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SIMULATION METHODS IN HEALTH SERVICES RESEARCH: APPLICATIONS
 FOR POLICY, MANAGEMENT, AND PRACTICE

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Microsimulation of Private Health Insurance and Medicaid Take-up Following the U.S. Supreme Court Decision Upholding the Affordable Care Act

Stephen T. Parente and Roger Feldman

Objective. To predict take-up of private health insurance and Medicaid following the U.S. Supreme Court decision upholding the Affordable Care Act (ACA).

Data Sources. Data came from three large employers and a sampling of premiums from ehealthinsurance.com. We supplemented the employer data with information on state Medicaid eligibility and costs from the Kaiser Family Foundation. National predictions were based on the MEPS Household Component.

Study Design. We estimated a conditional logit model of health plan choice in the large group market. Using the coefficients from the choice model, we predicted take-up in the group and individual health insurance markets. Following ACA implementation, we added choices to the individual market corresponding to plans that will be available in state and federal exchanges. Depending on eligibility for premium subsidies, we reduced the out-of-pocket premiums for those choices. We simulated several possible patterns for states opting out of the Medicaid expansion, as allowed by the Supreme Court.

Principal Findings. The ACA will increase coverage substantially in the private insurance market and Medicaid. HSAs will remain desirable in both the individual and employer markets.

Conclusions. If states opt out of the Medicaid expansion this could increase the federal cost of health reform, while reducing the number of newly covered lives.

Key Words. Insurance, health reform, microsimulation, uninsured, health savings accounts

On June 28, 2012, the Supreme Court of the United States (SCOTUS) upheld most of the provisions of the Patient Protection and Affordable Care Act and the health care provisions of the Health Care and Education Reconciliation Act (P.L. 111-148 and P.L. 111-152; henceforth referred to as the ACA).¹

1 Starting in 2014, individuals without an offer of insurance from their employer
2 and small businesses will be able to buy insurance on state and federal
3 exchanges, with premium subsidies depending on their incomes. Certain
4 employers that do not offer health insurance will be penalized, and individuals
5 will be required to have coverage or pay a penalty.

6 At the same time, however, the Supreme Court ruled that states could
7 opt out of the ACA expansion of Medicaid coverage for all individuals up to
8 age 65 with incomes less than 133 percent of poverty. Under the ACA as
9 enacted, but before the Supreme Court ruling, the Medicaid expansion was
10 mandatory for states that wanted to keep their federal matching funds for any
11 part of the Medicaid program. The Supreme Court's decision immediately
12 raised the prospect that some states might opt out of the Medicaid expansion.
13 The U.S. Congressional Budget Office (CBO 2012) estimated that six million
14 people previously covered by the Medicaid expansion in its March 2012 base-
15 line would not be covered; some of these would enroll in exchanges, but the
16 number of uninsured people would rise by four million.

17 Our research has two goals. First, we predict how many people will take
18 up private health insurance under provisions of the ACA. Second, we predict
19 Medicaid take-up under several possible patterns for states opting out of the
20 Medicaid expansion. Unlike the CBO, which did not make estimates for spe-
21 cific states but instead utilized average probabilities of opting out, we make
22 predictions for specific states.² We also predict enrollment in specific types of
23 private plans (e.g., the "metallic" plans offered in health insurance exchanges).

24 We find the ACA will increase coverage substantially in the private
25 health insurance market and Medicaid. However, if states opt out of the Med-
26 icaid expansion this could *increase* the federal cost of health reform, while
27 reducing the number of newly covered lives. If six states (Florida, Louisiana,
28 Mississippi, Nebraska, South Carolina, and Texas) opt out, the number of
29 uninsured people will increase by 7.9 million with a drop in Medicaid cover-
30 age of 4.4 million by 2021, compared with the pre-SCOTUS situation.

31 Our predictions are based on a microsimulation model of health plan
32 choice that we originally developed to predict the effect of the Medicare Mod-
33 ernization Act of 2003 (MMA) on take-up of high-deductible health plans in
34

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1 the individual health insurance market (Feldman et al. 2005; Parente et al.
2 2005).³ We begin the study with a section that describes the model. This is
3 followed by our simulation of the ACA effects on private health insurance and
4 Medicaid take-up.

6 MICROSIMULATION MODEL

7
8
9 Our approach has three major components: (1) Model Estimation; (2) Choice
10 Set Assignment and Prediction; and (3) Policy Simulation. Integral to this anal-
11 ysis was the use of data from three large employers working with the study
12 investigators.

13 The model estimation had several steps. First, we pooled the data from
14 the employers to estimate a conditional logit plan choice model. In the second
15 step, we used the estimated choice model coefficients to predict health plan
16 choices for individuals in the MEPS-HC. To complete this step, it was neces-
17 sary to assign the number and types of health insurance choices that are avail-
18 able to each respondent in the MEPS-HC. For this purpose we turned to the
19 smaller, but more detailed MEPS Household Component–Insurance Com-
20 ponent linked file, which contained the needed information. The third step
21 was to generate HSA premiums and benefit designs. The final step was to
22 apply plan choice model coefficients to the MEPS data with premium infor-
23 mation as well as ACA provisions to estimate insurance take up and subsidy
24 costs. We expand on each of these steps.

26 *Estimate Plan Choice Model*

27
28 We pooled the data from the employers to estimate a conditional logit model
29 of health plan choice similar to our earlier work (Parente, Feldman, and Chris-
30 tianson 2004). Conceptually, the choice model is based on utility maximiza-
31 tion, where utility is considered to be a function of health plan attributes (such
32 as the out-of-pocket premium) and interactions of health plan and personal
33 attributes, formally stated as:

$$34 U_{ij} = f(Z_j, X_{ij})$$

35
36 where i is the decision making employee choosing among j health plans on
37 the basis of Z_j health plan attributes and X_{ij} interactions of health plan and per-
38 sonal attributes. The interaction terms prevent personal attributes from “dif-
39 ferencing away” when the individual compares plans. An example is an

1 interaction of the worker's health status with the plan's premium, which allows
2 people of different health status to have difference choice responses to plan
3 premiums.

4 A very important constraint in our modeling was that any variable used
5 in the plan choice model from the employer data also had to be available in
6 the MEPS-HC to permit a simulation. As a result, the health plan variables
7 used in the choice model were: the after-tax premium paid by the employee;
8 the amount of money in the employee's health reimbursement account
9 (HRA), if any; the difference between the employee's plan deductible and
10 HRA; and the plan coinsurance rate. The personal attributes were: whether
11 the employee or dependent has a chronic illness; the employee's age and
12 gender; the employee has a two-person or family contract (with single contract
13 as the reference); and the employee's annual wage income.

14 Also included in the regression were alternative-specific constants
15 (intercepts) for each health plan choice. These intercepts capture plan-spe-
16 cific features not represented by other identifiers of plan design. They are
17 also included as interaction terms with age, gender, family status, and
18 income. The intercept terms include seven choices: low-option PPO (restrict-
19 tive network, high copay, 15 percent coinsurance); medium-option PPO
20 (better network, lower copay and coinsurance); high-option PPO (open net-
21 work, lowest copay, no coinsurance); Health Reimbursement Arrangement
22 (HRA); Health Savings Account (HSA) with employer contribution, mod-
23 eled on higher premium HRA; HSA without employer contribution,
24 modeled on lower premium HRA; and Health Maintenance Organization
25 (HMO). The intercept terms for each person represent only the choices he
26 or she has from the employer; most employees have a subset of the choices
27 available. The low-option PPO, available to all employees, was the refer-
28 ence choice.

