

1 **Title**

2 Mass evacuation and increases in long-term care benefits: lessons from the Fukushima
3 nuclear disaster

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5 **Running Title**

6 Mass evacuation and long-term care benefits

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26

27 **Abstract**

28 **Background**

29 Though mass evacuation may increase the need for long-term care (LTC) services, how
30 the need for LTC services increases and how the public LTC system affects it is not
31 well understood. We evaluated changes in public LTC benefits for the people living in
32 the mandatory evacuation areas established after the 2011 Fukushima nuclear disaster
33 and examined the roles of the universal LTC insurance system in Japan.

34

35 **Methods**

36 In order to evaluate the effect of the mandatory evacuation on LTC benefits, we
37 examined the trends of LTC benefits in the Fukushima evacuation group and the
38 nationwide non-evacuation group. We first decomposed per-elderly-individual benefits
39 at the municipality level into the LTC certification rate and per-certified-individual
40 benefits, and then implemented difference-in-differences analysis using these variables
41 as outcomes.

42

43 **Results**

44 Per-elderly-individual benefits significantly increased from 2012 onward in the
45 evacuation group, and this was explained by an increase in the certification rate rather
46 than in per-certified-individual benefits. Increases in per-elderly-individual benefits and
47 the certification rate in the post-disaster period were observed in all but the highest care
48 level, and the corresponding outcomes for the highest care level decreased immediately

49 after the disaster. We also found that the increase in the certification rate had been
50 mostly realized by an increase in the number of certified individuals.

51

52 **Conclusions**

53 The increase in LTC benefits can be associated with the impact of the increase in the
54 number of people newly certified to receive LTC benefits after the mandatory
55 evacuation. In order to cope with the increase in utilization of long-term care and
56 associated costs after disasters in aging societies, both formal long-term care services
57 and social support for informal care for evacuees should be considered important.

58

59 **Keywords:** Disasters, Long-Term Care, Displacement, Aging,
60 Difference-in-Differences

61

62

63 **1. Introduction**

64 Natural and manmade disasters, such as earthquakes, floods, and fires, often lead to
65 mass evacuation and/or relocation. In 2015, there were 27.8 million people newly
66 involved in evacuation or relocation caused by disasters worldwide [1]. Damage to
67 communities following a mass evacuation has become a social issue. As the number of
68 elderly people increases throughout the world, the health effects due to damage to
69 communities seem likely to become more severe, but these adverse effects, especially in
70 the long-term, are not clearly understood.

71

72 Mass evacuation in an aging society is exemplified by the case of the 2011 Fukushima
73 nuclear disaster, which occurred after the Great East Japan Earthquake and Tsunami on
74 March 11, 2011. Following a series of government evacuation orders, more than
75 160,000 citizens were evacuated from the areas surrounding the nuclear power plant [2].
76 In particular, the entire population of the nine municipalities closest to the damaged
77 nuclear plant, nearly 80,000 citizens, was forced to evacuate.

78

79 This mass evacuation not only raised the mortality of the elderly in the short term [3],
80 but may also have increased the utilization of public long-term care (LTC) services
81 [4-6]. However, *how* such an evacuation affects the utilization of LTC services is still
82 not well understood. Do the elderly people who had already been using LTC services
83 before the evacuation use more LTC services after it? Or do the elderly people who had
84 not been using any LTC services before the evacuation start doing so after they have
85 been evacuated? The objective of this study is to assess the factors behind the increase

86 in LTC utilization after an evacuation, which are important for policy makers in
87 countries that frequently experience natural disasters and subsequent evacuations. We
88 examined how the utilization of public LTC benefits changed after the evacuation
89 among citizens who had lived in the mandatory evacuation areas of the 2011 Fukushima
90 nuclear disaster using the difference-in-differences (DID) method.

91

92 **2. Materials and methods**

93 **2.1. Settings**

94

95 After the 2011 Fukushima nuclear disaster, evacuation instructions were issued by the
96 Japanese government on March 12, 2011 for a 20km area surrounding the nuclear plant
97 and in areas where the annual cumulative radiation dose was expected to exceed 20
98 mSv/year. As a result, eight whole municipalities (the villages of Iitate, Katsurao, and
99 Kawauchi and the towns of Namie, Futaba, Okuma, Tomioka, and Naraha) were
100 designated as mandatory evacuation areas. One additional municipality, the town of
101 Hirono, also decided to issue a mandatory evacuation order to all of its residents at the
102 direction of its mayor. As a result, a total of 78,768 residents were forced to evacuate.
103 According to the National Population Census, elderly residents (aged 65 and older)
104 accounted for 25.1% (19,792/78,768) of the population in these nine municipalities at
105 the time of the disaster. Parts of three municipalities (the cities of Minamisoma and
106 Tamura and the town of Kawamata) were also designated as mandatory evacuation
107 areas. In total, 164,865 Fukushima residents were evacuated.

108

109 In Japan, all people aged 65 and older have been insured by public LTC insurance since
110 2000. When an individual wants to make use of this insurance, municipalities conduct
111 interviews and surveys concerning his or her living situation. An initial assessment is
112 then used to assign the applicant to one of the eight LTC need levels (not certified,
113 support level 1–2, and care level 1–5). The final LTC need level is decided by the Care
114 Needs Certification Board, a committee of medical and other professionals. The
115 maximum amount of expenditure covered by the LTC insurance is fixed in accordance
116 with the applicant’s level as determined by this process [7].

117

118 **2.2. Data**

119

120 In order to evaluate the effect of the mandatory evacuation on LTC benefits, we
121 examined the LTC benefits in the eight “evacuation” municipalities and compared them
122 to those observed nationwide.[8]

123

124 We first define Q_{it} as the aggregated quantity of LTC benefits, which is measured as
125 the amount of LTC benefits per elderly individual for municipality i in fiscal year t .

126 Then we can decompose Q_{it} as follows [9]:

127

$$128 \quad Q_{it} \equiv \frac{\text{Total benefits}}{\# \text{ of elderly}} = \frac{\# \text{ of certified elderly}}{\# \text{ of elderly}} \frac{\text{Total benefits}}{\# \text{ of certified elderly}} \equiv C_{it} B_{it} ,$$

129

130 where C_{it} is the LTC certification rate (i.e. the ratio of elderly people certified to
131 receive LTC benefits to the total elderly population in question) and B_{it} is the amount

132 of LTC benefits per certified individual. We call Q “per-elderly-individual benefits”, C
133 “certification rate”, and B “per-certified-individual benefits” and examine these three
134 variables.

