

# Vaccination Rates & New Case Counts

University of California, Berkeley  
4/13/2021

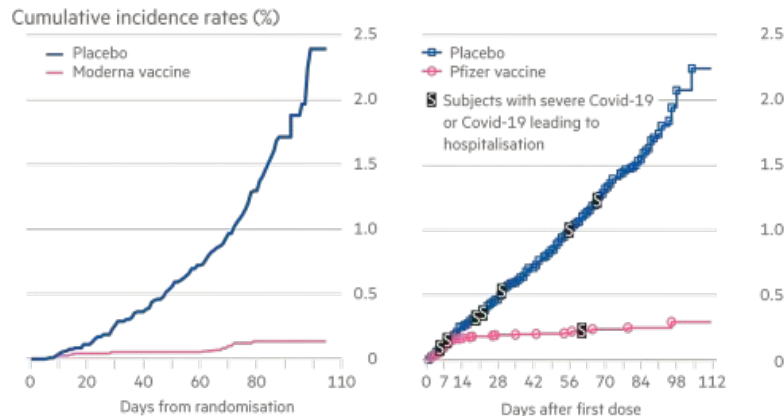
Dera C., Pavan E., Greg T., Kayla W.



# Research Question

Are vaccinations causing a decrease in new COVID-19 cases?

Covid-19 cases in the placebo group overtake the vaccine group soon after first dose



Sources: FDA; Pfizer/BioNTech  
© FT

We see from controlled trials that these vaccines prevent cases.

What do we see in the real world?

Other factors driving cases:

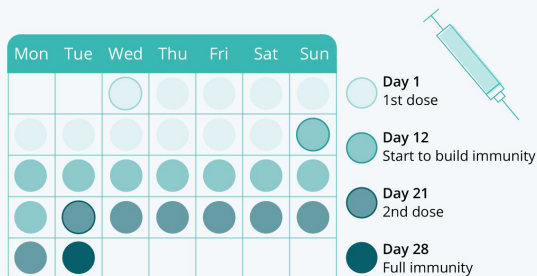
- “Risky Behaviors”
- Not social distancing
- Not adhering to wearing masks
- Creating space for viral spread

# Data & Preparation

## 2+ Week Lag for Efficacy

### The Pfizer/BioNTech Vaccination Process

Vaccination process for the Pfizer/BioNTech Covid-19 BNT162b2 vaccine



Source: Pfizer/BioNTech via BBC



statista

## CDC Data on New Covid Cases

Aggregated weekly and presented as new cases per 100k

## OWID Data on Vaccinations

Reported daily and presented as people per 100, or % vaccinated (sourced from CDC)

## Policy Actions by KFF

Taken for the last week in March 2021

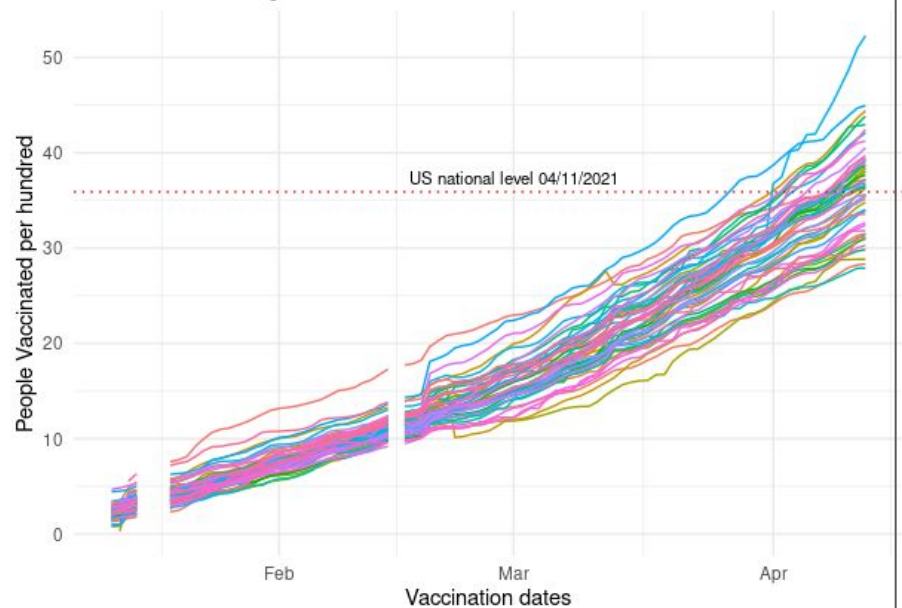
## Community Movement from Google

Logged as percent change from baseline for given community. Categorized by High, Med, Low based on quartiles.

