

# Appendices

## A Model Details

The baseline model is adapted and expanded from Carroll, Slacalek, Tokuoka, and White (2017). The economy consists of a continuum of expected utility maximizing households with a common CRRA utility function over consumption,  $u(\mathbf{c}, \eta) = \eta \mathbf{c}^{1-\rho} / (1 - \rho)$ , where  $\eta$  is a marginal utility shifter. Households are *ex ante* heterogeneous: household  $i$  has a quarterly time discount factor  $\beta_i \leq 1$  and an education level  $e_i \in \{D, HS, C\}$  (for dropout, high school, and college, respectively). Each quarter, the household receives (after tax) income, chooses how much of their market resources  $\mathbf{m}_{it}$  to consume  $\mathbf{c}_{it}$  and how much to retain as assets  $\mathbf{a}_{it}$ ; they then transition to the next quarter by receiving shocks to mortality, income, their employment state, and their marginal utility of consumption.

For each education group  $e$ , we assign a uniform distribution of time preference factors between  $\beta_e - \nabla$  and  $\beta_e + \nabla$ , chosen to match the distribution of liquid wealth and retirement assets. Specifically, the calibrated values in Table 1 fit the ratio of liquid wealth to permanent income in aggregate for each education level, as computed from the 2004 Survey of Consumer Finance. The width of the distribution of discount factors was calibrated to minimize the difference between simulated and empirical Lorenz shares of liquid wealth for the bottom 20%, 40%, 60%, and 80% of households, as in Carroll, Slacalek, Tokuoka, and White (2017).

When transitioning from one period to the next, a household with education  $e$  that has already lived for  $j$  periods faces a  $D_{ej}$  probability of death. The quarterly mortality probabilities are calculated from the Social Security Administration's actuarial table (for annual mortality probability) and adjusted for education using Brown, Liebman, and Pollett (2002); a household dies with certainty if it (improbably) reaches the age of 120 years. The assets of a household that dies are completely taxed by the government to fund activities outside the model. Households who survive to period  $t + 1$  experience a return factor of  $R$  on their assets, assumed constant.

Household  $i$ 's state in period  $t$ , at the time it makes its consumption–saving decision, is characterized by its age  $j$ ,<sup>1</sup> a level of market resources  $\mathbf{m}_{it} \in \mathbb{R}_+$ , a permanent income level  $\mathbf{p}_{it} \in \mathbb{R}_{++}$ , a discrete employment state  $\ell_{it} \in \{0, 1, 2\}$  (indicating whether the individual is employed, normal unemployed, or deeply unemployed), and a discrete state  $\eta_{it} \in \{1, \underline{\eta}\}$  that represents whether its marginal utility of consumption has been temporarily reduced ( $\underline{\eta} < 1$ ). Denote the joint discrete state as  $n_{it} = (\ell_{it}, \eta_{it})$ .

Each household inelastically participates in the labor market when it is younger than 65 years ( $j < 164$ ) and retires with certainty at age 65. The transition from working life to retirement is captured in the model by a one time large decrease in

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<sup>1</sup>Households enter the model aged 24 years, so model age  $j = 0$  corresponds to being 24 years, 0 quarters old.

permanent income at age  $j = 164$ .<sup>2</sup> Retired households face essentially no income risk: they receive Social Security benefits equal to their permanent income with 99.99% probability and miss their check otherwise; their permanent income very slowly degrades as they age. The discrete employment state  $\ell_{it}$  is irrelevant for retired households.

Labor income for working age households is subject to three risks: unemployment, permanent income shocks, and transitory income shocks. Employed ( $\ell_{it} = 0$ ) households' permanent income grows by age-education-conditional factor  $\Gamma_{ej}$  on average, subject to a mean one lognormal permanent income shock  $\psi_{it}$  with age-conditional underlying standard deviation of  $\sigma_{\psi j}$ . The household's labor income  $\mathbf{y}_{it}$  is also subject to a mean one lognormal transitory shock  $\xi_{it}$  with age-conditional underlying standard deviation of  $\sigma_{\xi j}$ . The age profiles of permanent and transitory income shock standard deviations are approximated from the results of Sabelhaus and Song (2010), and the expected permanent income growth factors are adapted from Cagetti (2003). Normal unemployed and deeply unemployed households receive unemployment benefits equal to a fraction  $\underline{\xi} = 0.3$  of their permanent income,  $\mathbf{y}_{it} = \underline{\xi}\mathbf{p}_{it}$ ; they are not subject to permanent nor transitory income risk, but their permanent income degrades at rate  $\underline{\Gamma}$ , representing "skill rot".<sup>3</sup>

