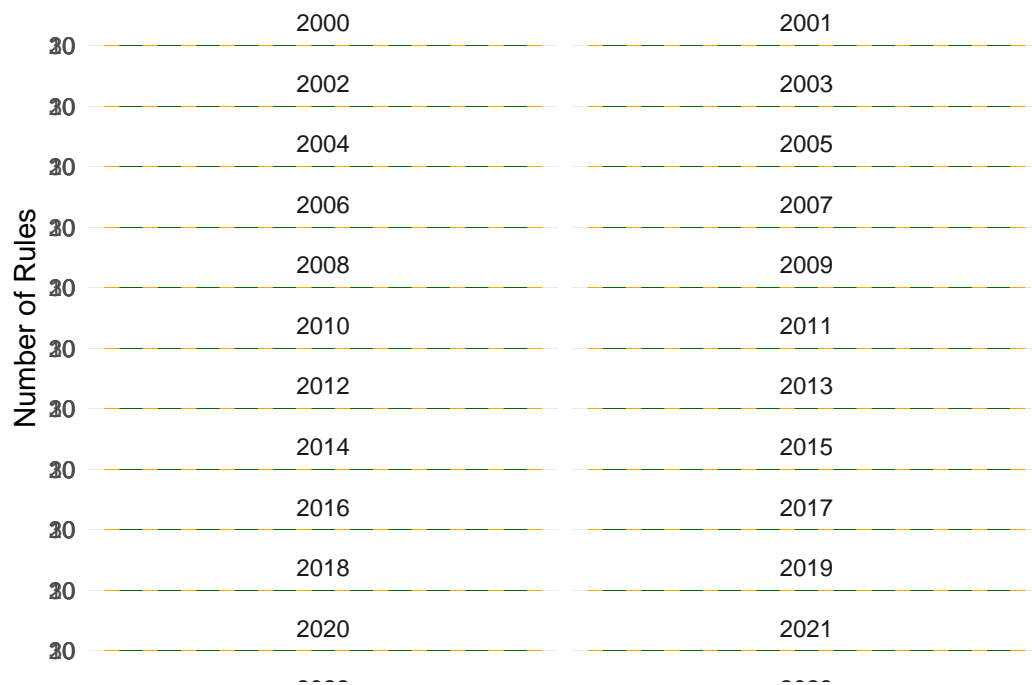
## Running Code

Now, create plots from df data frame

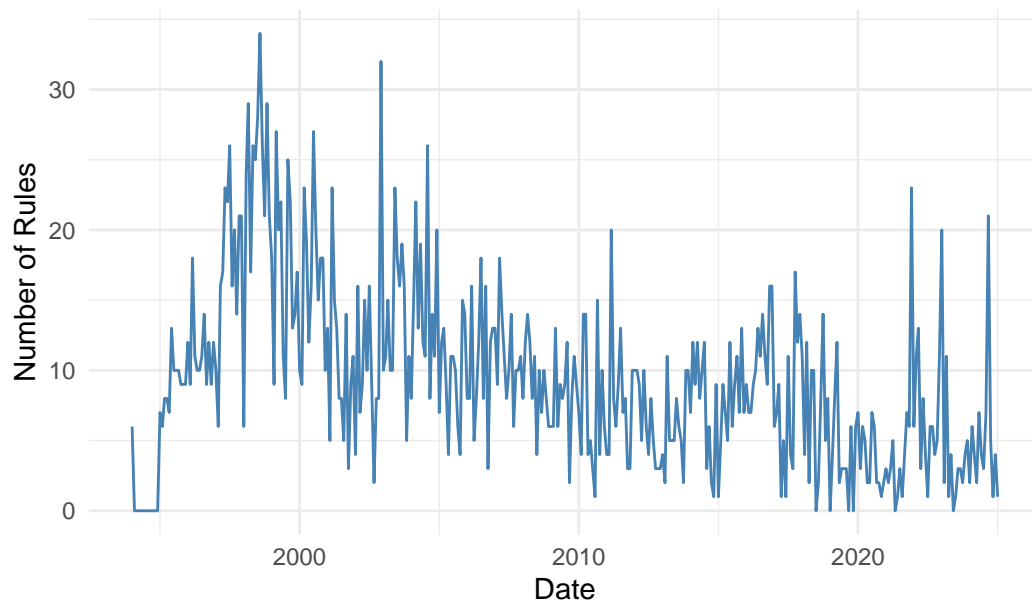


Next, let's analyze the data broken into its constituent months (the first visualization doesn't work yet)

``geom_line()``: Each group consists of only one observation.  
i Do you need to adjust the group aesthetic?