# EDA of % Population Vaccinated

## % Vaccinated is rising!

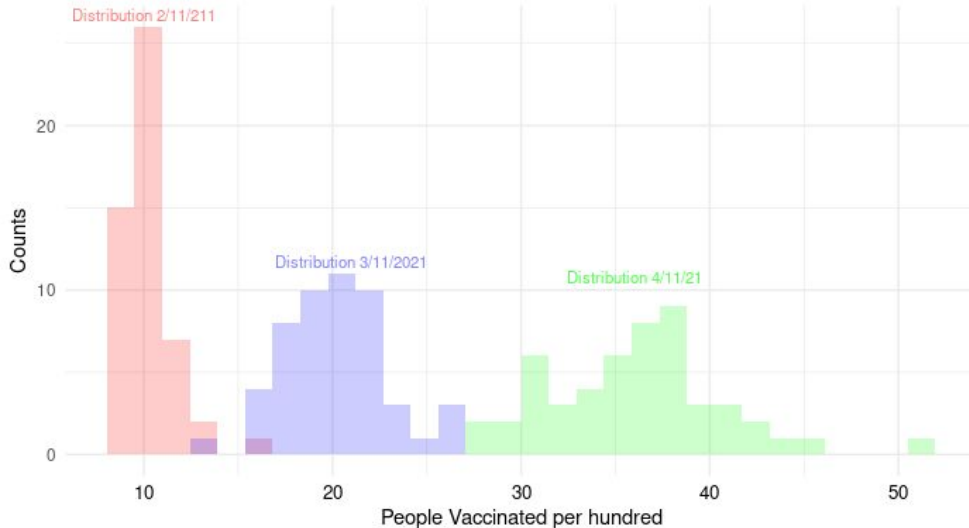
Vaccinations are rising!



## Distribution becomes more normal over time

% Vaccinated distribution appears more normal over time

People Vacc per hundred on dates: Feb 11 (red), Mar 11 (blue), Apr 11 (green)



# Base Model

$$f(\text{Log}(\text{Count New Cases})) = \beta_0 + \beta_1(\% \text{ Population Vaccinated})$$

## T test

t test of coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	3.690186	0.556829	6.6271	2.509e-08 ***
people_vaccinated_per_hundred	0.054373	0.028701	1.8945	0.06407 .

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

% Population  
Vaccinated is  
**not statistically  
significant**

# Improved Model

## - Add population density & “risky behavior” variables

$$f(\text{Log}(\text{Count New Cases})) = \beta_0 + \beta_1(\% \text{ Population Vaccinated}) + \beta_2(\log(\text{Density})) + \beta_3(\text{Workplace Movement}) + \beta_4(\text{Retail Movement}) + \beta_5(\text{Transit Movement})$$

### T test

t test of coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	2.665557	0.635621	4.1936	0.000127 ***
people_vaccinated_per_hundred	0.068029	0.020325	3.3471	0.001657 **
log(Density)	0.165464	0.050027	3.3075	0.001857 **
Workplace_flag	0.391674	0.157921	2.4802	0.016939 *
Retail_flag	-0.168079	0.200767	-0.8372	0.406913
Transit_flag	-0.236087	0.142132	-1.6610	0.103658
---				
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1				

% Population Vaccinated, Density and Workplace movement are **statistically significant**