The income process for a household can be represented mathematically as:

$$\mathbf{p}_{it} = \begin{cases} \psi_{it}\Gamma_{ej}\mathbf{p}_{it-1} & \text{if } \ell_{it} = 0, j < 164 & \text{Employed, working age} \\ \underline{\Gamma}\mathbf{p}_{it-1} & \text{if } \ell_{it} > 0, j < 164 & \text{Unemployed, working age ,} \\ \Gamma_{ret}\mathbf{p}_{it-1} & \text{if } j \geq 164 & \text{Retired} \end{cases}$$

$$\mathbf{y}_{it} = \begin{cases} \xi_{it}\mathbf{p}_{it} & \text{if } \ell_{it} = 0, j < 164 & \text{Employed, working age} \\ \underline{\xi}\mathbf{p}_{it} & \text{if } \ell_{it} > 0, j < 164 & \text{Unemployed, working age .} \\ \mathbf{p}_{it} & \text{if } j \geq 164 & \text{Retired} \end{cases}$$

A working-age household's employment state  $\ell_{it}$  evolves as a Markov process described by the matrix  $\Xi$ , where element  $k, k'$  of  $\Xi$  is the probability of transitioning from  $\ell_{it} = k$  to  $\ell_{it+1} = k'$ . During retirement, all households have  $\ell_{it} = 0$  (or any other trivializing assumption about the "employment" state of the retired). We assume that households treat  $\Xi_{0,2}$  and  $\Xi_{1,2}$  as zero: they do not consider the possibility of ever attaining the deep unemployment state  $\ell_{it} = 2$  from "normal" employment or unemployment, and thus it does not affect their consumption decision in those employment states.

We specify the unemployment rate during normal times as  $\mathfrak{U} = 5\%$ , and the expected duration of an unemployment spell as 1.5 quarters. The probability of transitioning from unemployment back to employment is thus  $\Xi_{1,0} = \frac{2}{3}$ , and the probability of becoming unemployed is determined as the flow rate that offsets

<sup>2</sup>The size of the decrease depends on education level, very roughly approximating the progressive structure of Social Security:  $\Gamma_{D164} \approx 0.56$ ,  $\Gamma_{HS164} \approx 0.44$ ,  $\Gamma_{C164} \approx 0.31$ .

<sup>3</sup>Unemployment is somewhat persistent in our model, so the utility risk from receiving 15% of permanent income for one quarter (as in Carroll, Slacalek, Tokunaka, and White (2017)) is roughly the same as the risk of receiving 30% of permanent income for 1.5 quarters in expectation.

this to generate 5% unemployment (about 3.5%). The deeply unemployed expect to be unemployed for *much* longer: we specify  $\Xi_{2,0} = 0$  and  $\Xi_{2,1} = \frac{1}{3}$ , so that a deeply unemployed person remains so for three quarters on average before becoming “normal” unemployed (they cannot transition directly back to employment). Thus the unemployment spell for a deeply unemployed worker is 2 quarters at a minimum and 4.5 quarters on average.<sup>4</sup>

Like the prospect of deep unemployment, the possibility that consumption might become less appealing (via marginal utility scaling factor  $\eta_{it} < 1$ ) does not affect the decision-making process of a household in the normal  $\eta_{it} = 1$  state. If a household does find itself with  $\eta_{it} = \underline{\eta}$ , this condition is removed (returning to the normal state) with probability 0.5 each quarter; the evolution of the marginal utility scaling factor is represented by the Markov matrix  $H$ . In this way, the consequences of a pandemic are fully unanticipated by households, a so-called “MIT shock”; households act optimally once in these states, but did not account for them in their consumption–saving problem during “normal” times.<sup>5</sup>