30 *Choice Set Assignment and Prediction*

31
32 We used the health plan choice model to develop two sets of plan choice pre-
33 dictions: one set for workers with insurance offers and a second set for individ-
34 uals who do not have employer offers of coverage. The second set includes
35 both uninsured individuals and those who take up nongroup policies. Nonof-
36 ferred individuals who reported having employer group coverage through
37 another household member are excluded from the simulations. Below we
38 outline the analytic steps taken to develop the individuals' choice sets for the
39 simulations.

1 Unlike in the employer data, in the MEPS-HC we do not observe the
2 person's plan choice set, so we had to predict it from the linked MEPS HC-IC
3 (as described in Feldman et al. 2005). The predictions included both the number
4 and types of choices. For example, workers in large firms have more plan
5 choices than those in small firms, so a worker in a large firm might have four
6 choices: low-option PPO, medium-option PPO, high-option PPO, and
7 HMO.

8 To these choices we added two additional options, conditional on the
9 size of the establishment where the employee works. We assumed that an
10 employer-sponsored HRA and an employer-sponsored HSA are available to
11 all workers in establishments with more than 500 employees, but not available
12 to other workers. For the HSA choices, we estimated the amount of money
13 that employers and individuals contribute to their Health Savings Accounts,
14 letting the contributions vary by age and income of the policyholder. Finally,
15 all employees with an offer of group insurance have the option of a self-
16 financed (full cost) HSA or of turning down coverage.

17 Individuals who did not have health insurance offered to them at work
18 or who were not employed faced five health plan choices regardless of
19 income, age or gender: high-option PPO, medium-option PPO, low-option
20 PPO, self-financed HSA, or uninsured. A self-financed HRA is not an option
21 for this group because only employers may contribute to an HRA.

22 With the sets of possible choices for workers with insurance offers and
23 individuals without insurance offers, we used the plan choice model to predict
24 plan choice probabilities for each MEPS-HC sample respondent.⁴ However,
25 before we could predict the probabilities, we needed to develop assumptions
26 about benefits and premiums for individual plans. To get premium estimates,
27 we used the linked MEPS insurance data to estimate a hedonic model of pre-
28 miums for group insurance plans. To predict premiums for individuals with-
29 out employer offers of coverage, we used the smallest establishment size
30 category, based on the assumption that this most closely represents an individ-
31 ual policy in terms of the loading charge for plan administrative costs.

32 The plan characteristics that we used to define the three PPOs (low,
33 medium, and high) came from the eHealthinsurance.com survey of plans pur-
34 chased in the individual market. Roughly speaking, we used the 25th, 50th,
35 and 75th percentiles of coinsurance and deductibles for assigning the plan
36 characteristics.

37 We also recognized that premiums in the individual market vary by a
38 person's age. The MEPS included a table of average premiums by age cohort.
39 We created an index using the information from this table. The index was set

1 equal to 1.0 for the age group corresponding to the median age of adults in our
2 sample (35–39). Older individuals, who had higher premiums, had index val-
3 ues that were greater than 1.0. Younger individuals, who had lower premiums,
4 had index values less than 1.0. The index values ranged from .59 to 2.18 for
5 single coverage policies and .453 to 1.65 for family coverage policies. Finally,
6 we adjusted all premiums to current levels.

7 An important feature of our work was the creation of synthetic states as
8 MEPS only shows the region where a respondent lived. We needed to identify
9 states to gauge the state-specific impact of ACA given the differences in exist-
10 ing Medicaid programs among states. For this analysis, we used an approach
11 from our prior research (Parente et al. 2011) on the impact of interstate health
12 insurance markets. The creation of the synthetic states is a four-step process.
13 First, we used the American Community Survey (ACS) to define the strata that
14 would be used to generate the sample. The strata included four variables: age,
15 income, gender, and race. Using person weights in the ACS, we tabulated the
16 population frequencies for each of these strata by state. Second, we divided
17 the MEPS into four regions—Northeast, Midwest, South, and West. The Dis-
18 trict of Columbia is in the South region. We then drew a random sample with
19 replacement of 1,000 (approximately, given rounding) observations from the
20 region containing a particular state. The frequency of observations by strata
21 was matched to represent the population. After the random samples were
22 drawn, the data were appended to form a national dataset. In the third step, we
23 validated our state assignments using the CPS. In our fourth and final step,
24 we merged several other variables into the file and selected the sample to
25 mimic the one we have used previously in simulations (Feldman et al. 2005).

26 One significant issue with our simulation is that we were not able to pre-
27 dict whether an individual would take-up insurance in the employer-offered
28 market or be uninsured in the individual market. We faced this limitation
29 because the employer data used to estimate the plan choice model included
30 information only on workers who held coverage.

31 To address this issue, we “calibrated” the model to accurately reflect both
32 the actual percentage of people who turn down employer offers and the actual
33 percentage of people in the individual market who are uninsured. Calibration
34 means setting the plan-specific intercepts for these choices at values that repro-
35 duce the known probabilities (e.g., the probability of turning down an
36 employer offer). To obtain more accurate estimates, we performed these cali-
37 brations by four quartiles of income and then compared our results with
38 national turn-down and uninsurance rates. We also applied the national popu-
39 lation weights to the calibrated model to represent the entire adult population,

1 excluding full-time students, those with public insurance, and individuals with
2 employer-based coverage through another household member. This fairly
3 tedious process was performed for each reestimation and/or modification in
4 the conditional logit choice model.

5 6 *Policy Simulation* 7

8 Two of the most substantial advances in our microsimulation were inclusion of
9 chronic illness in the plan choice model and generation of premiums through
10 an iterative process using prior years' claims data to create actuarially fair esti-
11 mates of premiums. Below we describe in more detail the issues we addressed
12 in this analysis.

13 Health status is important in the simulations for two reasons. First, health
14 status may be an important factor in predicting plan choice, so the addition of
15 this variable should improve the fit of the choice model, other things equal.
16 Second, sick (healthy) people may prefer certain plans, which would drive the
17 premiums up (down). Specifically, if sick people are attracted to traditional
18 plans, it could lead to a "death spiral" of increasing premiums and falling
19 enrollment in the traditional plans. One of our goals in the simulations is to
20 determine whether favorable selection into the HRA and HSA choices will
21 tend to destabilize the health insurance market.

22 To account for health status, we used claims data for contract holders
23 (employees) from the employers in the plan choice model. We obtained the
24 claims data for the year prior to their possible enrollment in a high-deductible
25 health plan. We used the diagnosis code information from these prior-year
26 claims to calculate a set of 34 Adjusted Diagnosis Groups (ADGs) using a
27 methodology developed by Johns Hopkins University researchers (Weiner
28 et al. 1991). Several of these 34 ADGs identify a diagnosis indicating the pres-
29 ence of a chronic condition. With this information we constructed a dummy
30 variable indicating the presence of chronic illness and included this variable in
31 the plan choice model interacted with plan premiums and cost sharing.
32 Because the chronic condition indicator also is available on the MEPS-HC,
33 this aggregate measure of health status allowed us to predict health plan choice
34 in the MEPS database.