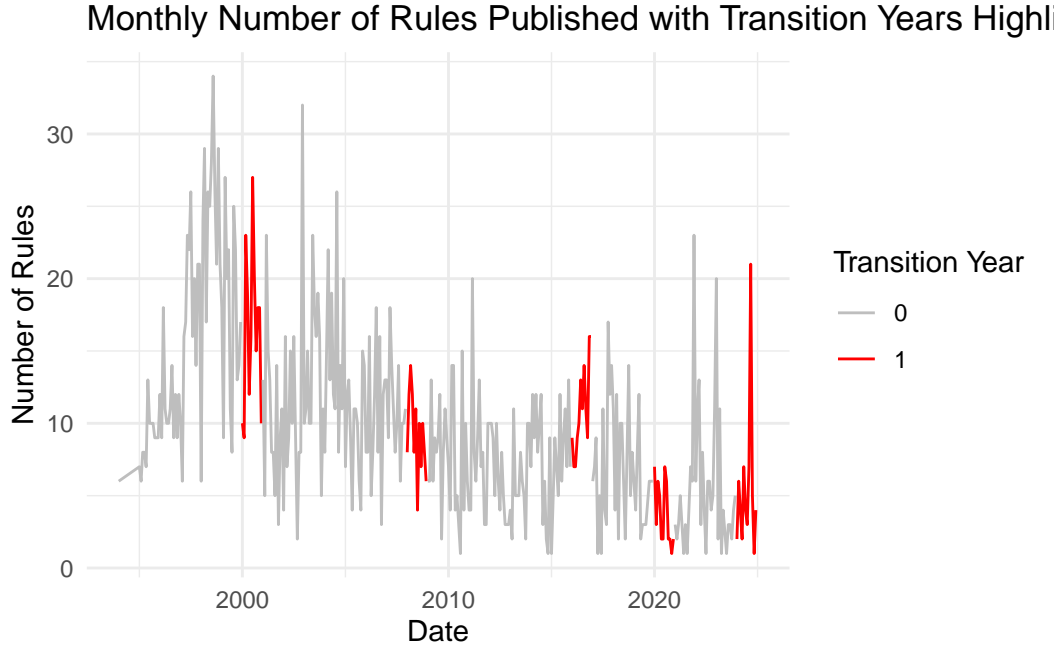
Monthly Number of Rules Published Over Time



The following graph depicts the data broken into constituent months and highlights Presidential transition years.

``summarise()`` has grouped output by 'year'. You can override using the

`.groups` argument.



### Regression Discontinuity Analysis

$$Y_i = \alpha + \tau D_i + \beta X_i + \gamma X_i^2 + \sum_{m=1}^{11} \delta_m M_{im} + \theta T_i + \epsilon_i$$

Where:

- $(Y_i)$ : Number of Rules Published in month  $(i)$ .
- $(\alpha)$ : Intercept term (baseline level of  $(Y)$ ).
- $(D_i)$ : Treatment Indicator for month  $(i)$ .  $D_i = 1$  if month  $i$  is after June 2024 and 0 otherwise.
- $(X_i)$ : Running Variable representing the distance from the cutoff (June 2024) for month  $i$ . Measured in months:  $X_i = \text{Number of months since June 2024}$ 
  - $(X_i > 0)$ : Post-June 2024 (treatment group)
  - $(X_i < 0)$ : Pre-June 2024 (control group)
- $(\beta)$ : The Treatment Effect