The household’s permanent income level can be normalized out of the problem, dividing all boldface variables (absolute levels) by the individual’s permanent income  $\mathbf{p}_{it}$ , yielding non-bold normalized variables, e.g.,  $m_{it} = \mathbf{m}_{it}/\mathbf{p}_{it}$ . Thus the only state variables that affect the choice of optimal consumption are normalized market resources  $m_{it}$  and the discrete Markov states  $n_{it}$ . After this normalization, the household consumption functions  $c_{e,j}$  satisfy:

$$\begin{aligned} v_{e,j}(m_{it}, n_{it}) &= \max_{c_{e,j}} u(c_{e,j}(m_{it}, n_{it}), \eta_{it}) + \beta_i(1 - D_{e,j}) \mathbb{E}_t \left[ \widehat{\Gamma}_{it+1}^{1-\rho} v_{e,j+1}(m_{it+1}, n_{it+1}) \right] \\ &\text{s.t.} \\ a_{it} &= m_{it} - c_{e,j}(m_{it}, n_{it}), \\ m_{it+1} &= (R/\widehat{\Gamma}_{it+1})a_{it} + y_{it}, \\ n_{it+1} &\sim (\Xi, H), \\ a_{it} &\geq 0, \end{aligned}$$

where  $\widehat{\Gamma}_{it+1} = \mathbf{p}_{it+1}/\mathbf{p}_{it}$ , the realized growth rate of permanent income from period  $t$  to  $t + 1$ . Consumption function  $c_{e,j}$  yields optimal *normalized* consumption, the ratio of consumption to the household’s permanent income level; the actual consumption level is simply  $\mathbf{c}_{it} = \mathbf{p}_{it}c_{e,j}(m_{it}, n_{it})$ .

Starting from the terminal model age of  $j = 384$ , representing being 120 years old (when the optimal choice is to consume all market resources, as death is certain), we solve the model by backward induction using the endogenous grid method, originally presented in Carroll (2006). Substituting the definition of next

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<sup>4</sup>Our computational model allows for workers’ *beliefs* about the average duration of deep unemployment to differ from the *true* probability. However, we do not present results based on this feature and thus will not further clutter the notation by formalizing it here.

<sup>5</sup>Our computational model also allows households’ beliefs about the duration of the reduced marginal utility state (via social distancing) to deviate from the true probability. The code also permits the possibility that the reduction in marginal utility is lifted as an aggregate or shared outcome, rather than idiosyncratically. We do not present results utilizing these features here, but invite the reader to investigate their predicted consequences using our public repository.

period's market resources into the maximand, the household's problem can be rewritten as:

$$v_{e,j}(m_{it}, n_{it}) = \max_{c_{it} \in \mathbb{R}_+} u(c_{it}, \eta_{it}) + \beta_i(1 - D_{e,j}) \mathbb{E}_t \left[ \widehat{\Gamma}_{it+1}^{1-\rho} v_{e,j+1}((R/\widehat{\Gamma}_{it+1})a_{it} + y_{it}, n_{it+1}) \right]$$

$$\text{s.t. } a_{it} = m_{it} - c_{it}, \quad a_{it} \geq 0, \quad n_{it+1} \sim (\Xi, H).$$

This problem has one first order condition, which is both necessary and sufficient for optimality. It can be solved to yield optimal consumption as a function of (normalized) end-of-period assets and the Markov state:

$$\underbrace{\eta_{it} c_{it}^{-\rho}}_{=\frac{\partial u}{\partial c}} - \underbrace{\beta_i R(1 - D_{e,j}) \mathbb{E}_t \left[ \widehat{\Gamma}_{it+1}^{-\rho} v_{e,j+1}^m((R/\widehat{\Gamma}_{it+1})a_{it} + y_{it}, n_{it+1}) \right]}_{\equiv v_{e,j}^a(a_{it}, n_{it})} = 0 \implies c_{it} = \left( \frac{v_{e,j}^a(a_{it}, n_{it})}{\eta_{it}} \right)^{-\frac{1}{\rho}}.$$

To solve the age- $j$  problem numerically, we specify an exogenous grid of end-of-period asset values  $a \geq 0$ , compute end-of-period marginal value of assets at each gridpoint (and each discrete Markov state), then calculate the unique (normalized) consumption that is consistent with ending the period with this quantity of assets while acting optimally. The beginning-of-period (normalized) market resources from which this consumption was taken is then simply  $m_{it} = a_{it} + c_{it}$ , the *endogenous gridpoint*. We then linearly interpolate on this set of market resources-consumption pairs, adding an additional bottom gridpoint at  $(m_{it}, c_{it}) = (0, 0)$  to represent the liquidity-constrained portion of the consumption function  $c_{e,j}(m_{it}, n_{it})$ .