35 This variable also permitted us to develop a medical care cost regression
36 to predict medical expenditure of the MEPS population enrolled in each plan
37 type. To capture the relation between cost and health risk, we estimated a
38 medical care cost regression for the individuals who chose each plan. We used
39 that model to predict premiums that fed back into the choice model. We

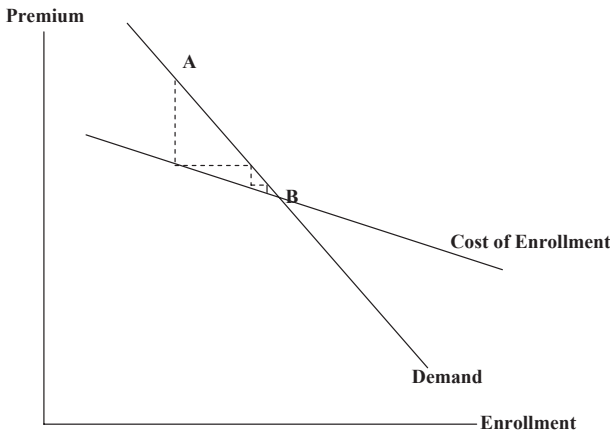
1 “iterated”—that is, went back and forth—between the choice model and the
 2 cost model until the market converged to a stable set of choices and premiums.
 3 Our method is illustrated in Figure 1. Starting from a premium that is too high
 4 for equilibrium (point A), the premium falls and enrollment increases until the
 5 two lines converge to a single premium and enrollment (Point B). There is no
 6 guarantee that the model will be stable as shown here. We know that the
 7 demand for enrollment is downward sloping, but the cost of enrollment also
 8 slopes downward for plans with unfavorable selection. The model will be stable
 9 if the demand curve is the steeper of the two.

10 To implement this iterative approach, we had to construct premiums
 11 from expected medical care costs in the individual and ESI markets. Premiums
 12 obviously depend on expected costs, but they also depend on how costs
 13 are aggregated across individuals. How many individuals are in the insurance
 14 pool? Does the premium for a particular person depend on his or her experience,
 15 or on the experience of the group? In other words, how are premiums
 16 “rated” in the individual and ESI markets?

17 The two rating methods we used were individual experience rating
 18 (IER) for the individual health insurance market and group experience rating
 19 (GER) for the ESI market.⁵ The premium for each rating method was generated
 20 as follows:

21
 22 *Individual Market.* Given that the premium for each person in the individual
 23 market is based on his or her own health care costs, we estimated how much
 24 each person would spend under each type of insurance plan (PPO, HSA, etc.).

25
 26 Figure 1: Model of Health Plan Demand and Cost



1 We added a loading fee of 30 percent to arrive at the premium for each choice
2 in the individual market.⁶
3
4

5 *Employer-sponsored Market.* The first step in GER is to define the “pool” that
6 determines the premium rates. We used three pools based on establishment
7 size—small establishments, medium-size establishments, and large establish-
8 ments.⁷ We predicted the cost of each person in each plan. Then, we calculated
9 the average cost across all people who work for employers in each of the three
10 pools. For example, the average cost of the HMO for employees of small
11 establishments may be \$6,000 for a single policy and \$12,000 for a family pol-
12 icy. The average cost of the HMO was different in medium-size and large
13 groups. Then, we added loading fees to get predicted premiums for each pool
14 in the ESI market.
15

16 17 ACA SIMULATION

18
19 We used the methods described above to simulate the impact of the ACA fol-
20 lowing the June 28, 2012 Supreme Court decision to allow states to opt out of
21 the Medicaid expansion. The two primary ACA elements we modeled were
22 as follows: (1) expansion of the private insurance market through state and
23 federal insurance exchanges; and (2) expansion of the Medicaid program as
24 outlined in the law for those with incomes less than 133 percent of the federal
25 poverty level (FPL). The data used for state-specific estimates were the syn-
26 thetic state assignment of MEPS-HC participants described earlier and in
27 detail in Parente et al. (2011).
28

29 *ACA at 2012 and 2014*

30
31 To simulate the impact of ACA, we needed to model the choices available in
32 the individual market exchange as well as the Medicaid expansions. For the
33 exchanges, we mapped existing health plan choices from our previous models
34 into the likely “metallic” plan choices that are part of ACA. We made this map-
35 pings based on our knowledge and assumptions about the existing health
36 insurance options and the likely order of generosity of benefit design. Specifi-
37 cally, we assumed: (1) Platinum plans will be the equivalent of a high-option
38 PPO; (2) Gold plans will map to a medium-option PPO; (3) Silver plans will
39 map to a low-option PPO; and (4) Bronze plans will map to a high-deductible

1 health plan (HDHP). We then needed to assign premiums for these plans in
2 the exchange. Here, we assumed the current pricing of the high-option PPO
3 was the reference point for 90 percent of the actuarial value requirement of
4 the ACA for the Platinum plan. From there, we arithmetically computed the
5 premiums for Gold, Silver, and Bronze plans that produced actuarial value
6 ranges of 80, 70, and 60 percent, respectively. This led to premiums in the
7 ACA that were out of sync from market conditions, particularly from the
8 HDHPs as Bronze plans. The existing premiums for HDHPs in our simula-
9 tion prior to 2014 are below the 60 percent actuarial value required in the
10 Bronze plan. We also calibrated our premiums to reflect higher loading costs
11 for community rating as well as to conform to the 3 : 1 modified adjusted
12 community rating requirement using age bands common for actuarial analy-
13 sis. In almost all of these calculations, there were no final rules written for
14 ACA operation so we needed to rely on assumptions based on prior research.

15 To simulate the Medicaid expansion we used Kaiser Family Foundation
16 (Kaiser Family Foundation 2012a,b) data on benefit coverage and costs of **2**
17 acute-care Medicaid programs by state. This gave us the ability to cost out the
18 Medicaid expansion in each state and then compare it with the cost of private-
19 sector coverage expansion. We also identified the percent of FPL ceilings cur-
20 rently used to qualify for Medicaid in each state, so we could identify which
21 states expanded their programs from the *status quo*. Finally, we used personal
22 income, age, and gender from the MEPS to identify the population in each
23 state that would be eligible for expanded Medicaid.

24 The results of our simulations are presented in Tables 1–3. In Table 1,
25 we provide estimates of 2012 and 2014 insurance coverage across the entire
26 under 65-year-old market affected by ACA. In 2012, we estimate the number
27 of uninsured to be 54.8 million. Due to ACA we forecast this number will
28 drop 38 percent to 34 million. The majority of the change will occur from
29 expansion of the private insurance market with over 21 million newly covered
30 enrollees. Over 17 million people will gain Medicaid coverage due to expan-
31 sion of the program in all 50 states.

32 At baseline, the iterations did not change the results of the individual
33 market health plan choices. Enrollment in bronze or catastrophic plans will
34 increase by 27 percent and enrollment in silver or more generous plans will
35 increase by 55 percent. HSAs that qualify as bronze plans may have too high a
36 premium to maintain the growth they had prior to 2012. However, HSAs
37 appear likely to continue to grow from our results.