135

136 In addition, because the certified elderly are categorized into seven care levels (support
137 level 1-2 and care level 1–5), we also investigate LTC benefits disaggregated by care
138 level. For example, we use disaggregated outcomes for care level 1 as

139

$$140 \quad Q_{it}^{Care1} \equiv \frac{\text{Total benefits for care level 1}}{\text{\# of elderly}},$$

141

$$142 \quad C_{it}^{Care1} \equiv \frac{\text{\# of certified elderly for care level 1}}{\text{\# of elderly}}, \text{ and}$$

143

$$144 \quad B_{it}^{Care1} = \frac{\text{Total benefits for care level 1}}{\text{\# of certified elderly for care level 1}}.$$

145

146 We can also define disaggregated outcomes for support levels 1-2 and care levels 2-5 in
147 the same manner. See Table 1 for the descriptive statistics of the aggregated and
148 disaggregated outcome variables. All the variables are constructed from the Status
149 Report on the Long-term Care Insurance collected by the Ministry of Health, Labour
150 and Welfare of Japan. This report provides municipality-level data based on
151 administrative data about all the benefits dispensed through public LTC insurance for
152 each municipality in an open public database. From this public database, we extracted
153 the annual data on LTC benefits, the number of elderly people certified to receive LTC
154 benefits, and the population over 65 years of age in each municipality. Some

155 municipalities organized or joined inter-municipality insurance coalitions, and some
156 municipalities merged during the sample period (2007-2014). In these cases, we
157 aggregated the LTC benefits data and population data based on municipalities as they
158 were constituted in 2014.

159

160 When it comes to the number of certified elderly individuals and the population over 65
161 years of age, these values are measured in the last month of the fiscal year (March of the
162 following calendar year because the Japanese fiscal year starts in April). LTC benefits,
163 which are measured in so-called “benefit points” (1 point = 10 yen in a standard area),
164 are in turn aggregated from March to February of the following calendar year. This is
165 somewhat confusing because the first month is March, not April (the first month of the
166 fiscal year). This is advantageous in our case, however, because the disaster occurred on
167 the 11th of March, 2011, and therefore the survey of the year for 2011 consists of data
168 concerning only post-disaster months (March 2011-February 2012).

169

170 It should be noted that the residential status (registered place of residence) of citizens
171 remained the same after the mandatory evacuation if they wanted it to, and they were
172 given financial support unless they changed it [10]. Most of the LTC benefits for these
173 citizens were therefore financed by these nine municipalities before, during, and after
174 the evacuation. These circumstances enable us to assess the impact of the mandatory
175 evacuation on LTC benefits.

176

177 **2.3. Methods**

178

179 In order to investigate how this evacuation has affected the utilization of LTC services,
180 we examine how per-elderly-individual benefits (Q), certification rate (C), and
181 per-certified-individual benefits (B) have changed over time in the evacuation group (i.e.
182 those who had been living in evacuation areas before the disaster) by implementing
183 simple difference-in-differences (DID) estimations. The estimation model is expressed
184 as follows:

185

$$186 \quad Y_{it} = \pi_i + \theta_t + \sum_{\tau \neq 2009} \beta_{\tau}(T_{\tau} \cdot D_i) + \varepsilon_{it}, \quad (1)$$

187

188 where Y_{it} is either Q_{it} , C_{it} , or B_{it} , π_i is a municipality fixed effect, θ_t is a time
189 fixed effect, T_{τ} is a time dummy variable that takes one if $t = \tau$, D_i is a “treatment”
190 dummy variable that takes one if municipality i is entirely included in evacuation areas
191 in Fukushima, and ε_{it} is a random error.

192

193 As explained in Section 2.1, the entirety of each of the eight municipalities is included
194 in the “evacuation areas” and one municipality issued a mandatory evacuation order to
195 all of its residents, resulting in nine “treated” municipalities in our sample. Note that we
196 exclude municipalities with partial evacuation orders or without a mandatory evacuation
197 order in the Fukushima region (Fukushima prefecture) and in tsunami-affected coastal
198 areas from the “control” group ($D_i = 0$): these municipalities presumably have been
199 affected by the nuclear disaster and/or tsunami in 2011 and are not suitable for
200 constructing the counterfactual trends to the evacuation areas.

201

202 The parameter of interest is β_τ , which reflects a differential trend in the outcome
203 variable for the treated (evacuation) group at $t = \tau$, compared with a corresponding
204 trend for the control (non-evacuation) group. Because we exclude the time dummy
205 variable at $t = 2009$, β_τ captures $E(Y_{i\tau} - Y_{i2009} | D_i = 1) - E(Y_{i\tau} - Y_{i2009} | D_i = 0)$,
206 which is identical to a conventional DID parameter. Note that fiscal year 2009 is the
207 latest pre-disaster year when the complete annual data is available for the evacuation
208 group. That is, the data for fiscal year 2010 is not available because the data for
209 February 2011, which was supposed to be included in fiscal year 2010, is not available
210 due to administrative disorder during the evacuation period (the Fukushima nuclear
211 disaster occurred in March 2011).

212

213 As is now common in DID and event study literature [11, 12], in the pre-disaster period,
214 $\tau < 2009$, β_τ serves as a placebo parameter whose estimated value should be near zero
215 if no differential trends exist between treated and control municipalities. In the
216 post-disaster period, $\tau \geq 2011$, β_τ captures how the trend of mean outcomes for the
217 treated municipalities deviates from the trend for the control municipalities given that
218 no differential trends exist in the pre-disaster period.

219

220 Before examining estimation results, Figure 1 shows the trends of outcome variables on
221 average for the treated group and the control group. These graphs indicate that in the
222 pre-disaster period (2007-2009) the trends were more or less similar between the
223 evacuation and control groups, indirectly validating the common trend assumption of
224 DID estimation. In the post-disaster period (2011-2014) they significantly differed. In

225 2011, the year of the disaster and evacuation, the certification rate (C) sharply increased
226 in the evacuation group, whereas per-certified-individual benefits (B) sharply decreased,
227 resulting in a modest increase in per-elderly-individual benefits (Q). Since 2012, the
228 certification rate (C) has remained high, and per-certified-individual benefits (B) have
229 more or less recovered to their pre-disaster trend. As a result, per-elderly-individual
230 benefits (Q) have sharply increased.

231

232 **3. Results**

233 Figure 2 presents DID estimation results based on equation (1). The three graphs in this
234 figure show estimation results using different aggregated outcomes, namely
235 per-elderly-individual benefits (Q), certification rate (C), and per-certified-individual
236 benefits (B) respectively. These outcome variables are based on all LTC benefits and
237 certified individuals.

238

239 The results of this analysis imply that per-elderly-individual benefits (Q , left graph)
240 increased from 2012 onward, and that this was explained by an increase in the
241 certification rate (C , center graph) rather than per-certified-individual benefits (B , right
242 graph). DID estimates for the certification rate are positive and statistically significant
243 immediately after the evacuation, and the magnitude of the estimates implies that the
244 evacuation has increased the certification rate by around 6 percentage points. DID
245 estimates for per-certified-individual benefits decreased sharply in 2011 (and modestly
246 in 2012), and this in turn is presumably a result of the disaster (earthquake and tsunami)
247 and the ensuing evacuation necessitated by the nuclear plant accident. Note also that

248 placebo estimates in the pre-disaster period are around zero for all three outcomes,
249 suggesting that the “parallel trends” assumption of DID seems to be plausible in the
250 post-disaster period.