- $(X_i^2)$ : Quadratic Term to capture potential non-linear trends in the data.
- $(M_{im})$ : Monthly Dummy Variables for each month ( $m$ , January to December), excluding one month to avoid multicollinearity (January is the reference category).
- $(\delta_m)$ : Coefficients for each monthly dummy variable, capturing the effect of being in month ( $m$ ) relative to the reference month.
- $(T_i)$ : Presidential Transition Indicator for month ( $T_i$ ):  $T_i = 1$  for 2008, 2016, 2020, or 2024 and 0 otherwise.
- $(\theta)$ : Coefficient capturing the effect associated with presidential transition years.
- $(\epsilon_i)$ : Error Term capturing unobserved factors affecting  $(Y_i)$ .

```
# A tibble: 6 x 4
  year month    rules guidance
  <dbl> <chr>   <int>     <int>
1  2000 January     10         4
2  2000 February    9         2
3  2000 March      23         5
4  2000 April      19         3
5  2000 May        12         0
6  2000 June       16         1
```

Call:

```
lm(formula = rules ~ treatment * distance + month + transition,
    data = rd_data_all)
```

Residuals:

```
      Min       1Q   Median       3Q      Max
-9.6733 -2.9071 -0.5818  2.2703 17.4854
```

Coefficients:

```
              Estimate Std. Error t value Pr(>|t|)
(Intercept)    3.524881   0.547545   6.438 5.13e-10 ***
treatment       5.622432   4.010943   1.402  0.16207
distance      -0.032746   0.003122 -10.490 < 2e-16 ***
month.L         0.683134   0.907327   0.753  0.45213
month.Q         0.735963   0.916250   0.803  0.42251
month.C         2.978612   0.907329   3.283  0.00116 **
month^4        -0.062377   0.905578  -0.069  0.94513
month^5         1.779356   0.909297   1.957  0.05134 .
```

month <sup>6</sup>	1.119657	0.907425	1.234	0.21826
month <sup>7</sup>	-2.233488	0.908183	-2.459	0.01452 *
month <sup>8</sup>	2.052115	0.907428	2.261	0.02448 *
month <sup>9</sup>	-1.587479	0.908290	-1.748	0.08158 .
month <sup>10</sup>	0.534394	0.907461	0.589	0.55640
month <sup>11</sup>	-0.347802	0.908496	-0.383	0.70213
transition	1.363058	0.680451	2.003	0.04611 *
treatment:distance	-1.010874	0.878174	-1.151	0.25065

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 4.537 on 285 degrees of freedom

Multiple R-squared: 0.3422, Adjusted R-squared: 0.3076

F-statistic: 9.883 on 15 and 285 DF, p-value: < 2.2e-16

Call:

```
lm(formula = guidance ~ treatment * distance + month + transition,
    data = rd_data_all)
```

Residuals:

Min	1Q	Median	3Q	Max
-9.650	-2.611	-0.677	1.718	101.692

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	8.343186	0.853611	9.774	< 2e-16 ***
treatment	4.185854	6.252975	0.669	0.50377
distance	0.017403	0.004867	3.576	0.00041 ***
month.L	3.057850	1.414503	2.162	0.03147 *
month.Q	0.175262	1.428414	0.123	0.90243
month.C	-0.105815	1.414507	-0.075	0.94042
month <sup>4</sup>	-1.794588	1.411776	-1.271	0.20471
month <sup>5</sup>	-1.841802	1.417575	-1.299	0.19490
month <sup>6</sup>	-0.923469	1.414656	-0.653	0.51442
month <sup>7</sup>	-1.902585	1.415838	-1.344	0.18009
month <sup>8</sup>	0.233432	1.414661	0.165	0.86905
month <sup>9</sup>	-1.385906	1.416005	-0.979	0.32854
month <sup>10</sup>	-0.474372	1.414712	-0.335	0.73764
month <sup>11</sup>	-0.518115	1.416326	-0.366	0.71477
transition	-0.370870	1.060809	-0.350	0.72689
treatment:distance	-0.317373	1.369054	-0.232	0.81684

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 7.074 on 285 degrees of freedom