38 For the group market, the iterations lead to substantial migration out of
39 the “turned down”, “PPO high”, and “PPO medium” choices with losses of 7,

1 Table 1: Affordable Care Act Impact from 2012 to 2014 in Insurance
 2 Demand

<i>Individual Market</i>	<i>2012</i>	<i>2014</i>	<i>Difference</i>	<i>% Change</i>
Silver or higher	13,077,268	20,285,299	7,208,030	55
Bronze or catastrophic Medicaid	19,539,213 34,855,438	24,721,721 52,012,641	5,182,508 17,157,203	27 49
Uninsured	48,854,416	28,542,663	(20,311,753)	-42
Group market				
HMO	5,896,887	5,870,908	(25,979)	0
HRA	18,060,845	17,665,282	(395,563)	-2
HSA_Funded	2,045,882	3,909,279	1,863,397	91
ESI 2 self-pay low PPO	102,199	3,537,014	3,434,815	>400
PPO high	22,395,612	18,486,371	(3,909,241)	-17
PPO low	2,821,926	3,391,734	569,808	20
PPO medium	81,931,286	69,754,544	(12,176,743)	-15
ESI 2 other insurance/Exchange	19,584,084	30,358,810	10,774,725	55
Employee refuses coverage	5,916,908	5,507,608	(409,300)	-7
Total uninsured	54,771,324	34,050,271	(20,721,053)	-38
Total private insurance	205,669,993	226,904,491	21,234,498	10
Total Medicaid	34,855,438	52,012,641	17,157,203	49

17, and 15 percent, respectively. The plans forecast to have the largest in-migration are self-funded HSAs, medium-option PPOs, and an opt-out low-option PPO where the consumer buys insurance on a qualified exchange and the employer pays a penalty. A large number of individuals (10.8 million) will find new coverage as a consequence of a switch in spousal coverage or are potentially dropped from an employer's plan and directed to a health insurance exchange.

Ten-Year Estimates of Health Reform following the 2012 Supreme Court Decision

We simulated the impact of ACA over a 10-year period from 2012 to 2021 as our baseline simulation. To test how our microsimulation could forecast policy changes, we also examined the effect of two policy changes based on the recent Supreme Court decision that states can opt out of the Medicaid expansion. This decision could greatly impact both the insurance coverage and the cost of the law. Shortly after the SCOTUS decision, 15 States were reported as not taking or very unlikely to take the Medicaid expansion in 2014 (The Hill 2012). Below we provide estimates the coverage and federal cost of ACA. These ACA estimates are compared with different possible outcomes from the SCOTUS ruling on Medicaid expansion from 2014 to 2021.

3
 Table 2: Pre- and Post-Supreme Court of the United States (SCOTUS) PPACA Coverage Impact Scenario: 6 States Decline to Take Expansion—FL, LA, MS, NE, SC, TX

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Pre-SCOTUS ruling										
Individual coverage										
Insured	32.6	31.3	45.0	46.9	47.0	45.3	43.7	42.0	40.4	38.9
Medicaid	31.7	32.1	41.9	42.6	43.5	44.9	46.1	47.4	48.7	50.0
Uninsured	48.9	50.9	28.5	27.1	27.3	28.8	30.3	31.9	33.4	34.9
Group coverage										
Insured	152.6	152.4	144.9	137.7	131.8	132.2	132.8	133.8	135.1	136.8
Take Medicaid coverage	3.2	3.8	10.1	11.5	12.7	13.8	14.7	15.6	16.2	16.9
Refuse coverage	5.9	7.2	5.5	5.9	6.3	6.8	7.2	7.4	7.3	7.1
Take individual coverage	0.1	0.1	4.5	11.7	17.6	17.3	17.0	16.8	16.5	16.2
Total coverage										
Insured	185.4	183.7	194.4	196.2	196.3	194.7	193.6	192.6	192.1	191.9
Medicaid	34.9	35.9	52.0	54.1	56.2	58.7	60.9	63.0	65.0	66.8
Uninsured	54.8	58.1	34.1	33.0	33.6	35.6	37.5	39.3	40.7	42.0
Total	275.0	277.7	280.5	283.3	286.1	289.0	291.9	294.8	297.8	300.7
Post-SCOTUS ruling										
Individual coverage										
Insured	32.6	31.3	46.9	48.1	47.3	44.7	42.2	39.3	36.5	33.9
Medicaid	31.7	32.1	38.9	39.6	40.5	41.7	42.8	44.2	45.6	47.1
Uninsured	48.9	50.9	29.6	28.9	29.9	32.6	35.1	37.9	40.4	42.7
Group coverage										
Insured	152.6	152.4	145.8	138.6	132.8	133.2	133.9	135.0	136.4	138.1

continued

Table 2. *Continued*

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Take Medicaid coverage	3.2	3.8	8.7	9.9	11.0	12.0	12.9	13.7	14.4	15.0
Refuse coverage	5.9	7.2	5.6	6.0	6.5	7.0	7.3	7.5	7.5	7.2
Take individual coverage	0.1	0.1	4.9	12.2	18.2	17.8	17.6	17.2	17.0	16.7
Total coverage	185.4	183.7	197.7	198.9	198.2	195.8	193.8	191.5	189.9	188.7
Insured	34.9	35.9	47.7	49.5	51.5	53.7	55.7	57.9	60.0	62.1
Medicaid	54.8	58.1	35.2	34.9	36.4	39.5	42.5	45.4	47.9	50.0
Uninsured	275.0	277.7	280.5	283.3	286.1	289.0	291.9	294.8	297.8	300.7
Net Impact										
Insured	0.0	0.0	3.2	2.7	1.9	1.0	0.2	-1.0	-2.2	-3.2
Medicaid	0.0	0.0	-4.4	-4.6	-4.7	-4.9	-5.2	-5.1	-5.0	-4.8
Uninsured	0.0	0.0	1.1	1.9	2.8	3.9	5.0	6.1	7.2	7.9

Table 3: Pre- and Post-Supreme Court of the United States (SCOTUS) PPACA Cost Impact (Billions) Scenario: 6
 States Decline to Take Expansion—FL, LA, MS, NE, SC, TX

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Pre-SCOTUS ruling										
Individual federal \$\$										
Insured	\$0.0	\$0.0	\$219.7	\$218.9	\$220.1	\$217.5	\$211.8	\$203.0	\$194.6	\$187.0
Medicaid	\$105.4	\$108.9	\$143.8	\$149.4	\$155.6	\$163.3	\$171.1	\$179.3	\$187.9	\$196.6
Uninsured	—	—	—	—	—	—	—	—	—	—
Group federal \$\$										
Insured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take Medicaid coverage	\$10.7	\$13.1	\$35.6	\$41.1	\$46.3	\$51.3	\$55.8	\$60.1	\$64.0	\$67.9
Refuse coverage	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take individual coverage	\$0.0	\$0.0	\$26.9	\$25.4	\$25.2	\$25.2	\$25.5	\$25.4	\$26.0	\$26.1
Total federal \$\$	\$0.0	\$0.0	\$246.6	\$244.2	\$245.3	\$242.7	\$237.3	\$228.5	\$220.6	\$213.1
Insured	\$116.1	\$122.0	\$179.4	\$190.5	\$201.9	\$214.5	\$226.9	\$239.4	\$251.9	\$264.4
Medicaid	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Uninsured	\$116.1	\$122.0	\$426.0	\$434.7	\$447.1	\$457.3	\$464.2	\$467.9	\$472.5	\$477.5
Total										
Post-SCOTUS ruling										
Individual federal \$\$										
Insured	\$0.0	\$0.0	\$236.0	\$236.0	\$237.3	\$237.4	\$237.7	\$227.6	\$214.4	\$202.1
Medicaid	\$105.4	\$108.9	\$135.9	\$141.1	\$147.2	\$154.3	\$161.3	\$169.7	\$178.7	\$188.3
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Group federal \$\$										
Insured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take Medicaid coverage	\$10.7	\$13.1	\$31.0	\$35.7	\$40.4	\$45.1	\$49.3	\$53.5	\$57.2	\$60.8
Refuse coverage	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take individual coverage	\$0.0	\$0.0	\$30.6	\$30.6	\$31.2	\$31.1	\$31.0	\$30.2	\$30.2	\$30.1