# The Kitchen Sink Model

## - Add more “risky behavior” and policy mandates

$$f(\text{Log}(\text{Count New Cases})) = \beta_0 + \beta_1(\% \text{ Population Vaccinated}) + \beta_2(\text{log}(\text{Density})) + \beta_3(\text{Workplace Movement}) + \beta_4(\text{Retail Movement}) +$$

$$\beta_5(\text{Transit Movement}) + \beta_6(\text{Business Closure Policy}) + \beta_7(\text{Bar Closures}) + \beta_8(\text{Mask Policy}) + \beta_9(\text{Parks Movement}) + \beta_{10}(\text{Grocery Movement})$$

t test of coefficients:

T test

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	2.556280	0.904969	2.8247	0.007346 **
people_vaccinated_per_hundred	0.075011	0.028445	2.6371	0.011851 *
log(Density)	0.147084	0.058204	2.5270	0.015559 *
Workplace_flag	0.283076	0.179545	1.5766	0.122758
Retail_flag	-0.213018	0.221253	-0.9628	0.341443
Transit_flag	-0.343816	0.174081	-1.9750	0.055194 .
Business_Flag	0.347654	0.290626	1.1962	0.238652
Bar_Flag	-0.232790	0.217567	-1.0700	0.291048
Mask_Flag	-0.022192	0.284057	-0.0781	0.938118
Parks_flag	0.088718	0.130827	0.6781	0.501594
Grocery_flag	0.090847	0.178501	0.5089	0.613585

---

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

% Population  
Vaccinated and  
Density are  
**statistically  
significant**

# Model Comparisons

Improved Model is the best performing model:

- As seen in Regression Table, all variables have have **lowest Standard error**
- F-Tests show that the **Improved model is better performing than Base Model**
  - No evidence that “Kitchen sink” is better

Table 1: The relationship between new covid cases and % of population vaccinated

	Dependent variable:		
	log(cumulative_new_case_7_per100000) (1)	(2)	(3)
people_vaccinated_per_hundred	0.054 (0.028)	0.068* (0.027)	0.075* (0.029)
log(Density)		0.165** (0.060)	0.147* (0.061)
Workplace_flag		0.392** (0.135)	0.283 (0.149)
Retail_flag		-0.168 (0.149)	-0.213 (0.174)
Transit_flag		-0.236 (0.143)	-0.344 (0.170)
Bussiness_flag			0.348 (0.216)
Bar_Flag			-0.233 (0.157)
Mask_Flag			-0.022 (0.225)
Parks_flag			0.089 (0.109)
Grocery_flag			0.091 (0.157)
Constant	3.690*** (0.905)	2.666*** (0.754)	2.556** (0.828)
Observations	51	51	51
R2	0.069	0.404	0.475
Adjusted R2	0.050	0.338	0.344
Residual Std. Error	0.547 (df = 49)	0.457 (df = 45)	0.454 (df = 40)

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001



# Conclusions

- Improved Model (the best model) indicates that:

**1 more person out a hundred with a vaccine is associated with a 6.8% percent increase in Cumulative Cases per 100k in Last 7 Days** (given all else equal)

- Our original hypothesis was wrong** (with our dataset)
- This is surprising, **but not novel!**

Could be explained by the **Peltzman Effect**

## The Effects of Automobile Safety Regulation

Sam Peltzman  
*University of Chicago*

Gives  
the feeling  
of invincibility



# Next Steps - Address major model limitations

- Test Peltzman Effect
- Reduce Scale from States to Counties (increase sample size)
- Find proxies for Omitted Variables:

○

Face Mask  
Adherence

Prevalence of  
Covid Variants per  
state

Age

Returning to  
school

Social gatherings

# Questions?

---

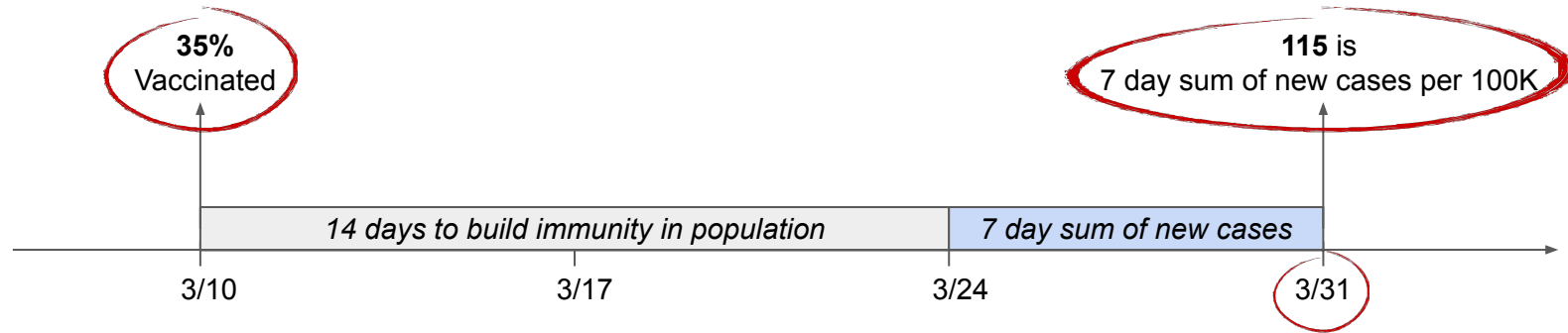
University of California, Berkeley  
4/13/2021

Dera C., Pavan E., Greg T., Kayla W.



# Appendix - Need time to build Immunity and Count Cases

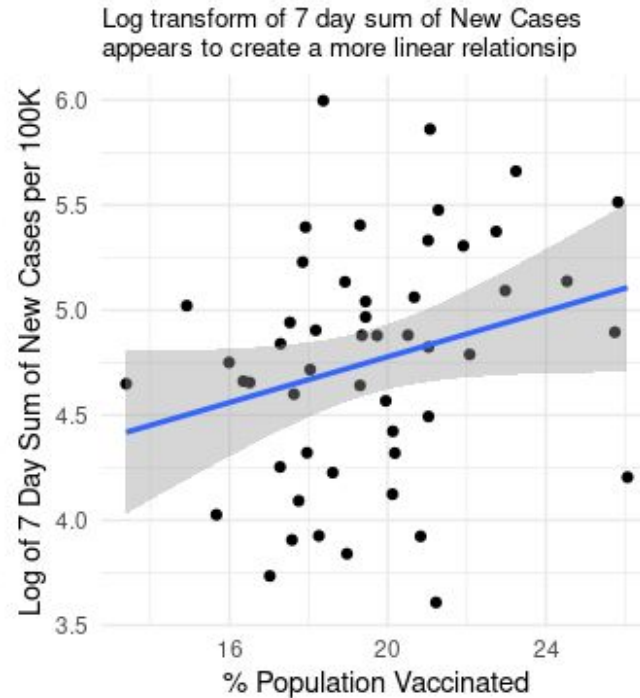
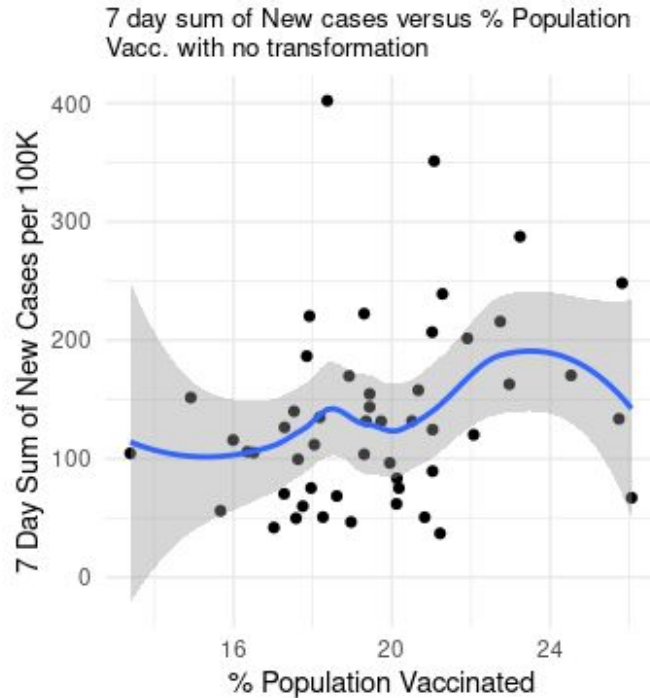
- 14 day buffer from vaccination date to sum of new cases