Multiple R-squared: 0.0924, Adjusted R-squared: 0.04463

F-statistic: 1.934 on 15 and 285 DF, p-value: 0.02003

Call:

```
lm(formula = total_documents ~ treatment * distance + month +  
    transition, data = rd_data_all)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-14.250	-4.221	-1.040	3.306	106.356

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	11.868067	1.068392	11.108	<2e-16 ***
treatment	9.808287	7.826317	1.253	0.2111
distance	-0.015343	0.006091	-2.519	0.0123 *
month.L	3.740985	1.770413	2.113	0.0355 *
month.Q	0.911225	1.787824	0.510	0.6107
month.C	2.872797	1.770418	1.623	0.1058
month^4	-1.856965	1.767001	-1.051	0.2942
month^5	-0.062445	1.774258	-0.035	0.9719
month^6	0.196187	1.770605	0.111	0.9119
month^7	-4.136074	1.772084	-2.334	0.0203 *
month^8	2.285547	1.770611	1.291	0.1978
month^9	-2.973385	1.772293	-1.678	0.0945 .
month^10	0.060021	1.770674	0.034	0.9730
month^11	-0.865917	1.772695	-0.488	0.6256
transition	0.992188	1.327725	0.747	0.4555
treatment:distance	-1.328247	1.713529	-0.775	0.4389

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

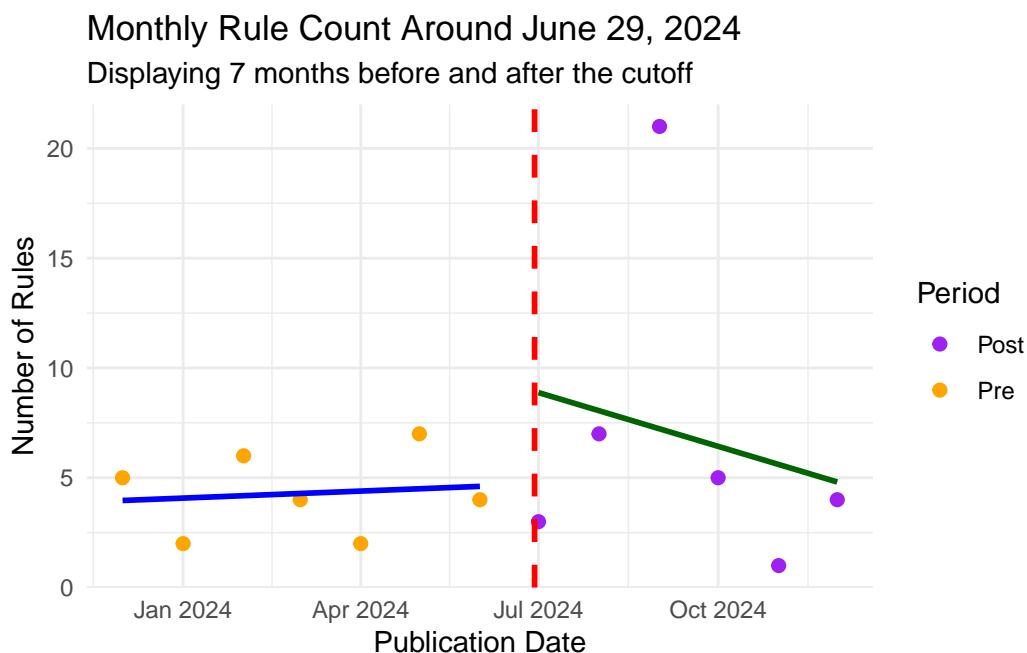
Residual standard error: 8.853 on 285 degrees of freedom

Multiple R-squared: 0.08617, Adjusted R-squared: 0.03807

F-statistic: 1.792 on 15 and 285 DF, p-value: 0.03539

Now, let's visualize the RD regression, displaying the number of rules published in the 7 months before and after the cutoff date

```
`geom_smooth()` using formula = 'y ~ x'
`geom_smooth()` using formula = 'y ~ x'
```



Now we'll try an ITS model.