continued

Table 3. *Continued*

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Total federal \$\$										
Insured	\$0.0	\$0.0	\$266.7	\$266.6	\$268.5	\$268.5	\$268.7	\$257.8	\$244.6	\$232.3
Medicaid	\$116.1	\$122.0	\$166.9	\$176.8	\$187.6	\$199.4	\$210.6	\$223.1	\$235.9	\$249.1
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Total	\$116.1	\$122.0	\$433.6	\$443.5	\$456.0	\$467.9	\$479.4	\$480.9	\$480.5	\$481.3
Net impact										
Insured	\$0.0	\$0.0	\$20.1	\$22.4	\$23.2	\$25.8	\$31.4	\$29.3	\$24.0	\$19.2
Medicaid	\$0.0	\$0.0	\$12.5	\$13.6	-\$14.3	\$15.2	-\$16.3	\$16.3	\$16.0	\$15.4
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Total	\$0.0	\$0.0	\$7.5	\$8.8	\$8.9	\$10.6	\$15.1	\$13.0	\$8.0	\$3.8
										\$75.8

UNCORRECTED PROOF

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1 To complete this simulation, we assumed: (1) Individuals who are not
2 eligible for Medicaid currently and are eligible for exchange subsidies starting
3 at 100 percent FPL will consider taking an exchange subsidy with a minimal
4 premium required for enrollment; (2) States that do not offer an exchange will
5 have a federal exchange available to their citizens and the federal exchange
6 will be able to route subsidies to citizens in these states; (3) States offering Medi-
7 caid will use the same health plans they use today (e.g., the State of Washing-
8 ton uses private insurer Molina as a Medicaid health plan) through 2021; (4)
9 Insurance exchange premiums will grow at a higher rate than Medicaid costs
10 because of the frequent budget constraints imposed on state Medicaid budgets
11 to meet balanced budget requirements; (5) Prior to the SCOTUS ruling, states
12 would have auto-enrolled anyone below 134 percent of FPL in a federally
13 financed Medicaid plan and not allowed anyone below 134 percent FPL to
14 enroll in a state or federal exchange for private insurance coverage.

15 In Tables 2 and 3, we report the coverage and federal cost impacts,
16 respectively, of ACA pre-SCOTUS to a scenario where the Governors of six
17 states have openly declared that they will not expand their Medicaid pro-
18 grams. These states all have Republican Governors and include Florida, Loui-
19 siana, Mississippi, Nebraska, South Carolina, and Texas. We predict a
20 7.9 million person increase in the number of uninsured by 2021, compared
21 with the pre-SCOTUS situation. Several million childless adults who are not
22 disabled or aged earn sufficient income to qualify for an exchange policy in the
23 six states choosing not to expand Medicaid. Medicaid coverage will drop by
24 4.4 million lives in 2014 compared with the pre-SCOTUS ruling. The number
25 of people taking up private insurance coverage will increase by 3.2 million in
26 2014. The cost of private insurance displacing the Medicaid expansion will be
27 an additional \$7.5 billion in 2014 and a total of \$75.8 billion from 2014 to
28 2021. If citizens in the states refusing to expand their Medicaid programs could
29 accept federal subsidies for private health insurance, the federal cost will be
30 greater and the loss of coverage larger than our baseline estimate.

31 We considered an alternative scenario where only states with Medicaid
32 benefits already at 100 percent of the FPL or above would take the expansion.
33 The rationale for this scenario is that each of these states already is investing
34 state revenue to support an expanded Medicaid program. As a result, the
35 ACA offers these states financial relief from funding their expanded programs.
36 These states include the prototype state for ACA, Massachusetts, with 133
37 percent FPL qualification as well as Minnesota, the state with highest FPL
38 threshold of more than 200 percent. The coverage and cost results of this sim-
39 ulation are presented in Tables 4 and 5, respectively. Regarding coverage,

Table 4: Pre- and Post-Supreme Court of the United States (SCOTUS) PPACA Coverage Impact (Millions) Only States with Medicaid Federal Poverty Level $\geq 100\%$ Take the Expansion

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Pre-SCOTUS ruling										
Individual coverage										
Insured	32.6	31.3	45.0	46.9	47.0	45.3	43.7	42.0	40.4	38.9
Medicaid	31.7	32.1	41.9	42.6	43.5	44.9	46.1	47.4	48.7	50.0
Uninsured	48.9	50.9	28.5	27.1	27.3	28.8	30.3	31.9	33.4	34.9
Group coverage										
Insured	152.6	152.4	144.9	137.7	131.8	132.2	132.8	133.8	135.1	136.8
Take Medicaid coverage	3.2	3.8	10.1	11.5	12.7	13.8	14.7	15.6	16.2	16.9
Refuse coverage	5.9	7.2	5.5	5.9	6.3	6.8	7.2	7.4	7.3	7.1
Take individual coverage	0.1	0.1	4.5	11.7	17.6	17.3	17.0	16.8	16.5	16.2
Total coverage										
Insured	185.4	183.7	194.4	196.2	196.3	194.7	193.6	192.6	192.1	191.9
Medicaid	34.9	35.9	52.0	54.1	56.2	58.7	60.9	63.0	65.0	66.8
Uninsured	54.8	58.1	34.1	33.0	33.6	35.6	37.5	39.3	40.7	42.0
Total	275.0	277.7	280.5	283.3	286.1	289.0	291.9	294.8	297.8	300.7
Post-SCOTUS ruling										
Individual coverage										
Insured	32.6	31.3	49.5	50.4	49.2	46.2	43.2	39.7	36.4	33.3
Medicaid	31.7	32.1	33.4	34.4	35.8	37.4	39.0	41.0	43.1	45.1
Uninsured	48.9	50.9	32.6	31.8	32.8	35.4	37.9	40.6	43.1	45.4
Group coverage										
Insured	152.6	152.4	147.2	140.8	135.8	137.1	138.6	140.3	142.3	144.5
Take Medicaid coverage	3.2	3.8	6.8	6.9	6.9	6.8	6.7	6.7	6.7	6.7
Refuse coverage	5.9	7.2	5.8	6.3	6.9	7.5	8.0	8.3	8.3	8.0
Take individual coverage	0.1	0.1	5.2	12.6	18.8	18.6	18.5	18.2	18.0	17.7

continued

Table 4. *Continued*

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Total coverage										
Insured	185.4	183.7	202.0	203.8	203.8	201.9	200.3	198.2	196.6	195.6
Medicaid	34.9	35.9	40.2	41.3	42.6	44.2	45.7	47.7	49.8	51.8
Uninsured	54.8	58.1	38.4	38.1	39.7	42.9	45.9	48.9	51.3	53.4
	275.0	277.7	280.5	283.3	286.1	289.0	291.9	294.8	297.8	300.7
Net impact										
Insured	0.0	0.0	7.5	7.6	7.5	7.1	6.7	5.7	4.5	3.7
Medicaid	0.0	0.0	-11.8	-12.8	-13.6	-14.4	-15.1	-15.3	-15.2	-15.1
Uninsured	0.0	0.0	4.3	5.1	6.1	7.3	8.4	9.6	10.6	11.4

Note. States taking expansion—MN, DC, ME, NJ, WI, CT, IL, VT, RI, NY, MA, TN, DE, MD, AZ, CA, CO, HI.