The standard envelope condition applies in this model, so that the marginal value of market resources equals the marginal utility of consumption when consuming optimally:

$$v_{e,j}^m(m_{it}, n_{it}) = \eta_{it} c_{e,j}(m_{it}, n_{it})^{-\rho}.$$

The marginal value function for age  $j$  can then be used to solve the age  $j - 1$  problem, iterating backward until the initial age  $j = 0$  problem has been solved.

When the pandemic strikes, we draw a new employment state (employed, unemployed, deeply unemployed) for each working age household using a logistic distribution. For each household  $i$  at  $t = 0$  (the beginning of the pandemic and lockdown), we compute logistic weights for the employment states as:

$$\mathbb{P}_{i,\ell} = \alpha_{\ell,e} + \alpha_{\ell,p} \mathbf{p}_{i0} + \alpha_{\ell,j} j_{i0} \quad \text{for } \ell \in \{1, 2\}, \quad \mathbb{P}_{i,0} = 0,$$

where  $e \in \{D, H, C\}$  for dropouts, high school graduates, and college graduates and  $j$  is the household's age. The probability that household  $i$  draws employment state  $\ell \in \{0, 1, 2\}$  is then calculated as:

$$\Pr(\ell_{it} = \ell) = \exp(\mathbb{P}_{i,\ell}) / \sum_{k=0}^2 \exp(\mathbb{P}_{i,k}).$$

Our chosen logistic parameters are presented in Table 2.

## B Aggregation

Households are modeled as individuals and incomes sized accordingly. We completely abstract from family dynamics. To get our aggregate predictions for income and consumption, we take the mean from our simulation and multiply by 253 million, the number of adults (over 18) in the United States in 2019. To size the unemployment benefits correctly, we multiply the benefits per worker by 0.8 to account for the fact that 20 percent of the working-age population is out of the labor force, so the average working-age household consists of 0.8 workers and 0.2 non-workers. With this adjustment, there are 151 million workers eligible for unemployment benefits in the model. Aggregate consumption in our baseline for 2020 is just over \$11 trillion, a little less than total personal consumption expenditure, accounting for the fact that some consumption does not fit in the usual budget constraint.<sup>6</sup> Aggregating in this way underweights the young, as our model excludes those under the age of 24.

Our model estimates the aggregate size of the stimulus checks to be \$267 billion, matching the the Joint Committee on Taxation’s estimate of disbursements in 2020.<sup>7</sup> This is somewhat of a coincidence: we overestimate the number of adults who will actually receive the stimulus, while excluding the \$500 payment to children.

The aggregate cost of the extra unemployment benefits depends on the expected level of unemployment. Our estimate is \$137 billion, much less than the \$260 billion mentioned in several press reports, but in line with the extent of unemployment in our pandemic scenario. We do not account for the extension of unemployment benefits to the self-employed and gig workers.

Households enter the model at age  $j = 0$  with zero liquid assets. A ‘newborn’ household has its initial permanent income drawn lognormally with underlying standard deviation of 0.4 and an education-conditional mean. The initial employment state of households matches the steady state unemployment rate of 5%.<sup>8</sup>

We assume annual population growth of 1%, so older simulated households are appropriately down-weighted when we aggregate idiosyncratic values. Likewise, each successive cohort is slightly more productive than the last, with aggregate productivity growing at a rate of 1% per year. The profile of average income by age in the population at any moment in time thus has more of an inverted-U shape than implied by the permanent income profiles from Cagetti (2003).

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<sup>6</sup>PCE consumption in Q4 2019, from the NIPA tables, was \$14.8 trillion. Market based PCE, a measure that excludes expenditures without an observable price was \$12.9 trillion. Health care, much of which is paid by employers and not in the household’s budget constraint, was \$2.5 trillion.

<sup>7</sup>The JCT’s 26 March 2020 publication JCX-11-20 predicts disbursements of \$267 billion in 2020, followed by \$24 billion in 2021.