```
Series: ts_rules
Regression with ARIMA(1,1,1)(0,0,2)[12] errors
```

Coefficients:

	ar1	ma1	sma1	sma2	Intervention	TimeAfterIntervention
	-0.0250	-0.7833	0.0634	0.1128	5.9283	-1.3589
s.e.	0.0677	0.0442	0.0522	0.0519	3.7620	0.9109

```
sigma^2 = 23.73: log likelihood = -1114.53
```

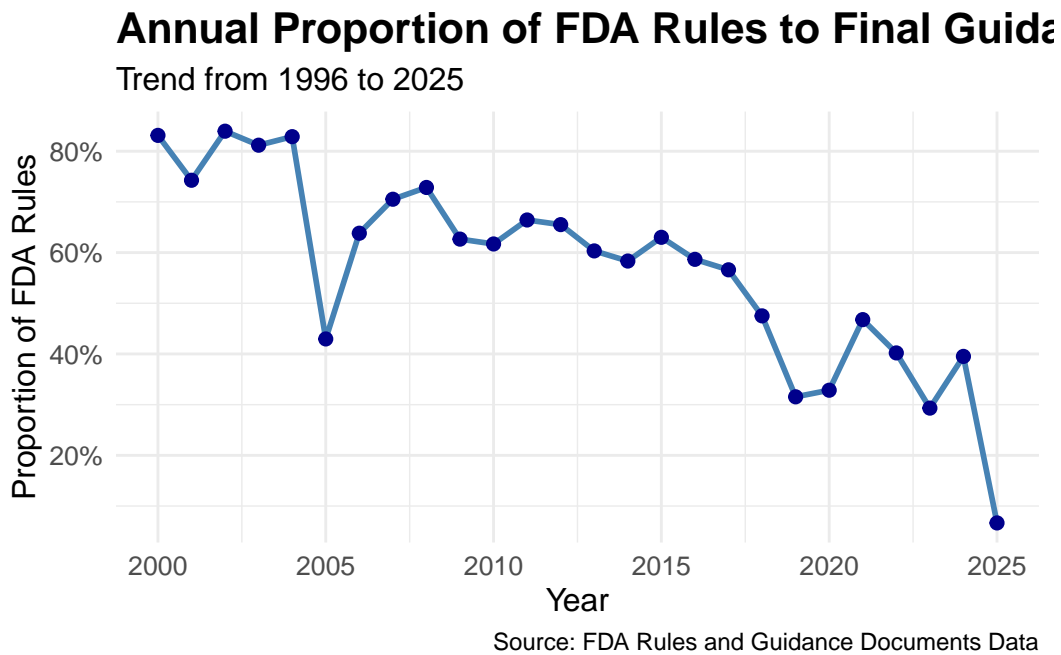
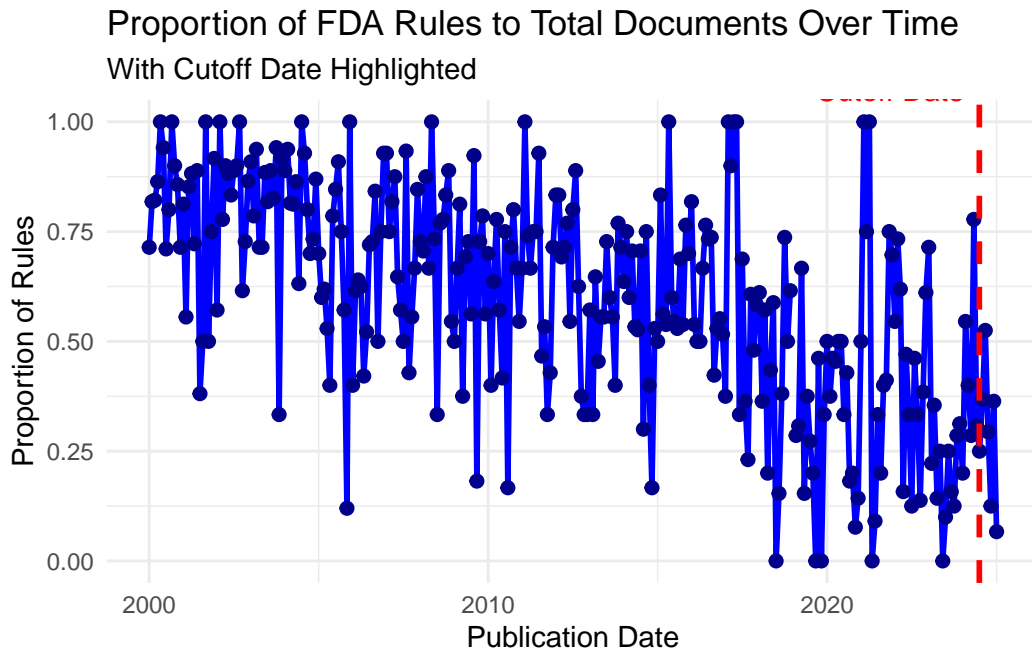
```
AIC=2243.05 AICc=2243.36 BIC=2270.48
```

Training set error measures:

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1
Training set	0.002791429	4.825765	3.629884	-Inf	Inf	0.6842757	0.0003992911