Table 5: Pre- and Post-Supreme Court of the United States (SCOTUS) PPACA Cost Impact (Billions) Only States with Medicaid Federal Poverty Level \geq 100% Take the Expansion

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Pre-SCOTUS ruling										
Individual federal \$\$										
Insured	\$0.0	\$0.0	\$219.7	\$218.9	\$220.1	\$217.5	\$211.8	\$203.0	\$194.6	\$187.0
Medicaid	\$105.4	\$108.9	\$143.8	\$149.4	\$155.6	\$163.3	\$171.1	\$179.3	\$187.9	\$196.6
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Group federal \$\$										
Insured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take Medicaid coverage	\$10.7	\$13.1	\$35.6	\$41.1	\$46.3	\$51.3	\$55.8	\$60.1	\$64.0	\$67.9
Refuse coverage	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take individual coverage	\$0.0	\$0.0	\$26.9	\$25.4	\$25.2	\$25.2	\$25.5	\$25.4	\$26.0	\$26.1
Total federal \$\$	\$0.0	\$0.0	\$246.6	\$244.2	\$245.3	\$242.7	\$237.3	\$228.5	\$220.6	\$213.1
Insured	\$116.1	\$122.0	\$179.4	\$190.5	\$201.9	\$214.5	\$226.9	\$239.4	\$251.9	\$264.4
Medicaid	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Total	\$116.1	\$122.0	\$426.0	\$434.7	\$447.1	\$457.3	\$464.2	\$467.9	\$472.5	\$477.5
Post-SCOTUS ruling										
Individual federal \$\$										
Insured	\$0.0	\$0.0	\$247.7	\$244.7	\$242.2	\$237.5	\$231.9	\$214.4	\$194.6	\$176.6
Medicaid	\$105.4	\$108.9	\$113.4	\$119.6	\$126.7	\$135.1	\$143.7	\$154.1	\$165.2	\$176.3
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Group federal \$\$										
Insured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take Medicaid coverage	\$10.7	\$13.1	\$24.0	\$24.6	\$24.9	\$25.1	\$25.2	\$25.5	\$25.9	\$26.6
Refuse coverage	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Take individual coverage	\$0.0	\$0.0	\$33.3	\$34.5	\$36.5	\$37.8	\$38.9	\$38.9	\$39.8	\$40.3

continued

Table 5. *Continued*

	2012	2013	2014	2015	2016	2017	2018	2019	2020	2021
Total federal \$\$										
Insured	\$0.0	\$0.0	\$281.0	\$279.2	\$278.7	\$275.2	\$270.8	\$253.3	\$234.4	\$216.9
Medicaid	\$116.1	\$122.0	\$137.4	\$144.2	\$151.6	\$160.2	\$168.9	\$179.5	\$191.1	\$202.9
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Total	\$116.1	\$122.0	\$418.5	\$423.4	\$430.3	\$435.4	\$439.7	\$432.9	\$425.4	\$419.9
Net impact										
Insured	\$0.0	\$0.0	\$34.4	\$35.0	\$33.4	\$32.5	\$33.5	\$24.8	\$13.7	\$3.9
Medicaid	\$0.0	\$0.0	-\$42.0	-\$46.3	-\$50.3	-\$54.4	-\$58.0	-\$59.9	-\$60.8	-\$61.5
Uninsured	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0	\$0.0
Total	\$0.0	\$0.0	-\$7.6	-\$11.3	-\$16.8	-\$21.9	-\$24.6	-\$35.1	-\$47.1	-\$57.7

Note: States taking expansion—MN, DC, ME, NJ, WI, CT, IL, VT, RI, NY, MA, TN, DE, MD, AZ, CA, CO, HI.

1 11.8 million fewer people are in Medicaid compared with participation by all
2 50 states in 2014. However, 7.5 more people are insured through private insur-
3 ance. This still leaves a net reduction in uninsured in 2014, but less potential
4 Medicaid crowd-out of the private market. By 2021, the total net difference in
5 uninsured has grown to 11.4 million as private insurance market take-up
6 erodes compared with 2014. With respect to cost, Table 5 shows that the net
7 federal cost will be \$7.6 billion less in 2014 due to less Medicaid expansion
8 and \$57.7 billion less by 2021 compared with fully implemented ACA. The
9 sum of net federal savings from 2014 to 2021 will be \$221.9 billion. Under this
10 scenario, states that already made a commitment to expand Medicaid cover-
11 age would get federal financing relief and states that chose not to expand to
12 100 percent would get fewer federal dollars for Medicaid and more uninsured.

15 CAVEATS

16
17 While we developed an improved model for this analysis, several caveats are
18 critical to note. The first is that we do not observe the uninsured or Medicaid
19 in our health plan choice model. Thus, we have to add intercept terms in our
20 prediction equations and calibrate the level of uninsured to match that
21 reported in the markets for both the individual and ESI populations.

22 The second caveat is that both the employer and individual market pre-
23 mium data are several years old and need to be inflation adjusted for this anal-
24 ysis. However, we feel confident making these adjustments because the plan
25 designs in our analysis are largely the same as when the plan choices were
26 observed and our premium estimates are based on claims expenditures with a
27 medical care inflation rate applied. The 2001 linked MEPS insurance compo-
28 nent and household survey is the oldest component of the analysis, but the
29 linked MEPS data have not been available since 2001.

30 The third caveat is that the estimated numbers of individuals enrolled in
31 plans from the simulations are actually summed probabilities of a person's
32 enrollment in different plans. For example, we do not predict that 100 actual
33 people will join a Bronze plan. Instead, we predict that 1,000 people have (on
34 average) a 10 percent probability of joining a Bronze plan (which sums to
35 100). This is what a plan choice model enables us to do. It also provides a plat-
36 form to predict changes in policy, but not to the point of saying that a person
37 will absolutely choose a particular health plan.

38 A fourth caveat is that we do not observe actual HSA plan choices in the
39 employer data. Instead, we used the low-option Health Reimbursement

1 Arrangement (HRA) design that later became the standard benefit design tem-
2 plate for an HSA. We are obtaining new plan choice data that includes HSAs
3 offered by a large employer to enhance the model in the future.
4

5 6 SUMMARY

7
8 Our simulation model provides a tool to gauge the national impact of federal
9 health reform. The application of the simulation predicts significant reduc-
10 tions in levels of uninsured following full implementation of the ACA in 2014.
11 Our approach employed novel characteristics that reflect an evolving insur-
12 ance market with greater demand for HSAs as well as considering the influ-
13 ence of health status on health plan choice.

14 Our simulation model predicted nontrivial impacts of the recent
15 Supreme Court decision on federal cost and the level of uninsured, depending
16 on whether a state decides to accept or decline the Medicaid expansion. Gen-
17 erally, the number of uninsured will rise and uninsured and federal costs will
18 fall as more states opt out. The one exception is our six-state model where the
19 consequence of a small population of uninsured entering private exchanges as
20 opposed to a Medicaid expansion has greater federal cost as well as more
21 uninsured. However, when we consider the effect of all states providing more
22 than 100 percent of the FPL for Medicaid accepting expansion and thus
23 defraying some of the states' own outlays, we predict a sizable reduction in the
24 number of covered lives resulting from the ACA as well as a substantially
25 lower federal cost.