Data Table

State	Effective Date	% Vaccinated	7 day sum of new cases per 100K
CA	March 31, 2021	35	115

# Appendix - Data Transformations: Log-Linear Base Model



# Appendix- CLM of Improved Model

1. IID Sampling
2. Linear Conditional Expectation
3. No Perfect Collinearity
4. Homoskedastic Errors
5. Normally Distributed Errors

# Appendix- CLM of Improved Model

## 1. IID Sampling

- 2. Linear Conditional Expectation
- 3. No Perfect Collinearity
- 4. Homoskedastic Errors
- 5. Normally Distributed Errors

**Covid case  
counts**

**Vaccination  
Rates**

**Density**

**Policy Data**

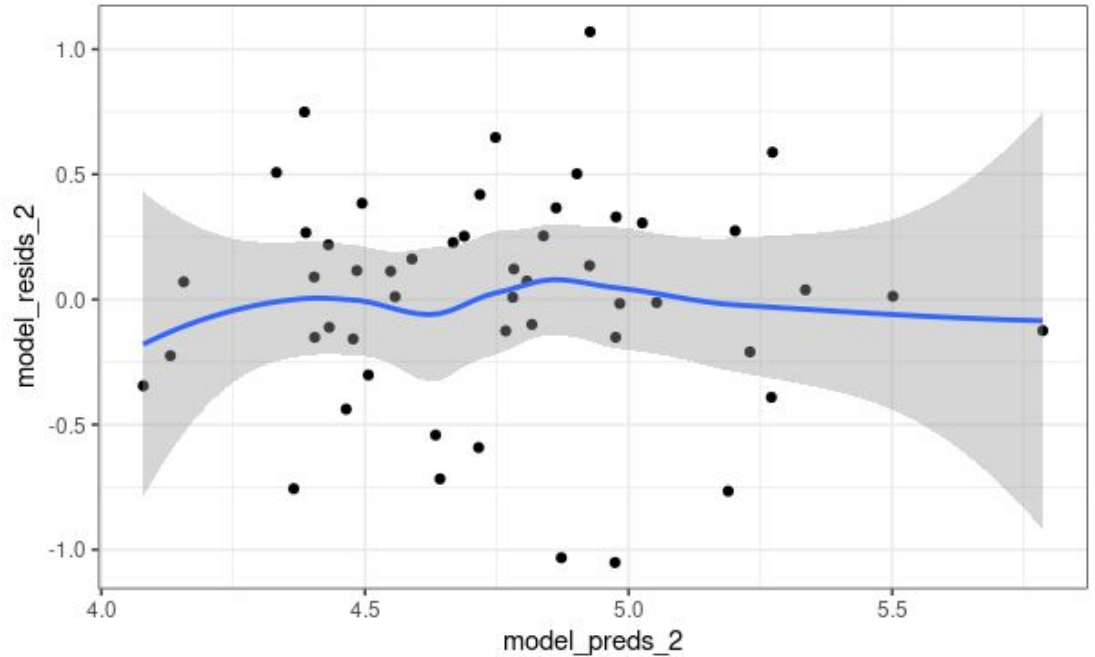
**Movement  
Data**

*Issues with Independence*

*No Issues with IID assumptions*

# Appendix- CLM of Improved Model

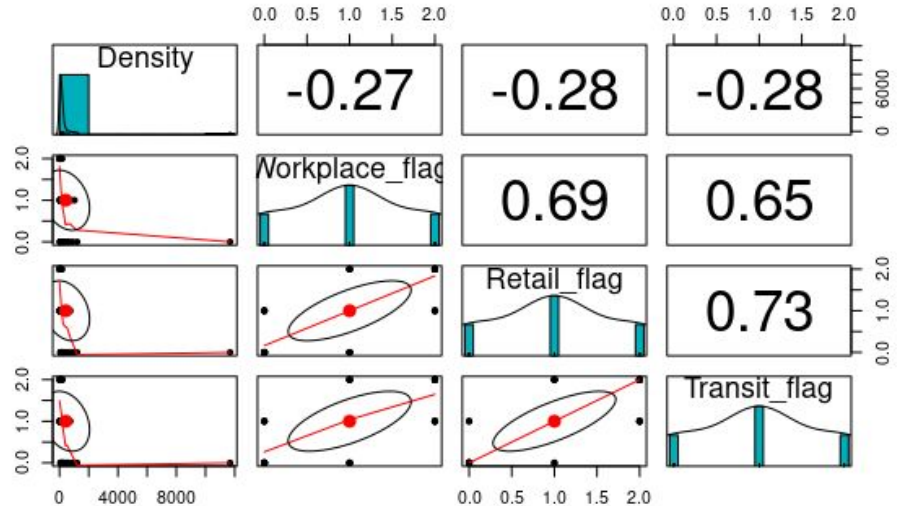
1. IID Sampling
- 2. Linear Conditional Expectation**
3. No Perfect Collinearity
4. Homoskedastic Errors
5. Normally Distributed Errors