251

252 Figure 3 shows DID estimates for disaggregated outcomes by care level (see appendix
253 Figure A.1-A.2 for the same estimation results with separated graphs). Firstly, the
254 results of this analysis present a markedly different pattern in DID estimates for care
255 level 5 outcomes. Negative estimates of the care level 5 certification rate (C) after 2011
256 suggest that some elderly people in this category may not have survived the disasters
257 and evacuation (center graph). The estimates in this category gradually approach zero
258 from 2011 to 2014, however, implying that the care level 5 certification rate was
259 returning to its pre-evacuation level. DID estimates for care level 5
260 per-certified-individual benefits (B) are also strongly negative in 2011 and 2012. This is
261 probably due to the evacuation, although the placebo estimates in 2008 fluctuate
262 unstably (right graph). The estimates in this category also approached zero in 2013 and
263 2014.

264

265 Secondly, when it comes to DID estimates for support levels 1-2 and care levels 1-4 in
266 Figure 3, these show more or less similar trends that are consistent with the DID results
267 in Figure 2. There are, however, two points worth mentioning. First, for these
268 support/care levels, DID estimates for per-elderly-individual benefits (Q) and the
269 certification rate (C) are always positive and almost always significantly different from
270 zero from 2012 onward (left and center graphs. See 95% C.I. in appendix Figure A.1.
271 and A.2.). The contribution to the aggregated positive impact on per-elderly individual

272 benefits is higher in care levels 1-3 than the other levels, but the DID estimates for the
273 certification rate suggest that the number of certified individuals has been sharply rising
274 in all care-need categories but care levels 4 and 5. Another interesting finding is that
275 DID estimates for per-certified-individual benefits (B) are also positive for 2012-2014
276 (right graph), although the 95% C.I. is consistently above zero only for care levels 1 and
277 2 (appendix Figure A.1, A.2).

278

279 In summary, the Fukushima disaster and the ensuing evacuation increased the overall
280 certification rate but not overall per-certified-individual benefits (Figure 2). If we look
281 at disaggregated outcomes by care level, the disaster and evacuation decreased the
282 certification rate and per-certified-individual benefits for the highest care level, and
283 increased the certification rates for the other care levels and the per-certified-individual
284 benefits for some care levels (Figure 3).

285

286 Finally, given the finding that the positive impact of the evacuation on LTC benefits
287 was mainly owing to an increase in the certification rate, it is worth looking into what
288 drove this increase in this group. We therefore compared the trends of the numerator
289 and the denominator of the certification rate (i.e. the number of certified people and the
290 total number of elderly people) between the evacuation and the control groups. Figure 4
291 clearly shows the main source of the increase in the certification rate in the evacuation
292 group is an increase in the number of certified elderly (the numerator), although the
293 number of those aged 65+ (the denominator) also decreased modestly in 2011.

294

295 **4. Robustness checks**

296 Figure 2 clearly shows that the increase in LTC benefits per elderly individual after the
297 disaster is caused by the increase in the certification rate, not in LTC benefits per
298 certified individual. In this section we provide the results of several robustness checks
299 on these primary findings. We implement robustness checks by estimating the same
300 equation (1) using different combinations of the following two estimation settings.

301

302 First, we use two different weighting schemes for estimation: (1) not weighting
303 observations as in our baseline estimation (i.e. ordinary least squares regression: OLS)
304 and (2) weighting observations by the number of elderly individuals for the outcomes of
305 Q and C and by the number of certified individuals for the outcome of B (i.e. weighted
306 least squares regression: WLS). There are pros and cons for weighting observations in
307 aggregated data, but our purpose in using different weights is to ensure that our main
308 findings are robust to different weighting schemes [13].

309

310 Second, we check the sensitivity of DID estimates by using three different samples (i.e.
311 three different control groups). In the baseline estimation we used all the Japanese
312 municipalities except for partial-evacuation and non-evacuation municipalities in
313 Fukushima prefecture and tsunami-affected coastal areas. In this section we use the
314 following three different samples: the same municipalities as in the baseline estimation
315 (the baseline sample), the municipalities in the baseline sample whose outcome values
316 (*all of Q , C and B*) are within the minimum and maximum of the outcome values of the
317 treated municipalities in all the three pre-disaster years (the trimmed sample), and
318 evacuation municipalities and non-evacuation municipalities in Fukushima prefecture
319 (the Fukushima sample).

320

321 In the case of the trimmed sample, 399 out of 1,496 control municipalities in the
322 baseline sample are kept in the sample. This trimming procedure is somewhat arbitrary,
323 but it enables us to exclude municipalities whose outcome values are far away from and
324 not comparable to those of the evacuation municipalities in the pre-disaster period.
325 Appendix Figure A.3 shows that this trimming procedure does in fact make the
326 averaged outcomes of the control areas quite similar to those of the evacuation areas in
327 the pre-disaster period.

328

329 In the Fukushima sample, we use 47 non-evacuation municipalities in Fukushima,
330 which are excluded from the baseline sample, as an alternative control group. We keep
331 partial-evacuation municipalities in Fukushima out of the control group. As we already
332 discussed, non-evacuation municipalities in Fukushima may not be an appropriate
333 control group. It is nevertheless worthwhile to compare the evacuation municipalities in
334 Fukushima with the non-evacuation municipalities in Fukushima because of their
335 geographic and socio-economic similarities in the pre-disaster period.

336

337 Figure 5 provides the results of our robustness checks. Because we use two weighting
338 schemes (OLS and WLS) and three samples (baseline, trimmed, and Fukushima
339 samples), we implemented six, including one baseline, DID estimations for each
340 outcome. Some estimated coefficients are different from the baseline DID estimation
341 (i.e. OLS with the baseline sample) but overall tendencies are similar to the baseline
342 analysis for all three outcomes.

343

344 **5. Discussion**

345 It has been reported that LTC benefits have increased in the peripheral area of the
346 damaged nuclear plant [14]. Our findings suggest that the increase in LTC benefits can
347 be associated with the impact of the increase in the number of people newly certified to
348 receive LTC benefits after the mandatory evacuation.

349

350 There are several reasons for this. First, as the communities in the mandatory evacuation
351 zones were severely damaged, the amount of informal care provided by young members
352 of the community seems to have declined after the disaster. Second, evacuation could
353 have worsened ADL, dementia, and other health problems among the elderly population
354 [15]. Our study was consistent with previous studies demonstrating that the number of
355 people relying on public LTC increased after the disaster [4-6]. Third, residents in the
356 mandatory evacuation areas were exempted from copayment according to the
357 compensation program. This financial support could have increased the demand for
358 long-term care [16].