```
Warning: Removed 1 row containing missing values or values outside the scale range
(`geom_point()`).
```





RD model for proportion of guidance docs to total documents with only 6 mos before and after cutoff

Call:

```
lm(formula = proportion_rules ~ treatment * distance, data = rd_data_filtered)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.19106	-0.14332	-0.01185	0.09594	0.31151

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.33805	0.14179	2.384	0.0443 *
treatment	0.07867	0.42529	0.185	0.8578
distance	0.03153	0.04582	0.688	0.5108
treatment:distance	-0.04268	0.06480	-0.659	0.5286

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1915 on 8 degrees of freedom

Multiple R-squared: 0.1421, Adjusted R-squared: -0.1797

F-statistic: 0.4416 on 3 and 8 DF, p-value: 0.7297

RD model for proportion of guidance docs to total documents with all data (does not include Jan 2025).

Call:

```
lm(formula = proportion_rules ~ treatment * distance + month +  
    transition, data = rd_data_proportion_filtered)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.53664	-0.12497	0.01234	0.11296	0.56897

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.3458669	0.0238869	14.479	<2e-16 ***
treatment	0.0268312	0.1508588	0.178	0.8590
distance	-0.0017115	0.0001353	-12.654	<2e-16 ***
month.L	-0.0930523	0.0398610	-2.334	0.0203 *
month.Q	0.0543076	0.0398338	1.363	0.1739
month.C	0.0843200	0.0396159	2.128	0.0342 *
month^4	0.0126976	0.0393752	0.322	0.7473
month^5	0.0582927	0.0393846	1.480	0.1400

month <sup>6</sup>	0.0344916	0.0392709	0.878	0.3805