<sup>8</sup>This is the case even during the pandemic and lockdown, so the death and replacement of simulated agents is a second order contribution to the profile of the unemployment rate.

## References

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**Table 1** Parameter Values in the Baseline Model

Description	Parameter	Value
Coefficient of relative risk aversion	$\rho$	1
Mean discount factor, high school dropout	$\beta_D$	0.9637
Mean discount factor, high school graduate	$\beta_{HS}$	0.9705
Mean discount factor, college graduate	$\beta_C$	0.9756
Discount factor band (half width)	$\nabla$	0.0253
Employment transition probabilities:		
– from normal unemployment to employment	$\Xi_{1,0}$	2/3
– from deep unemployment to normal unemployment	$\Xi_{2,1}$	1/3
– from deep unemployment to employment	$\Xi_{2,0}$	0
Proportion of high school dropouts	$\theta_D$	0.11
Proportion of high school graduates	$\theta_{HS}$	0.55
Proportion of college graduates	$\theta_C$	0.34
Average initial permanent income, dropout	$\bar{p}_{D0}$	5000
Average initial permanent income, high school	$\bar{p}_{HS0}$	7500
Average initial permanent income, college	$\bar{p}_{C0}$	12000
Steady state unemployment rate	$\bar{U}$	0.05
Unemployment insurance replacement rate	$\xi$	0.30
Skill rot of all unemployed	$\underline{\Gamma}$	0.995
Quarterly interest factor	$R$	1.01
Population growth factor	$N$	1.0025
Technological growth factor	$\mathfrak{I}$	1.0025

**Table 2** Pandemic Assumptions

Description	Parameter	Value
Short-lived Pandemic		
Logistic parametrization of unemployment probabilities		
Constant for dropout, regular unemployment	$\alpha_{1,D}$	-1.15
Constant for dropout, deep unemployment	$\alpha_{2,D}$	-1.5
Constant for high school, regular unemployment	$\alpha_{1,H}$	-1.3
Constant for high school, deep unemployment	$\alpha_{2,H}$	-1.75
Constant for college, regular unemployment	$\alpha_{1,C}$	-1.65
Constant for college, deep unemployment	$\alpha_{2,C}$	-2.2
Coefficient on permanent income, regular unemployment	$\alpha_{1,p}$	-0.1
Coefficient on permanent income, deep unemployment	$\alpha_{2,p}$	-0.2
Coefficient on age, regular unemployment	$\alpha_{1,j}$	-0.01
Coefficient on age, deep unemployment	$\alpha_{2,j}$	-0.01
Marginal Utility Shock		
Pandemic utility factor	$\underline{\eta}$	0.891
Prob. exiting pandemic each quarter	$H_{1,0}$	0.5
Long, Deep Pandemic		
Logistic parametrization of unemployment probabilities		
Constant for dropout, regular unemployment	$\alpha_{1,D}$	-1.45
Constant for dropout, deep unemployment	$\alpha_{2,D}$	-0.3
Constant for high school, regular unemployment	$\alpha_{1,H}$	-1.6
Constant for high school, deep unemployment	$\alpha_{2,H}$	-0.55
Constant for college, regular unemployment	$\alpha_{1,C}$	-1.95
Constant for college, deep unemployment	$\alpha_{2,C}$	-1.00
Coefficient on permanent income, regular unemployment	$\alpha_{1,p}$	-0.2
Coefficient on permanent income, deep unemployment	$\alpha_{2,p}$	-0.2
Coefficient on age, regular unemployment	$\alpha_{1,j}$	-0.01
Coefficient on age, deep unemployment	$\alpha_{2,j}$	-0.01
Marginal Utility Shock		
Pandemic utility factor	$\underline{\eta}$	0.891
Prob. exiting pandemic each quarter	$H_{1,0}$	0.25



**Table 3** Fiscal Stimulus Assumptions, CARES Act

Description	Value
Stimulus check	\$1,200
Means test start (annual)	\$75,000
Means test end (annual)	\$99,000
Stimulus check delay	1 quarter
Fraction that react on announcement	0.25
Extra unemployment benefit for:	
Normal unemployed	\$5,200
Deeply unemployed	\$7,800

Note: The unemployment benefits are multiplied by 0.8 to account for the fact that 20 percent of the working age population is out of the labor force. See aggregation details in Appendix B