359

360 In addition, it is possible that with their greater mobility and resilience the healthy
361 elderly population in a community who do not require long-term care may be more
362 likely than the rest of the population to change their resident registration from the
363 evacuation areas to other areas even though it may lead to the cessation of financial
364 support. Our analysis, however, found only a modest decline in the number of elderly
365 people in the evacuation group.

366

367 Another important finding is that both the certification rate and the
368 per-certified-individual benefits for the highest care level (care level 5) decreased
369 immediately after the disaster (2011), while the corresponding outcomes for the lower
370 care levels tended to gradually increase after the evacuation. Such heterogeneous effects
371 on the evacuated elderly suggest that special care soon after the disaster is particularly
372 important for the most vulnerable group, whereas increasing LTC needs among the
373 elderly due to the evacuation are of greater importance from a longer perspective.

374

375 This study demonstrates the possibility that an increase in the utilization of LTC
376 services immediately after a mass evacuation can be attributed to increases in people
377 newly certified to receive LTC benefits. This implies that universally provided LTC
378 insurance might have worked as a quick buffer that mitigates some of the negative
379 impacts of evacuation, such as the loss of informal care and evacuation-related health
380 deterioration [17]. At the same time, it highlights the importance of the question of how
381 such exceptional costs incurred by evacuations are to be financed as a policy issue to be
382 addressed in the context of the current municipality-based financing system of Japanese
383 LTC insurance. In order to cope with the increase in need for long-term care and
384 associated costs after disasters in aging societies, both formal long-term care services
385 and social support for informal care for evacuees should be considered important.

386

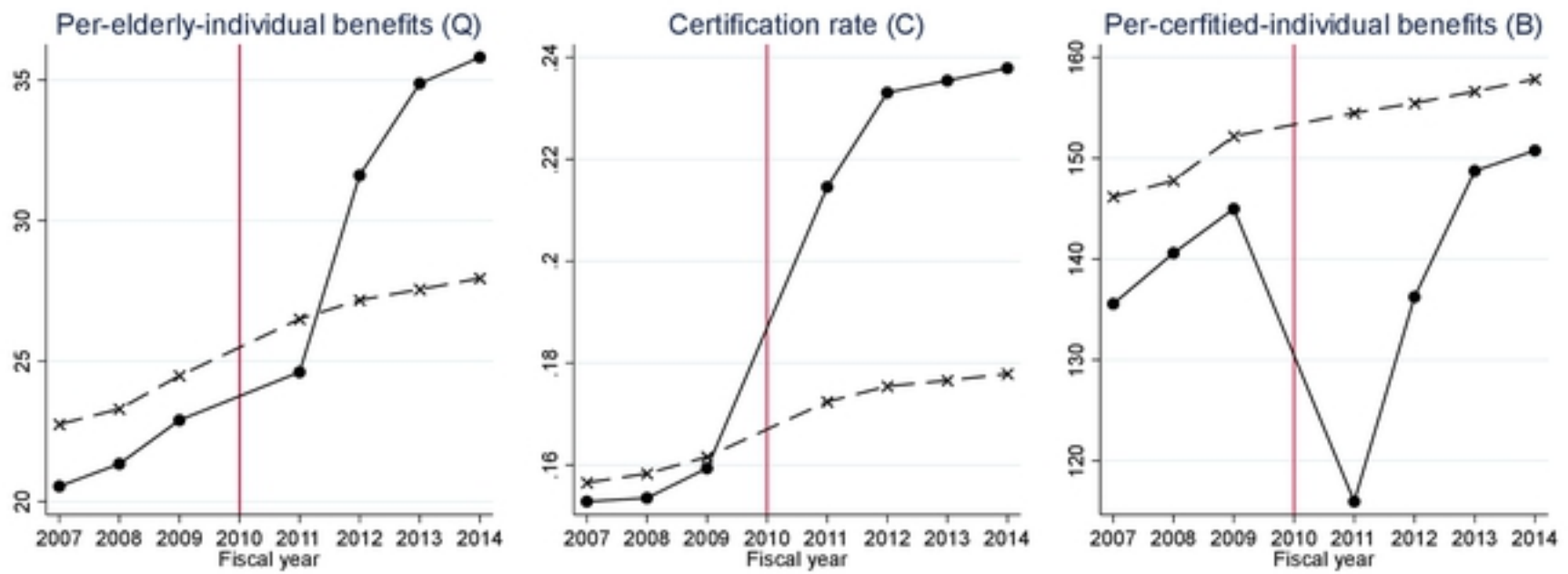
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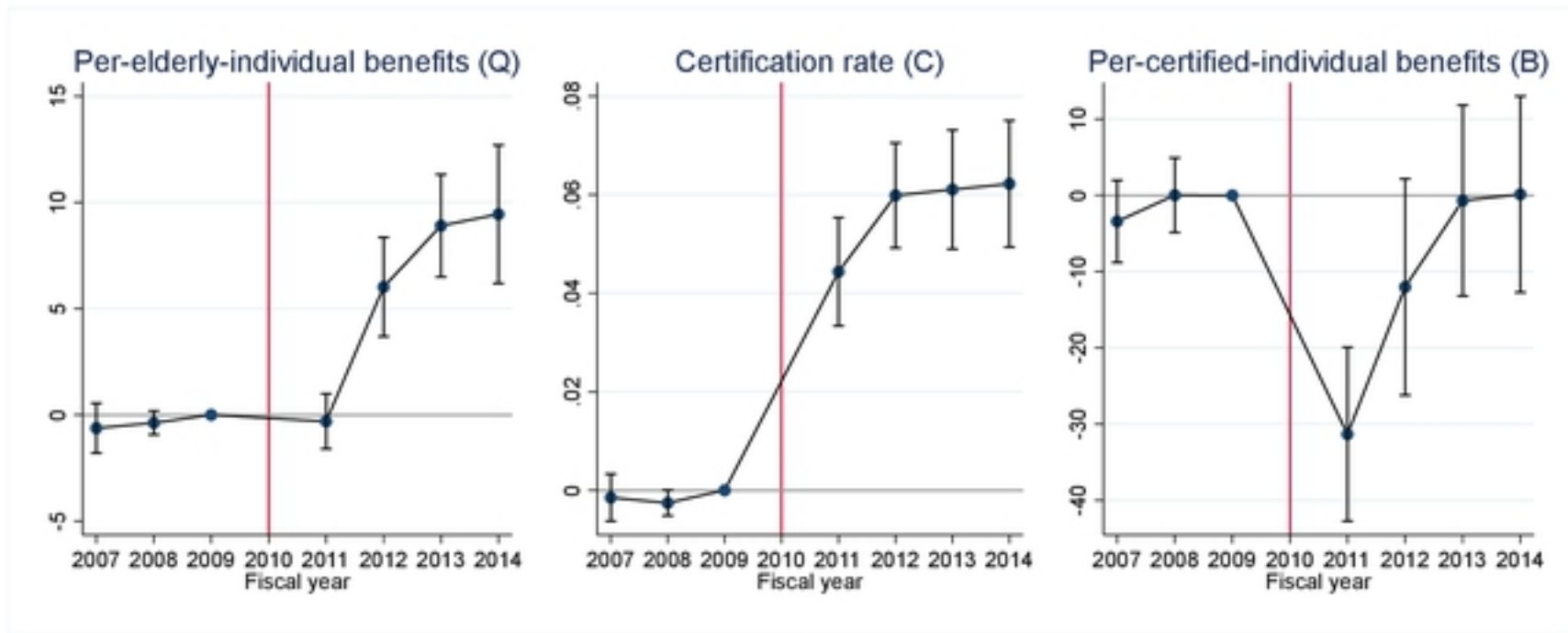
VARIABLES	Sample size	Mean	Min	Max
Per-elderly-individual benefits (all)	10,535	25.679	0.000	66.630
Per-elderly-individual benefits (support level 1)	10,535	0.402	0.000	2.330
Per-elderly-individual benefits (support level 2)	10,535	0.892	0.000	3.270
Per-elderly-individual benefits (care level 1)	10,535	3.018	0.000	21.250
Per-elderly-individual benefits (care level 2)	10,535	4.358	0.000	35.917
Per-elderly-individual benefits (care level 3)	10,535	5.421	0.000	20.333
Per-elderly-individual benefits (care level 4)	10,535	5.898	0.000	19.000
Per-elderly-individual benefits (care level 5)	10,535	5.692	0.000	23.552
Certification rate (all)	10,535	0.169	0.037	0.402
Certification rate (support level 1)	10,535	0.020	0.000	0.097
Certification rate (support level 2)	10,535	0.022	0.000	0.084
Certification rate (care level 1)	10,535	0.031	0.000	0.250
Certification rate (care level 2)	10,535	0.030	0.000	0.154
Certification rate (care level 3)	10,535	0.025	0.000	0.074
Certification rate (care level 4)	10,535	0.022	0.000	0.069
Certification rate (care level 5)	10,535	0.020	0.000	0.065
Per-certified-individual benefits (all)	10,535	152.9	0.0	474.0
Per-certified-individual benefits (support level 1)	10,476	20.2	0.0	112.0
Per-certified-individual benefits (support level 2)	10,498	41.1	0.0	192.8
Per-certified-individual benefits (care level 1)	10,524	98.2	0.0	417.0
Per-certified-individual benefits (care level 2)	10,531	149.1	0.0	862.0
Per-certified-individual benefits (care level 3)	10,528	221.2	0.0	691.0
Per-certified-individual benefits (care level 4)	10,525	264.6	0.0	881.0
Per-certified-individual benefits (care level 5)	10,525	287.8	0.0	732.0