month <sup>7</sup>	0.0104279	0.0393028	0.265	0.7910
month <sup>8</sup>	-0.0117084	0.0392479	-0.298	0.7657
month <sup>9</sup>	-0.0159263	0.0392855	-0.405	0.6855
month <sup>10</sup>	0.0262113	0.0392457	0.668	0.5048
month <sup>11</sup>	-0.0141505	0.0392973	-0.360	0.7190
transition	0.0224673	0.0296287	0.758	0.4489
treatment:distance	-0.0161162	0.0478745	-0.337	0.7366

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1962 on 283 degrees of freedom

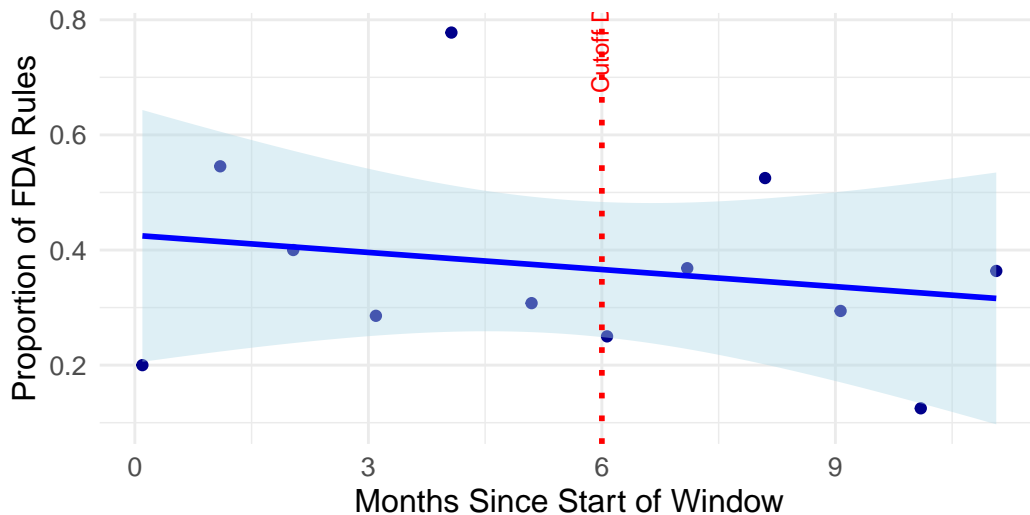
(1 observation deleted due to missingness)

Multiple R-squared: 0.4005, Adjusted R-squared: 0.3687

F-statistic: 12.6 on 15 and 283 DF, p-value: < 2.2e-16

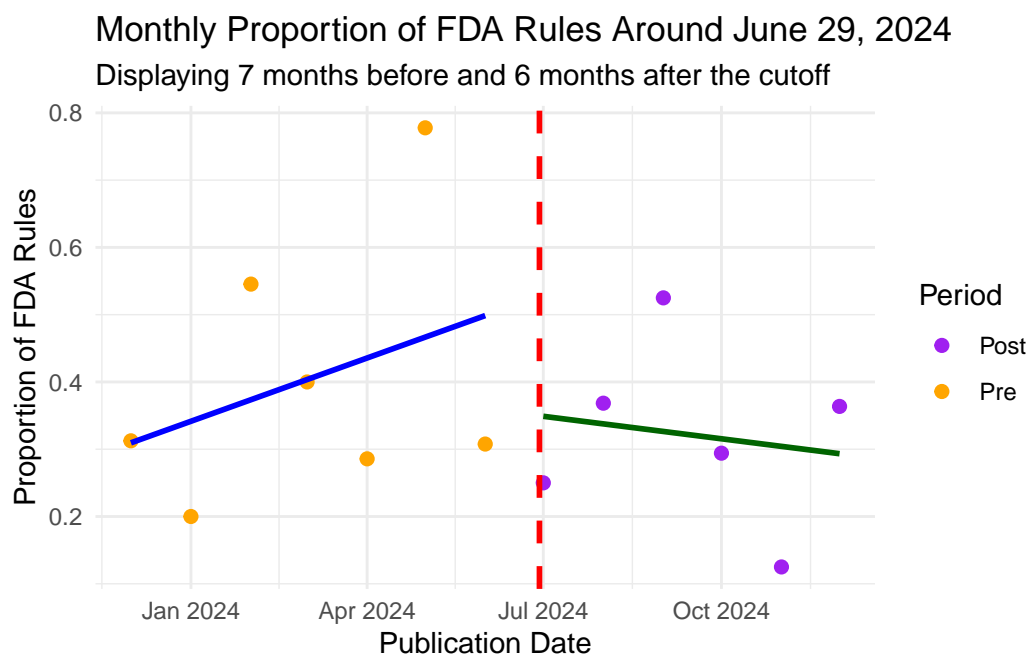
## Regression Discontinuity Analysis Within 6-Mc

Proportion of FDA Rules to Final Guidance Documents



Source: FDA Rules and Guidance Documents Data

```
`geom_smooth()` using formula = 'y ~ x'
`geom_smooth()` using formula = 'y ~ x'
```



Call:

```
lm(formula = proportion_rules ~ time + intervention + time_after,
    data = combined_counts)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.61351	-0.12564	0.01424	0.12761	0.58256

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.8531517	0.0230012	37.092	<2e-16 ***
time	-0.0017089	0.0001361	-12.560	<2e-16 ***
intervention	0.0390056	0.1390001	0.281	0.779
time_after	-0.0323138	0.0371518	-0.870	0.385

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.1977 on 296 degrees of freedom

(1 observation deleted due to missingness)

Multiple R-squared: 0.3734, Adjusted R-squared: 0.367

F-statistic: 58.79 on 3 and 296 DF, p-value: < 2.2e-16