26 At best, full implementation of the ACA will reduce the number of unin-
27 sured by more than 20 million. If achieved in 2014, this would be the largest
28 coverage expansion in recent U.S. history.
29

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32
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34 the U.S. Department of Health and Human Services (Contract HHS-
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37 *Disclosures:* None.

38 *Disclaimers:* None.
39

NOTES

1. National Federation of Independent Business v. Sebelius, 132 S. Ct. 2566 (2012).
2. Families USA (2012) recently released a 50-state estimate of the impact of ACA. However, this report did not create a separate estimate for the impact of the Medicaid expansion.
3. High-deductible health plans feature a large deductible coupled with a Health Savings Account (HSA) owned by the individual that can be used to pay for eligible medical expenses. The MMA made it possible for contributions to the HSA to be made on a tax-preferred basis. That is, contributions less than the size of the deductible are exempt from federal income taxes. If the contribution is made by an employer, it is exempt from Social Security taxes as well.
4. Regression coefficients for our plan choice model are available in Parente et al. (2007). We converted HMO copays to actuarially equivalent coinsurance rates to predict the probability of HMO enrollment.
5. Pauly and Herring (1999) have suggested that individual policies contain some degree of group experience rating and vice versa. According to Pauly and Herring, premiums in the individual market do not rise one-for-one with predictable expenses, and premiums in the group market have a positive association with predicted *individual* medical expenditures, contrary to the GER hypothesis. Notwithstanding these findings, we decided to use IER and GER as our rating assumptions because these methods are more tractable and because it is not clear how to combine them to form “mixed” ratings systems as suggested by Pauly and Herring.
6. See Pauly, Percy, and Herring (1999) for data on loading fees in the individual health insurance market.
7. The MEPS uses “establishment size” rather than employer size. The three size classes are less than 50 employees, 50–200, and more than 200. We assume the loading factors for these classes are 20, 15, and 10 percent, respectively.

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24 [or-leaning-against-expanded-medicaid-program](http://thehill.com/blogs/healthwatch/health-reform-implementation/236033-fifteen-governors-reject-or-leaning-against-expanded-medicaid-program) July 7, 2012
- 25 U.S. Congressional Budget Office. July 23, 2012. "Estimates for the Insurance Cover-
26 age Provisions of the Affordable Care Act Updated for the Recent Supreme
27 Court Decision." [accessed on November 19, 2012]. Available at <http://www.cbo.gov/publication/43472>
- 28 Weiner, J., B. Starfield, D. Steinwachs, and L. Mumford. 1991. "Development and
29 Application of a Population Oriented Measure of Ambulatory Care Case-Mix."
30 *Medical Care* 29: 452–72.

30 SUPPORTING INFORMATION

31
32 Additional supporting information may be found in the online version of this
33 article:

34
35 Appendix SA1: Author Matrix.
36
37
38
39

Author Query Form

Journal: HESR

Article: 12036

Dear Author,

During the copy-editing of your paper, the following queries arose. Please respond to these by marking up your proofs with the necessary changes/additions. Please write your answers on the query sheet if there is insufficient space on the page proofs. Please write clearly and follow the conventions shown on the attached corrections sheet. If returning the proof by fax do not write too close to the paper's edge. Please remember that illegible mark-ups may delay publication.

Many thanks for your assistance.

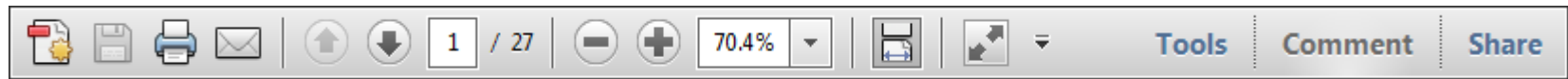
Query reference	Query	Remarks
1	AUTHOR: A running head short title was not supplied; please check if this one is suitable and, if not, please supply a short title of up to 40 characters that can be used instead.	
2	AUTHOR: KFF 2012 has been changed to Kaiser Family Foundation 2012a, 2012b so that this citation matches the Reference List. Please confirm that this is correct.	
3	AUTHOR: Tables 2A, 2B, 3A, and 3B are renumbered as Tables 2, 3, 4, and 5. Similar changes have also been made to text citations. Please check.	

USING e-ANNOTATION TOOLS FOR ELECTRONIC PROOF CORRECTION

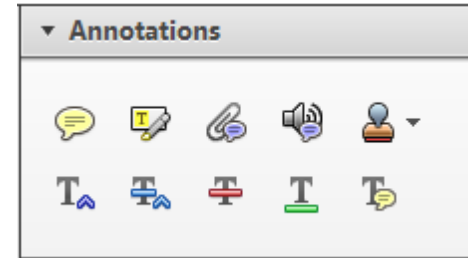
Required software to e-annotate PDFs: Adobe Acrobat Professional or Adobe Reader (version 7.0 or above). (Note that this document uses screenshots from Adobe Reader X)

The latest version of Acrobat Reader can be downloaded for free at: <http://get.adobe.com/uk/reader/>

Once you have Acrobat Reader open on your computer, click on the [Comment](#) tab at the right of the toolbar:



This will open up a panel down the right side of the document. The majority of tools you will use for annotating your proof will be in the [Annotations](#) section, pictured opposite. We've picked out some of these tools below:



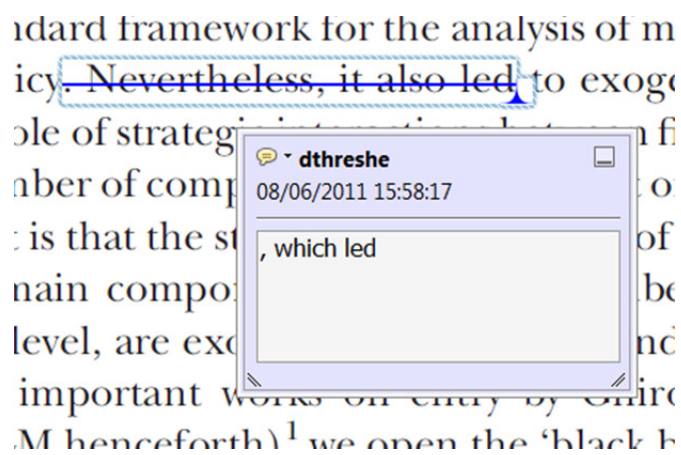
1. Replace (Ins) Tool – for replacing text.



Strikes a line through text and opens up a text box where replacement text can be entered.

How to use it

- Highlight a word or sentence.
- Click on the [Replace \(Ins\)](#) icon in the Annotations section.
- Type the replacement text into the blue box that appears.