Table 1: Descriptive statistics. Note that data for per-certified-individual benefits in some care levels are missing when the number of certified individuals in these care levels are is zero. In addition, the maximum value of per-certified-individual benefits for care level 2 appears to be unusually high, but this is due to the fact that this variable is calculated as the ratio of annual LTC benefits to the number of certified elderly individuals at a certain point. Thus this variable tends to fluctuate in municipalities with small numbers of certified elderly individuals. Source: The Survey of Long-term Care Benefit Expenditures. The unit of “benefit” is 1,000 points (i.e. 10,000 JPY or around 100 USD in standard areas).



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Figure 1: Time trends of long-term care benefits in the Fukushima mandatory evacuation group and the non-evacuation control group. Per-elderly-individual benefits: long-term-care benefits per elderly person. Certification rate: the percentage of people aged 65 and older who were certified to receive long-term care services. Per-certified-individual benefits: long-term care benefits per certified individual. The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas).



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Figure 2: Difference-in-differences (DID) estimates and their 95% confidence intervals. Confidence intervals are calculated by clustered robust standard errors clustered by municipality. All the DID estimates are estimates for β_τ in equation (1). The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas).

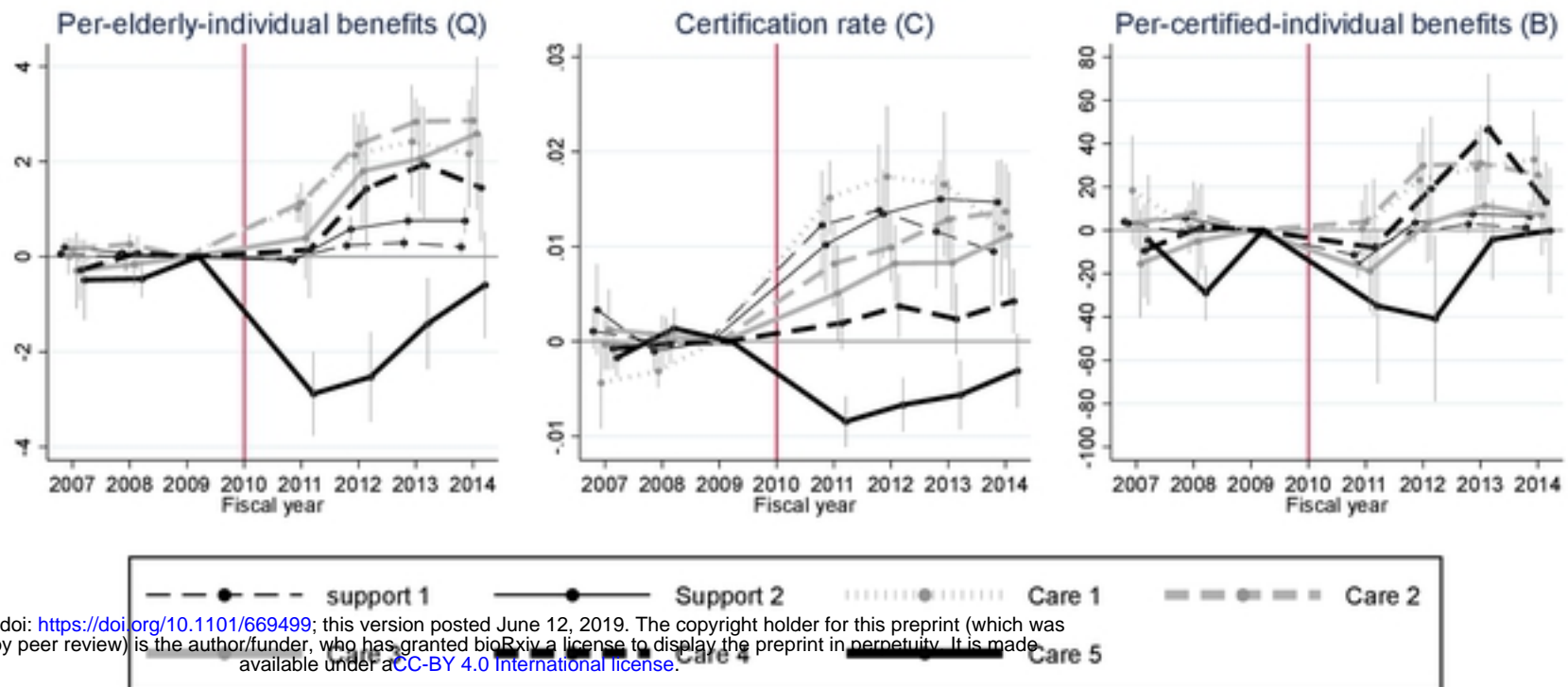
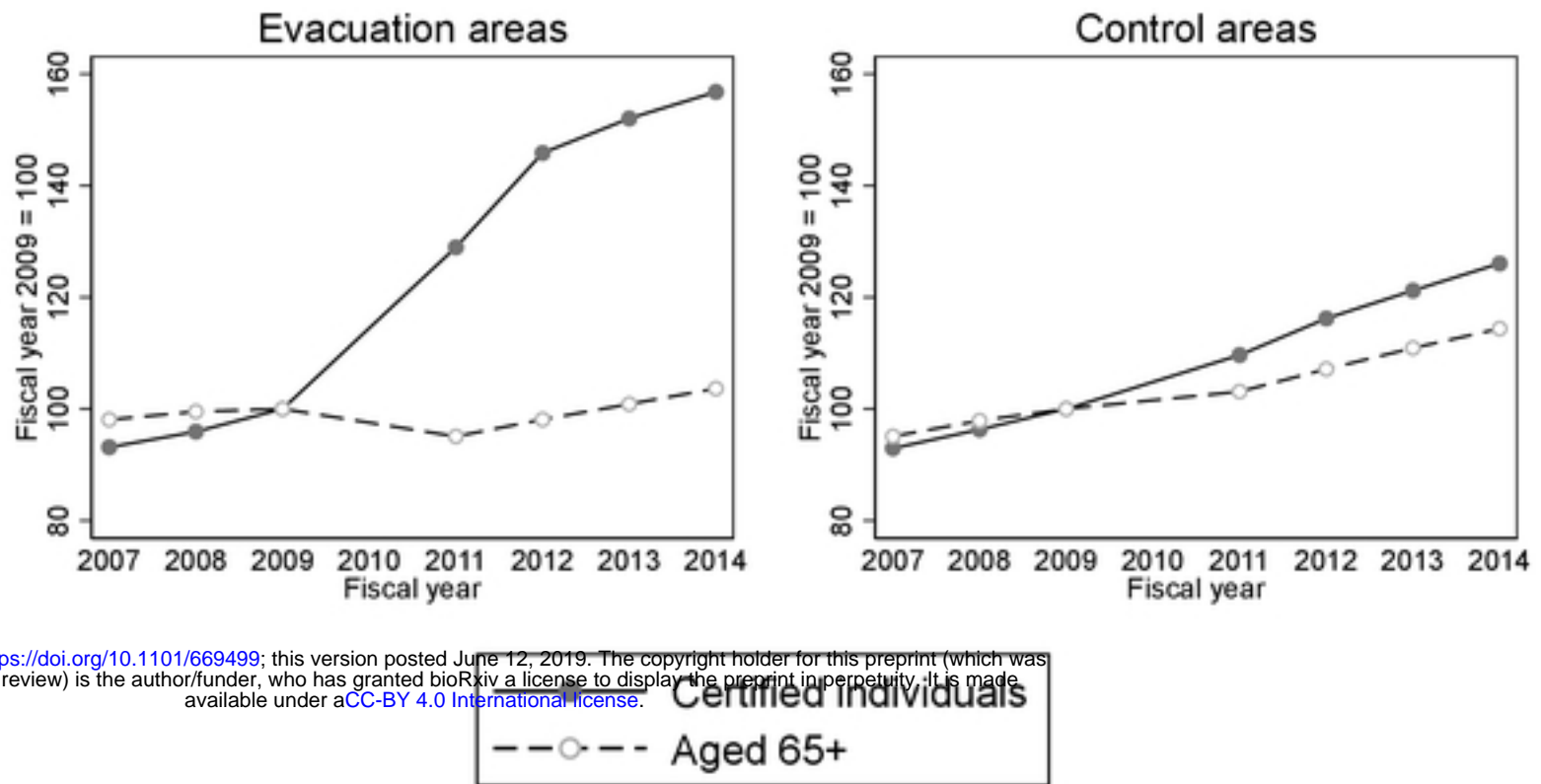
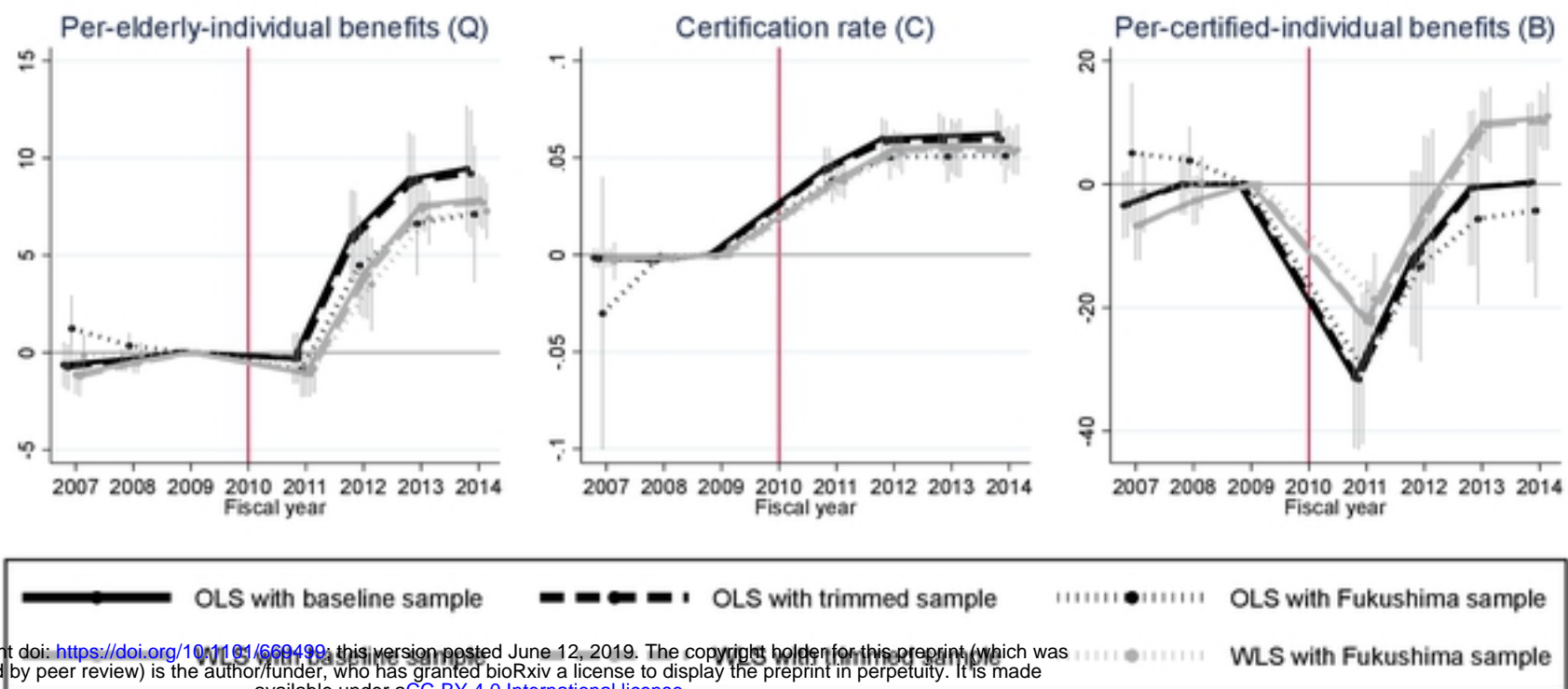


Figure 3: Difference-in-differences (DID) estimates for each care level and their 95% confidence intervals. Confidence intervals are calculated by clustered robust standard errors clustered by municipality. All DID estimates are estimates for β_τ in equation (1). The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas).



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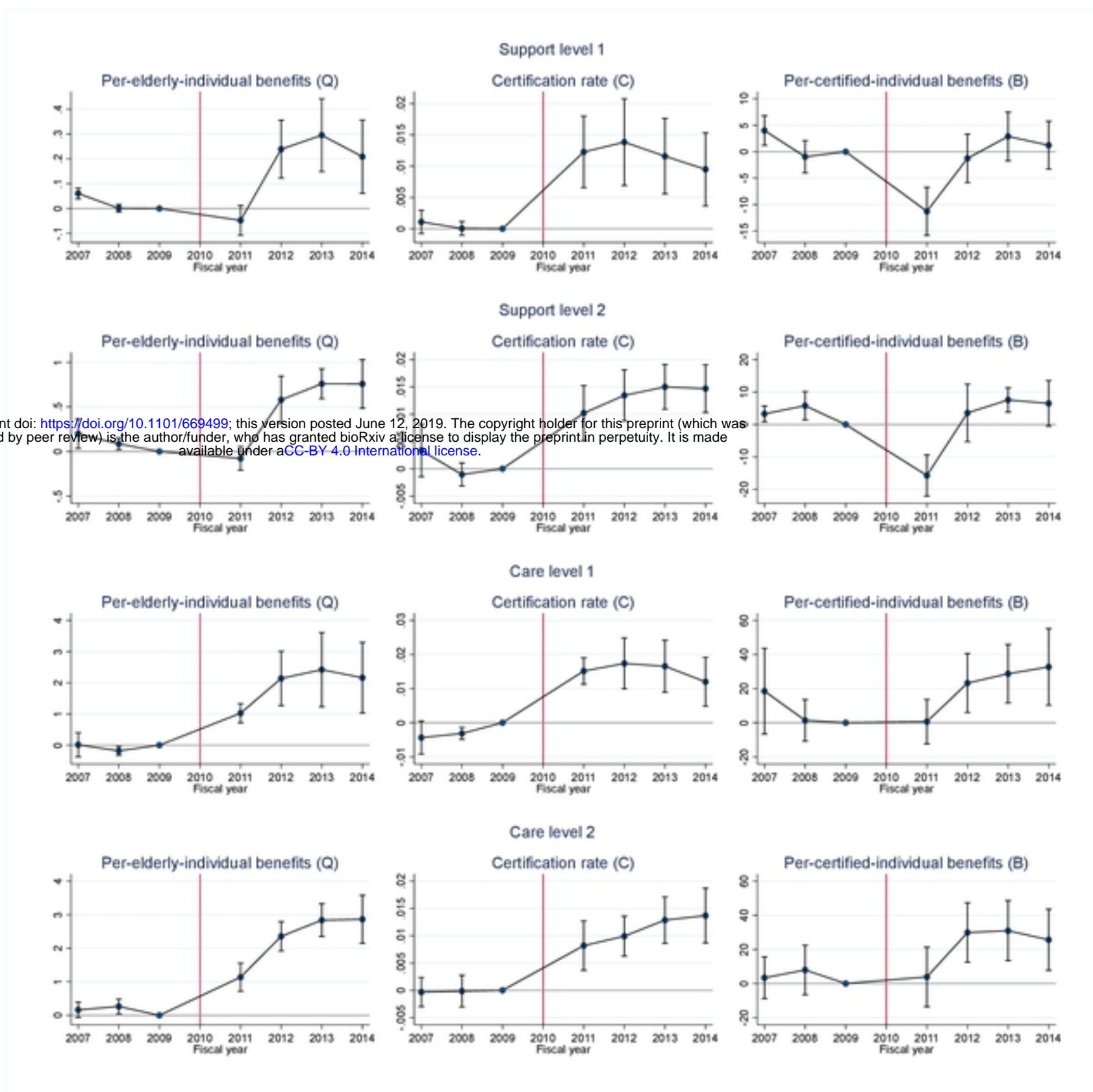
Figure 4: Trends in the numbers of certified individuals and aged 65+ in the evacuation and control areas. Certified individuals: the number of people aged 65 and older who were certified to receive long-term care services. Aged 65+: the number of people aged 65 and older. The numbers are standardized to 100 in 2009.



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Figure 5: Difference-in-differences (DID) estimates for different weighting schemes and samples and their 95% confidence intervals. Confidence intervals are calculated by clustered robust standard errors clustered by municipality. All DID estimates are estimates for β_τ in equation (1). The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas). “OLS” is a normal ordinary least squares regression and “WLS” is a weighted least squares regression in which the weight is the number of elderly individuals for the outcomes of Q and C and the number of certified individuals for the outcome of B . The trimmed sample includes the municipalities in the baseline sample whose outcome values are within the minimum and maximum of outcome values of the treated municipalities in all three pre-disaster years. The Fukushima sample contains the evacuation and non-evacuation municipalities in Fukushima prefecture.

APPENDICES



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Figure A.1: Difference-in-differences (DID) estimates and their 95% confidence intervals (lower care levels). Confidence intervals are calculated by clustered robust standard errors clustered by municipality. All DID estimates are estimates for β_τ in equation (1). The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas).

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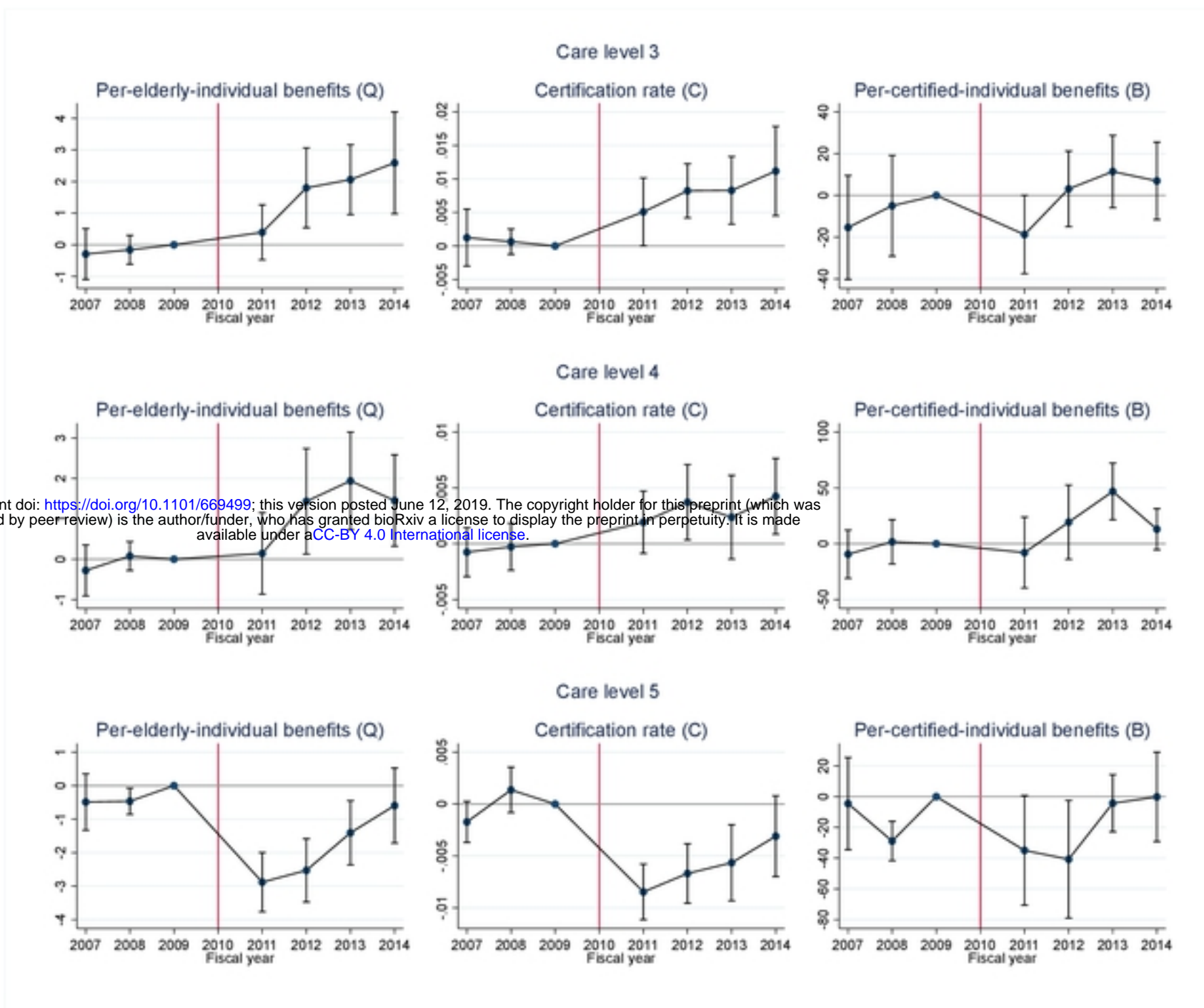
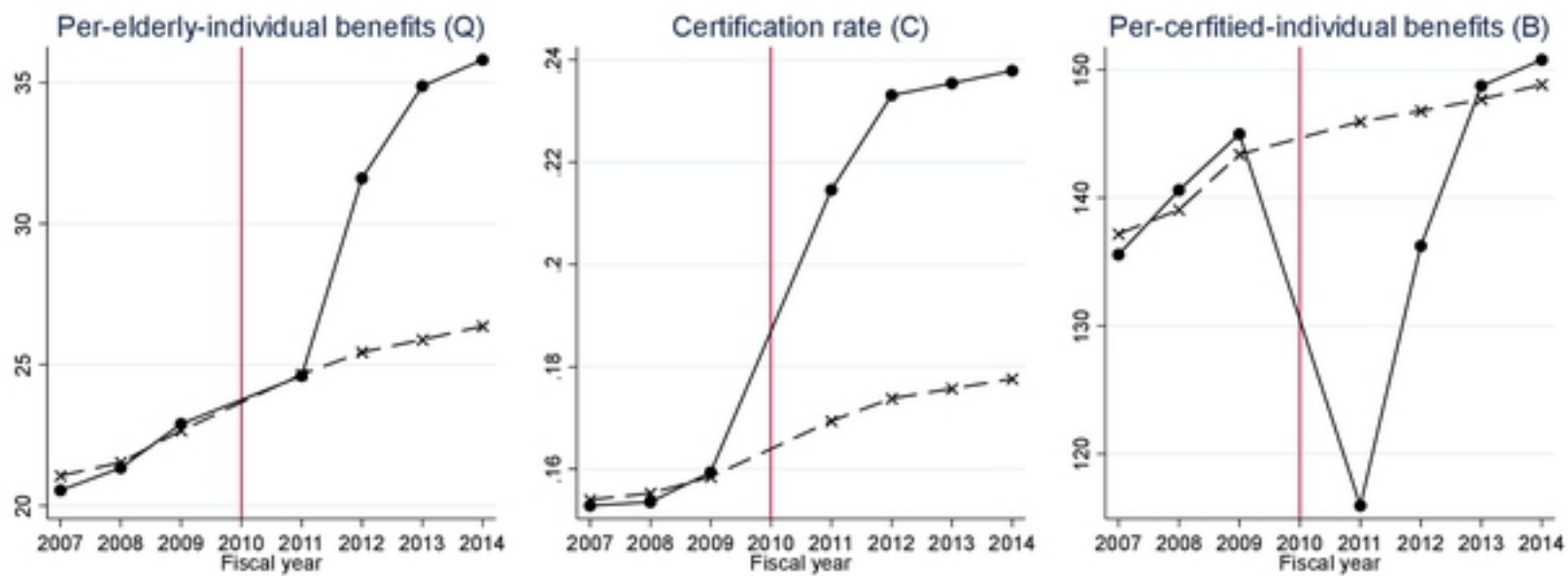


Figure A.2: Difference-in-differences (DID) estimates and their 95% confidence intervals (higher care levels). Confidence intervals are calculated by clustered robust standard errors clustered by municipality. All DID estimates are estimates for β_τ in equation (1). The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas).



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Figure A.3: Time trends of long-term care benefits in the Fukushima mandatory evacuation group and the non-evacuation control group with the trimmed sample. Per-elderly-individual benefits: long-term-care benefits per elderly person. Certification rate: the percentage of people aged 65 and older who were certified to receive long-term care services. Per-certified-individual benefits: long-term care benefits per certified individual. The unit of “benefit” is 1,000 points (10,000 JPY or around 100 USD in standard areas). The trimmed sample includes the municipalities in the baseline sample whose outcome values are within the minimum and maximum of outcome values of the treated municipalities in all three pre-disaster years.