# Appendix- CLM of Improved Model

1. IID Sampling
2. Linear Conditional Expectation
- 3. No Perfect Collinearity**
4. Homoskedastic Errors
5. Normally Distributed Errors

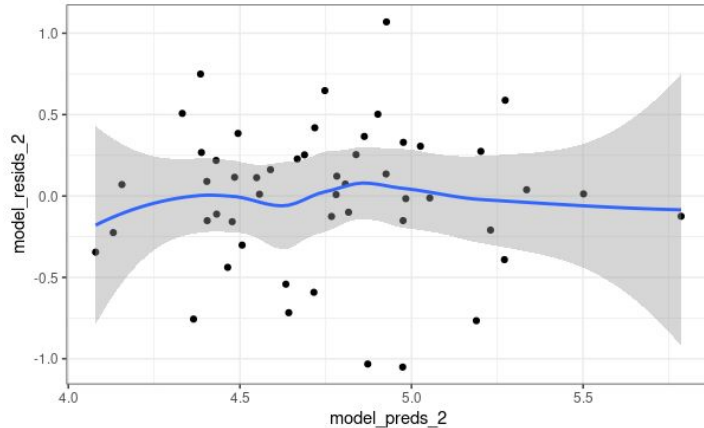


## VIF Results:

people vaccinated_per_hundred		log(Density)
Workplace flag	1.242642	2.009528
	2.262572	
	Retail_flag	Transit_flag
	2.777849	2.561925

# Appendix- CLM of Improved Model

1. IID Sampling
2. Linear Conditional Expectation
3. No Perfect Collinearity
- 4. Homoskedastic Errors**
5. Normally Distributed Errors

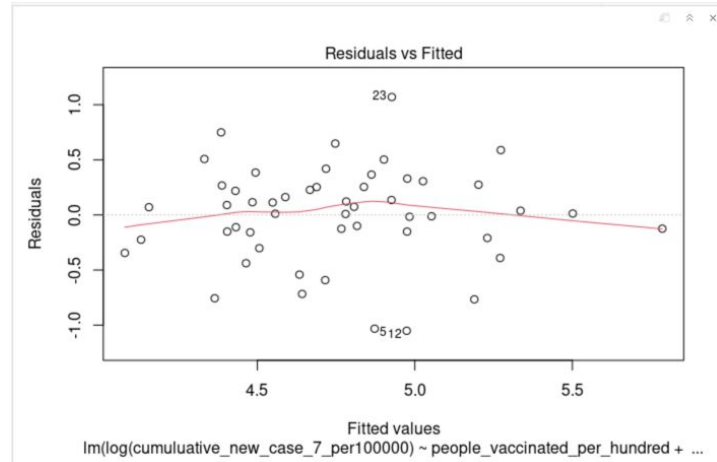


To further test for homoskedastic errors we also ran the Breusch Pagan test:

```
```{r}  
bptest(model_two)  
```
```

studentized Breusch-Pagan test

data: model\_two  
BP = 5.4459, df = 5, p-value = 0.3639



# Appendix- CLM of Improved Model

1. IID Sampling
2. Linear Conditional Expectation
3. No Perfect Collinearity
4. Homoskedastic Errors
- 5. Normally Distributed Errors**