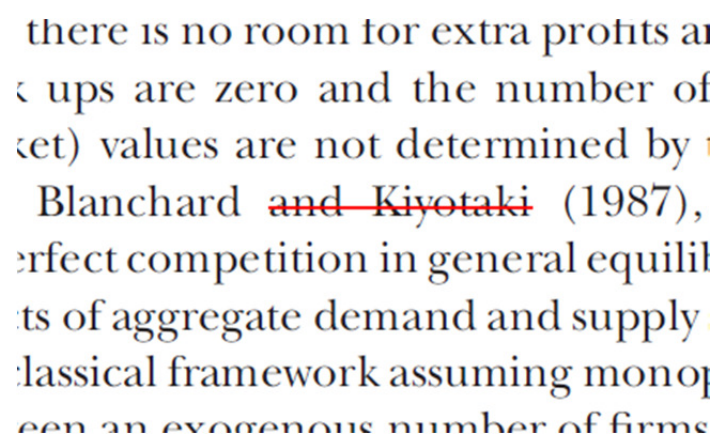
2. Strikethrough (Del) Tool – for deleting text.



Strikes a red line through text that is to be deleted.

How to use it

- Highlight a word or sentence.
- Click on the [Strikethrough \(Del\)](#) icon in the Annotations section.



3. Add note to text Tool – for highlighting a section to be changed to bold or italic.

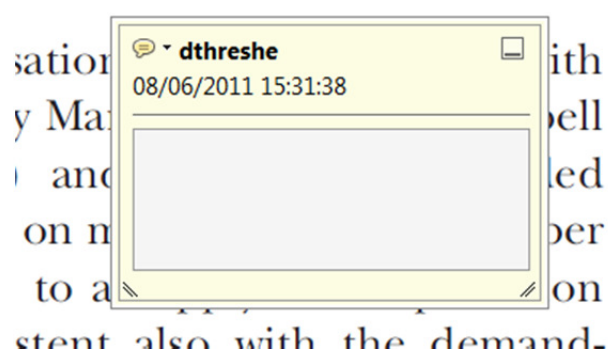


Highlights text in yellow and opens up a text box where comments can be entered.

How to use it

- Highlight the relevant section of text.
- Click on the [Add note to text](#) icon in the Annotations section.
- Type instruction on what should be changed regarding the text into the yellow box that appears.

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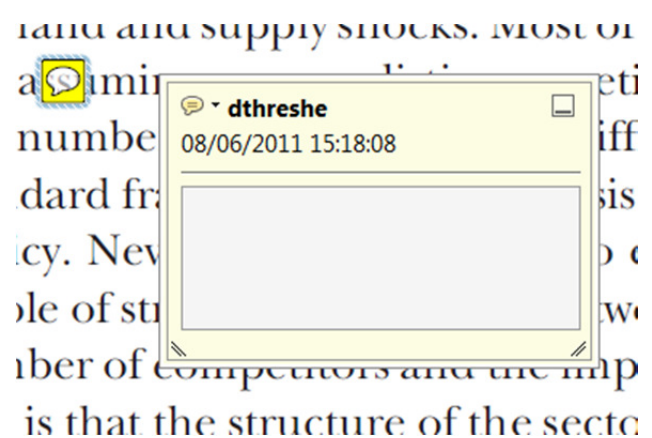
4. Add sticky note Tool – for making notes at specific points in the text.



Marks a point in the proof where a comment needs to be highlighted.

How to use it

- Click on the [Add sticky note](#) icon in the Annotations section.
- Click at the point in the proof where the comment should be inserted.
- Type the comment into the yellow box that appears.



USING e-ANNOTATION TOOLS FOR ELECTRONIC PROOF CORRECTION

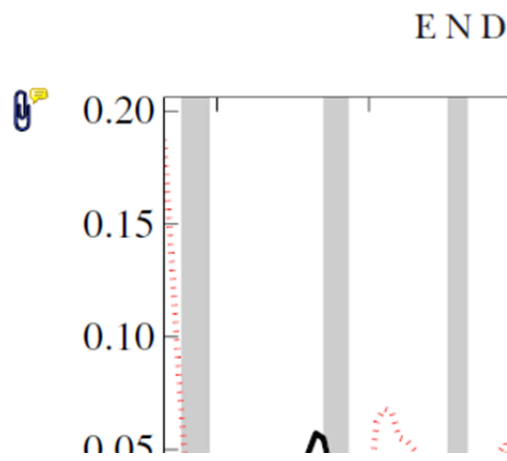
5. Attach File Tool – for inserting large amounts of text or replacement figures.



Inserts an icon linking to the attached file in the appropriate place in the text.

How to use it

- Click on the [Attach File](#) icon in the Annotations section.
- Click on the proof to where you'd like the attached file to be linked.
- Select the file to be attached from your computer or network.
- Select the colour and type of icon that will appear in the proof. Click OK.



6. Add stamp Tool – for approving a proof if no corrections are required.

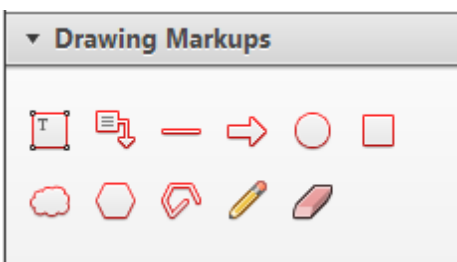


Inserts a selected stamp onto an appropriate place in the proof.

How to use it

- Click on the [Add stamp](#) icon in the Annotations section.
- Select the stamp you want to use. (The [Approved](#) stamp is usually available directly in the menu that appears).
- Click on the proof where you'd like the stamp to appear. (Where a proof is to be approved as it is, this would normally be on the first page).

of the business cycle, starting with the
 on perfect competition, constant return
 production. In this environment goods
 extra profits and the market
 he market. The New-Key
 otaki (1987), has introduced produc
 general equilibrium models with nomin
 ed and supply shocks. Most of this literat

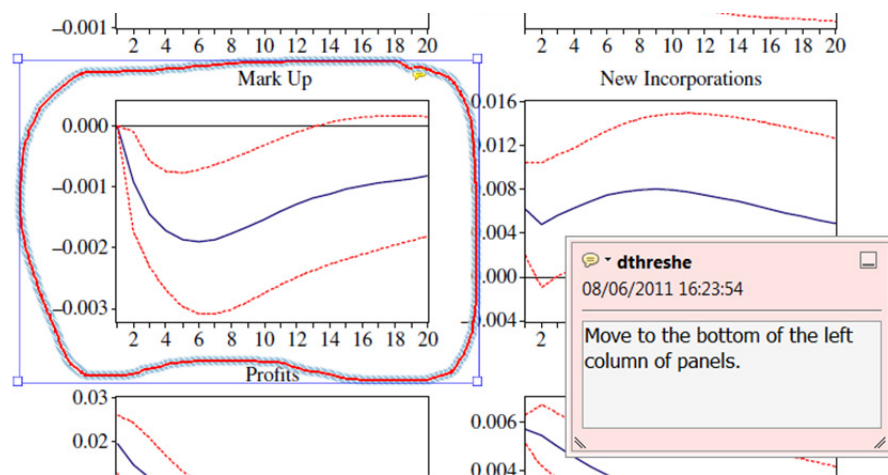


7. Drawing Markups Tools – for drawing shapes, lines and freeform annotations on proofs and commenting on these marks.

Allows shapes, lines and freeform annotations to be drawn on proofs and for comment to be made on these marks..

How to use it

- Click on one of the shapes in the [Drawing Markups](#) section.
- Click on the proof at the relevant point and draw the selected shape with the cursor.
- To add a comment to the drawn shape, move the cursor over the shape until an arrowhead appears.
- Double click on the shape and type any text in the red box that appears.



For further information on how to annotate proofs, click on the [Help](#) menu to reveal a list of further options:

