Negative Binomial Regression Fit Report

20250610

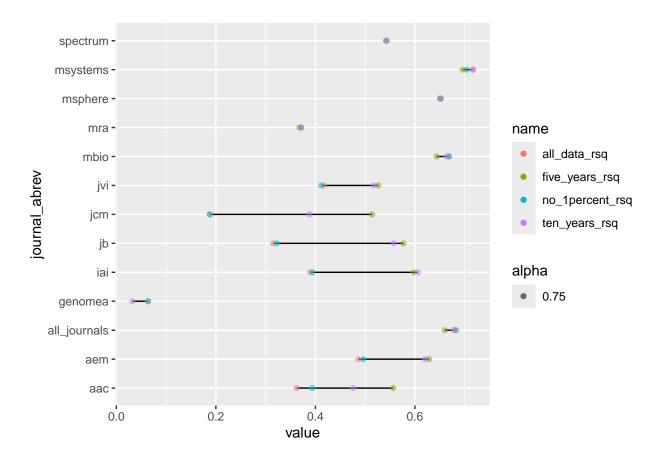
Project Summary

- We are using data from the American Society of Microbiology's (ASM) 12 published journals to investigate the relationship between the number of citations (variable 'is.referenced.by.count') a published scientific article receives and if the authors have included access to their raw sequencing data (variable 'da', data availability) in the manuscript.
- We are trying to understand if publishing raw data helps to improve citation metrics. We have data from 2000-2024, and will also adjust for time published (variable 'age.in.months'), as older papers have had the opportunity to accumulate more citations over time.

knitr::kable(rsquared, digits = 4)

journal_abrev	n	all_data_rsq	no_1percent_rsq	five_years_rsq	ten_years_rsq
aac	3237	0.3623	0.3938	0.5564	0.4759
aem	8638	0.4862	0.4964	0.6284	0.6204
genomea	6578	0.0636	0.0636	NA	0.0321
iai	1854	0.3896	0.3943	0.5975	0.6062
jb	4867	0.3152	0.3221	0.5767	0.5575
jcm	4374	0.1882	0.1878	0.5139	0.3880
jvi	4583	0.4172	0.4115	0.5264	0.5167
mbio	2498	0.6680	0.6685	0.6438	0.6633
mra	5738	0.3712	0.3712	0.3680	0.3712
msphere	1041	0.6523	0.6508	0.6510	0.6523
msystems	1436	0.7168	0.7040	0.6962	0.7168
spectrum	2957	0.5425	0.5425	0.5425	0.5425
all_journals	47808	0.6781	0.6829	0.6602	0.6801

```
rsquared %>%
pivot_longer(cols = all_data_rsq:ten_years_rsq) %>%
ggplot(aes(y = journal_abrev, x = value)) +
geom_line(na.rm = TRUE) +
geom_point(aes(color = name, alpha = 0.75), na.rm = TRUE)
```



How well do the models fit (by Cragg-Uhler pseduo R-squared metric)

• See above table "rsquared"

- Model format for all data from all journals MASS::glm.nb(is.referenced.by.count~ da_factor + log(age.in.months) + container.title + container.title*da_factor + log(age.in.months)*da_factor + container.title*log(age.in.months) + log(age.in.months)*da_factor*container.title, data = nsd_yes_metadata, link = log)
- Use model format for data from each journal MASS::glm.nb(is.referenced.by.count~da_factor + log(age.in.months) + log(age.in.months)*da_factor, data = <each journal>, link = log)

• Overall model fit with all data from all journals:

- $R^2 \text{ value} = 0.678$
- Removal of top 1% of data: R^2 value = 0.682
- Truncate data to last 5 years: R^2 value = 0.660
- Truncate data to last 10 years: R^2 value = 0.680
- Summary: Model fit by R^2 metric does not change by removing the top 1% of data or truncating to data from the last 5 or 10 years.

• Overall model fit for data from EACH journal individually:

- -4/12 journals have **overall model fit** with $R^2 > 0.5$
- -4/12 journals have fit with $R^2 > 0.5$ with top 1% of data removed
- 10/11 journals have model fits >0.5 when **truncated to the last 5 years**, so they are better than their fit overall (one journal has no data from this period)
- -8/12 journals have model fits >0.5 when **truncated to the last 10 years**, so they are better than their fit overall
- Summary: Data fits negative binomial model better with only more recent data considered.

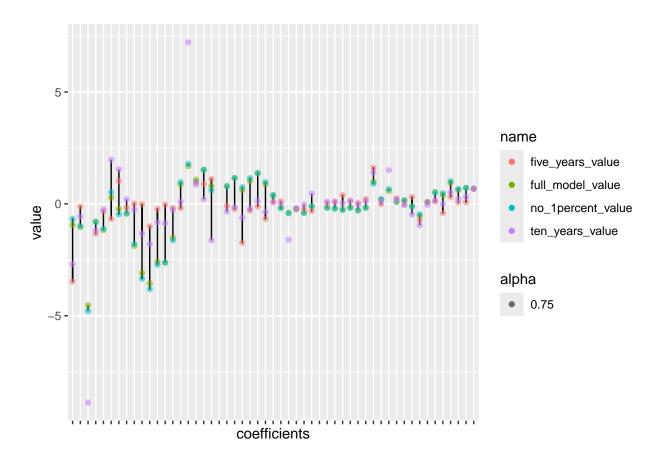
knitr::kable(all_journals, digits = 4)

coefficients	full_model_valueno	_1percent_valuativ	ve_years_value te	en_years_value
rsquared	0.6781	0.6829	0.6602	0.6801
(Intercept)	-0.9460	-0.6722	-3.4726	-2.6960
da_factorYes	-1.5082	-1.6296	-0.1780	-0.2667
log(age.in.months)	0.9932	0.9069	1.6141	1.3976
container.titleApplied and Environmental Microbiology	-1.0337	-0.9933	-0.1207	-0.5651
container.titleGenome Announcements	-4.5303	-4.7923	NA	-8.8809
container.titleInfection and Immunity	-0.7927	-0.8150	-1.3290	-1.1797
container.titleJournal of Bacteriology	-1.1706	-1.1146	-0.3361	-0.2502
container.titleJournal of Clinical Microbiology	0.2729	0.5180	-0.6776	1.9759
container.titleJournal of Microbiology & Biology Education	-0.2261	-0.4692	1.0220	1.5424

coefficients	full_model_valuen	o_1percent_valufi	ve_years_value te	en_years_valu
container.titleJournal of Virology	-0.4370	-0.4275	-0.1728	0.2013
container.titlemBio	-1.8817	-1.7910	0.0115	-0.2660
container.titleMicrobiology Resource Announcements	-3.0840	-3.3426	-0.0170	-1.3247
container.titleMicrobiology Spectrum	-3.5550	-3.8058	-1.0015	-1.7913
container.titlemSphere	-2.5767	-2.7184	-0.2216	-0.8267
container.titlemSystems	-2.6293	-2.6203	-0.0287	-0.8701
la_factorYes:container.titleApplied and Environmental Microbiology	0.8693	0.9714	-0.1907	0.1001
la_factorYes:container.titleGenome Announcements	1.6881	1.7938	NA	7.2193
la_factorYes:container.titleInfection and Immunity	1.0848	0.9853	0.9639	0.8601
la_factorYes:container.titleJournal of Bacteriology	1.5260	1.5178	0.8850	0.2051
la_factorYes:container.titleJournal of Clinical Microbiology	0.7994	0.6000	1.1103	-1.6342
la_factorYes:container.titleJournal of Microbiology & Biology Education	NA	NA	NA	NA
la_factorYes:container.titleJournal of Virology	0.8019	0.7830	-0.0860	-0.3469
la_factorYes:container.titlemBio	1.1512	1.1542	-0.2339	-0.1494
la_factorYes:container.titleMicrobiology Resource Announcements	0.6208	0.7466	-1.7384	-0.6183
la_factorYes:container.titleMicrobiology Spectrum	1.0273	1.1494	-0.3022	-0.2138
la_factorYes:container.titlemSphere	1.3657	1.3770	-0.1163	0.1374
la_factorYes:container.titlemSystems	0.8815	0.9763	-0.6710	-0.3577
la_factorYes:log(age.in.months)	0.3675	0.3986	0.0661	0.0957
og(age.in.months):container.titleApplied and Environmental Microbiology	0.2058	0.2073	-0.0156	0.1092
og(age.in.months):container.titleGenome Announcements	0.5778	0.6617	NA	1.5084
og(age.in.months):container.titleInfection and Immunity	0.0813	0.1089	0.2478	0.1951
og(age.in.months):container.titleJournal of Bacteriology	0.1570	0.1612	-0.0098	-0.0566
og(age.in.months):container.titleJournal of Clinical Microbiology	-0.0931	-0.1295	0.3100	-0.4809
og(age.in.months):container.titleJournal of Microbiology & Biology Education	-0.5427	-0.4649	-0.7173	-0.9522
og(age.in.months):container.titleJournal of Virology	0.0693	0.0849	0.0714	-0.0526
og(age.in.months):container.titlemBio	0.5267	0.4983	0.0984	0.1610
og(age.in.months):container.titleMicrobiology Resource Announcements	0.3927	0.4751	-0.4083	-0.0141
og(age.in.months):container.titleMicrobiology Spectrum	0.9389	1.0180	0.3094	0.5302
og(age.in.months):container.titlemSphere	0.6216	0.6690	0.0632	0.2172
og(age.in.months):container.titlemSystems	0.7187	0.7154	0.0735	0.3119
a_factorYes:log(age.in.months):container.titleApplied and Environmental	-0.1698	-0.1948	0.1092	0.0088
Microbiology				
la_factorYes:log(age.in.months):container.titleGenome Announcements	-0.3920	-0.4198	NA	-1.6043
la_factorYes:log(age.in.months):container.titleInfection and Immunity	-0.2392	-0.2182	-0.2194	-0.1933
la_factorYes:log(age.in.months):container.titleJournal of Bacteriology	-0.4067	-0.4056	-0.2304	-0.0223
la factorYes:log(age.in.months):container.titleJournal of Clinical Microbiology	-0.1115	-0.0999	-0.3246	0.4618

coefficients	full_model_valueno	_1percent_valufiv	e_years_value te	en_years_value
da_factorYes:log(age.in.months):container.titleJournal of Microbiology & Biology	NA	NA	NA	NA
Education da factorYes:log(age.in.months):container.titleJournal of Virology	-0.1740	-0.1730	0.0329	0.0975
da_factorYes:log(age.in.months):container.titlemBio	-0.2141	-0.2046	0.1095	0.0758
da_factorYes:log(age.in.months):container.titleMicrobiology Resource	-0.2456	-0.2778	0.3859	0.0256
Announcements				
da_factorYes:log(age.in.months):container.titleMicrobiology Spectrum	-0.1533	-0.1846	0.1478	0.1183
da_factorYes:log(age.in.months):container.titlemSphere	-0.2926	-0.2903	0.0498	-0.0242
da_factorYes:log(age.in.months):container.titlemSystems	-0.1622	-0.1815	0.2131	0.1090

```
all_journals %>%
  pivot_longer(cols = full_model_value:ten_years_value) %>%
  ggplot(aes(x = coefficients, y = value)) +
  geom_line(na.rm = TRUE) +
  geom_point(aes(color = name, alpha = 0.75), na.rm = TRUE) + theme(axis.text.x = element_blank())
```



All journal model is resistant to changes from removing top 1% of data, but less resistant to changes from truncating at 5 and 10 years.

- When working across the columns in the second table, we have coefficients on the left, followed by their values under the following conditions
 - full _model_value = all data included in the model
 - -no_1percent_value = top 1% of data removed
 - five _years_value = data truncated at 5 years in age of paper
 - -ten_years_value = data truncated at 10 years in age of paper

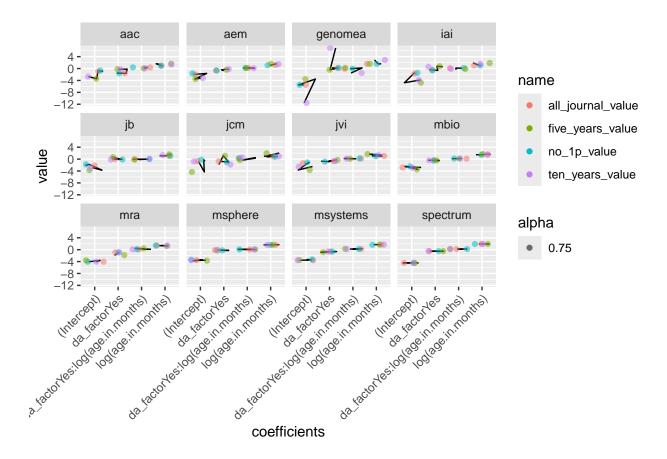
- **Note:** Journal of Microbiology and Biology Education(jmbe) has N=7 papers with new sequence data and has been excluded for these analyses, but is a part of the model, and appears as NAs in the table above.
- Is this graphic helpful? The labels are super cluttered and need some work.

knitr::kable(each_journal, digits = 4)

aac (Intercept) -0.9609 -0.6665 -3.4423 aac da_factorYes -1.5204 -1.6244 -0.1890 aac log(age.in.months) 0.9961 0.9057 1.6058 aac da_factorYes:log(age.in.months) 0.3708 0.3972 0.0691 aem (Intercept) -1.9654 -1.6400 -3.5278 aem da_factorYes -0.6343 -0.6462 -0.4172 aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea dog(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050	ten_years_value	five_years_value	no_1p_value	all_journal_value	coefficients	name
aac log(age.in.months) 0.9961 0.9057 1.6058 aac da_factorYes:log(age.in.months) 0.3708 0.3972 0.0691 aem (Intercept) -1.9654 -1.6400 -3.5278 aem da_factorYes -0.6343 -0.6462 -0.4172 aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea dog(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 -0.4172 genomea da_factorYes:log(age.in.months) 1.0675 1.0909 1.8523 iai da_factorYes:log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.0163 -0.1174 0.6339 jb	-2.6960	-3.4423	-0.6665	-0.9609	(Intercept)	aac
aac da_factorYes:log(age.in.months) 0.3708 0.3972 0.0691 aem (Intercept) -1.9654 -1.6400 -3.5278 aem da_factorYes -0.6343 -0.6462 -0.4172 aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes:log(age.in.months)	-0.2667	-0.1890	-1.6244	-1.5204	$da_factorYes$	aac
aem (Intercept) -1.9654 -1.6400 -3.5278 aem da_factorYes -0.6343 -0.6462 -0.4172 aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb da_factorYes:log(age.in.months)	1.3976	1.6058	0.9057	0.9961	$\log(\text{age.in.months})$	aac
aem da_factorYes -0.6343 -0.6462 -0.4172 aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) 1.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb da_factorYes:log(age.in.months) -0.0	0.0957	0.0691	0.3972	0.3708	$da_factor Yes: log(age.in.months)$	aac
aem log(age.in.months) 1.1963 1.1092 1.5802 aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb da_factorYes:log(age.in.months) 1.1499 1.0669 1.5749 jcm da_factorYes -0.081	-3.2253	-3.5278	-1.6400	-1.9654	(Intercept)	aem
aem da_factorYes:log(age.in.months) 0.1963 0.2004 0.1888 genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb da_factorYes:log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm da_factorYes	-0.1718	-0.4172	-0.6462	-0.6343	$da_factorYes$	aem
genomea (Intercept) -5.4902 -5.4902 -3.5278 genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months)	1.4985	1.5802	1.1092	1.1963	log(age.in.months)	aem
genomea da_factorYes 0.1984 0.1984 -0.4172 genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm da_factorYes:log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months)	0.1054	0.1888	0.2004	0.1963	$da_factorYes:log(age.in.months)$	aem
genomea log(age.in.months) 1.5740 1.5740 1.5802 genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm da_factorYes:log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months)	-11.5225	-3.5278	-5.4902	-5.4902	(Intercept)	genomea
genomea da_factorYes:log(age.in.months) -0.0284 -0.0284 0.1888 iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	6.9105	-0.4172	0.1984	0.1984	$da_factorYes$	genomea
iai (Intercept) -1.7012 -1.4506 -4.7673 iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	2.8942	1.5802	1.5740	1.5740	$\log(\text{age.in.months})$	genomea
iai da_factorYes -0.4050 -0.6297 0.7774 iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-1.4995	0.1888	-0.0284	-0.0284	$da_factorYes:log(age.in.months)$	genomea
iai log(age.in.months) 1.0675 1.0090 1.8523 iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-3.8227	-4.7673	-1.4506	-1.7012	(Intercept)	iai
iai da_factorYes:log(age.in.months) 0.1231 0.1761 -0.1510 jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	0.6081	0.7774	-0.6297	-0.4050	$da_factorYes$	iai
jb (Intercept) -2.1151 -1.7806 -3.7023 jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	1.5803	1.8523	1.0090	1.0675	$\log(\text{age.in.months})$	
jb da_factorYes 0.0163 -0.1174 0.6339 jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-0.1015	-0.1510	0.1761	0.1231	$da_factorYes:log(age.in.months)$	
jb log(age.in.months) 1.1499 1.0669 1.5749 jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-2.8669	-3.7023	-1.7806	-2.1151	(Intercept)	jb
jb da_factorYes:log(age.in.months) -0.0389 -0.0060 -0.1444 jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-0.1168	0.6339	-0.1174	0.0163	$da_factorYes$	
jcm (Intercept) -0.6815 -0.1575 -4.3089 jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	1.3225				$\log(\text{age.in.months})$	
jcm da_factorYes -0.7281 -1.0364 1.1310 jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	0.0859		-0.0060		$da_factorYes:log(age.in.months)$	jb
jcm log(age.in.months) 0.9018 0.7780 1.9686 jcm da_factorYes:log(age.in.months) 0.2605 0.3003 -0.3140	-0.7888		-0.1575		` /	$_{\rm jcm}$
jcm $da_{factorYes:log(age.in.months)}$ 0.2605 0.3003 -0.3140	-1.8444				$da_factorYes$	jcm
	0.9325	1.9686	0.7780		$\log(\text{age.in.months})$	$_{\rm jcm}$
	0.5445				$da_factorYes:log(age.in.months)$	•
jvi (Intercept) -1.3855 -1.1081 -3.6656	-2.4940	-3.6656	-1.1081	-1.3855	(Intercept)	jvi
jvi da_factorYes -0.7088 -0.8540 -0.2909	-0.6135	-0.2909	-0.8540	-0.7088	$da_factorYes$	jvi
jvi $log(age.in.months)$ 1.0630 0.9935 1.6911	1.3449	1.6911	0.9935	1.0630	$\log(\text{age.in.months})$	
jvi da_factorYes: $log(age.in.months)$ 0.1941 0.2275 0.1068	0.1932	0.1068	0.2275	0.1941	$da_factorYes:log(age.in.months)$	jvi
mbio (Intercept) -2.8112 -2.4410 -3.4651	-2.9554	-3.4651	-2.4410	-2.8112	(Intercept)	mbio

name	coefficients	all_journal_value	no_1p_value	five_years_value	ten_years_value
mbio	da_factorYes	-0.3597	-0.4783	-0.4089	-0.4184
mbio	log(age.in.months)	1.5161	1.4000	1.7137	1.5569
mbio	da_factorYes:log(age.in.months)	0.1539	0.1944	0.1747	0.1720
mra	(Intercept)	-4.0497	-4.0497	-3.5995	-4.0497
mra	da_factorYes	-0.8906	-0.8906	-1.8301	-0.8906
mra	log(age.in.months)	1.3911	1.3911	1.2378	1.3911
mra	da_factorYes:log(age.in.months)	0.1228	0.1228	0.4267	0.1228
msphere	(Intercept)	-3.5233	-3.3874	-3.7000	-3.5233
msphere	da_factorYes	-0.1065	-0.2401	-0.2685	-0.1065
msphere	log(age.in.months)	1.6149	1.5751	1.6790	1.6149
msphere	da_factorYes:log(age.in.months)	0.0657	0.1050	0.1088	0.0657
msystems	(Intercept)	-3.5575	-3.2742	-3.4729	-3.5575
msystems	da_factorYes	-0.6220	-0.6507	-0.8537	-0.6220
msystems	log(age.in.months)	1.7072	1.6174	1.6794	1.7072
msystems	da_factorYes:log(age.in.months)	0.2040	0.2163	0.2807	0.2040
spectrum	(Intercept)	-4.4608	-4.4608	-4.4608	-4.4608
spectrum	da_factorYes	-0.4802	-0.4802	-0.4802	-0.4802
spectrum	log(age.in.months)	1.9194	1.9194	1.9194	1.9194
spectrum	da_factorYes:log(age.in.months)	0.2139	0.2139	0.2139	0.2139

```
each_journal %>%
    rename(journal_abrev = name) %>%
    pivot_longer(cols = all_journal_value:ten_years_value) %>%
    ggplot(aes(x = coefficients, y = value)) +
    geom_line(na.rm = TRUE, position = "jitter") +
    geom_point(aes(color = name, alpha = 0.75), na.rm = TRUE, position = "jitter") +
    facet_wrap(vars(journal_abrev)) +
    theme(axis.text.x = element_text(angle = 45, hjust=1))
```



Each journal model are semi-resistant to changes from removing top 1% of data, and even less resistant to changes from truncating at 5 and 10 years.

- See above for mutations on these columns, but these models look less resistant to the transformation of removing the top 1% of data, and even less resistant to changes in coefficients from truncating at 5 and 10 years of data.
- Is this graphic helpful? I think the scale should be adjusted to better show the data.