

In this case, the economic condition of the elderly people is a major problem for both the state and the elderly people themselves. (Meng et al. 2020). According to Hange et al. (2016) of actual economic condition, 41% of the elderly people have a negative correlation between age and economic condition, while the other 59% of the elderly people have a non-negative correlation. A number of factors can cause a negative correlation between age and economic condition, e.g., lack of mobility, existing health conditions, limited social network, low income, and poor nutrition and health (Cohen and Milane 2005).

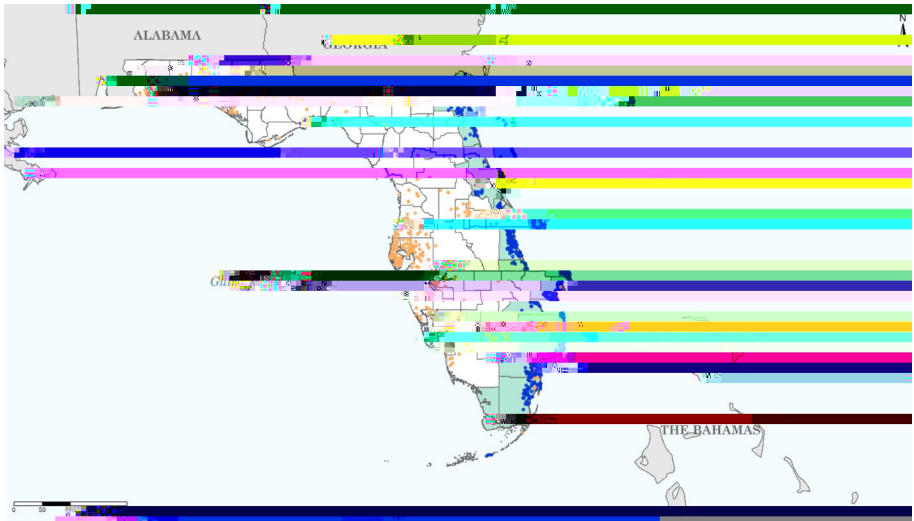


Fig. 1 Location of COVID-19 cases in Florida on February 2020 (in blue dot), June 2020 (in green color) and November 2020 (in orange dot)

As a result of the COVID-19 pandemic, the Florida economy is likely to be affected in the long run. The Florida economy is likely to be affected in the long run. The Florida economy is likely to be affected in the long run.

The average age of COVID-19 cases in Florida in June 2020 was 48 years old. The average age of COVID-19 cases in Florida in June 2020 was 48 years old. The average age of COVID-19 cases in Florida in June 2020 was 48 years old.

Moreover, the collected data on the economic impact of the COVID-19 pandemic in Florida in November 2020. The economic impact of the COVID-19 pandemic in Florida in November 2020. The economic impact of the COVID-19 pandemic in Florida in November 2020.

Table 1 Coding of variables used in regression models

Variable	Coding
Worried about flooding	<i>I am worried about the danger of a flood at my current residence</i> 1 = Strongly disagree, 5 = Strongly agree
Perceived flood probability ^a	<i>What is your best estimate of how often a flood will occur at your home?</i> categorical, 1 = less than 1/1,000 years, 7 = more often than 1/10 years
Age	<i>How old are you?</i> in years
Education	<i>What is your highest completed level of education?</i> 1 = Some high school, 5 = college
Income	<i>Which of the following describes your total household income for 2019 before taxes?</i> 1 = less than \$10,000, 6 = \$125,000 or more
Length of residence	<i>How long have you lived in your current residence (in years)?</i>

^a Ordinal scale based on hierarchical coding of household size: 1 = between 1 and 1,000 years, 2 = 1 in 100 years, 3 = between 1 in 1000 years, 4 = between 1 in 100 years, 5 = between 1 in 100 years, 6 = between 1 in 100 years, 7 = between 1 in 100 years

being impacted by wind and flooding caused by EA. The average age of respondents in November 2020 was 47 years, with an average household income of \$40,134 per year before taxes and 69% are female. Respondents reported their highest level of education as follows: some high school (3%), high school graduate (26%), some college (29%), college graduate (30%), and postgraduate (12%). Compared to the June 2020 survey, respondents in November 2020 have a lower income and education level, but are similar in terms of age and gender.

Table 1 describes the variables used in our statistical analysis. Each variable was coded the same in the household survey. To elicit these variables, respondents faced several questions in relation to their risk perception, association with flooding and COVID-19, own experience with flooding, income, length of residence and gender. The overall selection of these variables allows us to identify how the experience of COVID-19 has affected the association, controlling

For example, more people are worried about COVID-19 (63%) than about food (33%), as Fig. 2 illustrates. Moreover, only 21% disagree and 7% strongly disagree in the statement, “The probability of flooding is low. I am not concerned about the consequences of a flood.” These percentages are 28% and 30%, respectively.

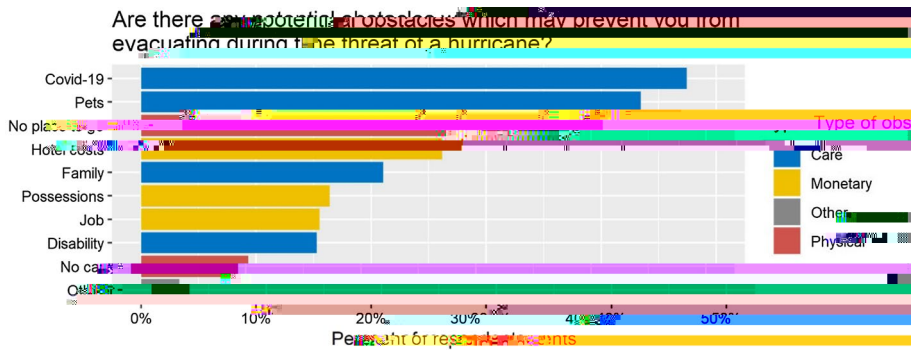


Fig. 3 Percentage of respondents who identified obstacles to evacuation, by obstacle type (based on the June 2020 survey). Same legend as in the online version of the journal

3.2 Evacuation intentions at the start of the 2020 hurricane season

When being asked about intentions to evacuate to a safe place and a plan to evacuate in the case of the hurricane season, 39% of respondents have indicated they are most likely to evacuate (see Table 2).

Despite the fact that COVID-19 is the most common obstacle to evacuation during the 2020 hurricane season, both of our surveys in February and June 2020 contained a question about the obstacles to evacuation during a hurricane season. More respondents indicated a lack of financial resources in the June survey than in the February survey (75% vs. 56%). The inability to afford the cost of evacuation was the most frequently mentioned obstacle during Hurricane Dorian (26% of the February 2020 survey respondents had an obstacle). However, as Fig. 3 illustrates, the cost of evacuation was the most frequently mentioned obstacle during the 2020 hurricane season, although the percentage of respondents who listed the cost of evacuation as an obstacle remained stable at 26%. Instead, COVID-19 was the most frequently mentioned obstacle, but almost half of the respondents in the June survey reported the inability to evacuate. Although the percentage of respondents who reported an inability to evacuate as an obstacle during the pandemic.

We conducted a logistic regression analysis to examine how evacuation intentions are related to a set of socio-demographic characteristics and the interaction of the hurricane and COVID-19 risk. An ordered probit model of the evacuation intentions is also presented in the online version of the journal. The socio-demographic characteristics that are most likely to be associated with evacuation intentions are (Table 3).⁷ As a result, the interaction of the hurricane and COVID-19 risk and length of residence are the most important variables in the model to examine the relationship between age and evacuation intentions. We will hold these variables constant for the remainder of the analysis. We include the variables length of residence, education, mobility, and housing, education, and infection mobility and concern about COVID-19 in model 2, since these variables are all significant predictors of evacuation intentions (Pearson correlation coefficient p

⁷ The main results in Table 3 and Tables 4, 5 and 6 show below are only those that include age and marital status. The variables age 65 and higher, which include age groups that are most likely to be infected and die from infection by COVID-19.

Table 3 Odds ratio model of variables of influence on ol n a e ac a ion in en ion (based on the June 2020 case)

*** $p < 0.01$; ** $p < 0.05$; * $p < 0.1$

Standard error and confidence intervals are shown below the coefficient. An odds ratio model is used to account for the ordinal nature of the dependent variable (1

at $p < 0.05$), and a 95% confidence interval of ol n a e ac a ion in en ion. The effect of these variables may vary. Some of the relationships between age and ol n a e ac a ion in en ion. Variables have been selected in age a e omi ed from the regression in model 2. In a mediation analysis (Table 4), we included significant variables when calculating indirect effects which is the effect of the relationships between age and ol n a e ac a ion has can be attributed to mediation variables. The other 95% confidence intervals are controlled for in a model has included variables.

We find that the likelihood of ol n a e ac a ion significantly increased in household, but significantly declined in household concerning the confidence of becoming infected by COVID-19 and the length of evidence (Table 3).⁸ Moreover, the

Table 4 Decomposition of the total effect of age on ol n a e ac a ion in o di ec and indi ec e ec ia conce n abo COVID-19, o abo ooding and leng h of eidence. Using the o de ed obi model (based on the June 2020 C e)

	With con ol a iable	Incl ding con ol a iable
Total e ec	0.012*** (0.003)	0.010*** (0.004)
Di ec e ec	0.005 (0.003)	0.003 (0.004)
Indi ec e ec	0.007*** (0.001)	0.007*** (0.002)
ia conce n abo COVID-19	0.001** (0.001)	0.003** (0.001)
ia o abo ooding	0.004*** (0.001)	0.003*** (0.001)
ia leng h of eidence	0.001* (0.001)	0.001 (0.001)
Media ion e cen age	59.57	72.51
ia conce n abo COVID-19	12.73	27.96
ia o abo ooding	35.88	33.35
ia leng h of eidence	10.95	11.20
Ob e a ion	527	362

*** $p < 0.01$; ** $p < 0.05$; * $p < 0.1$

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Table 4 diC la C he o al e ec of age on ol n a e ac a ion, di ided in o a di ec and indi ec e ec ia conce n abo he conce . enceCof becoming infec ed b COVID-19, o abo ooding, and leng h of eidence. O e all, he o al e ec ho C ha olde indi id alCha e lo e e ac a ion in en ionC Con olling fo conce n abo he conce . enceCof becoming infec ed b COVID-19, o abo ooding, and leng h of eidence lea eCan in igni can di ec e ec of age. The indi ec e ec, hich iC he sha e of he ela ionChi be en age and ol n a e ac a ion ha can be a ib ed o e ce - ionCof COVID-19 and ood i kCand leng h of eidence, iCe lained b he co c ien eCima e 0.007 (p al e < 0.01). Be en 60 and 73% (de ending on incl ded con ol a iableC) of he ela ionChi be en age and ol n a e ac a ion iCe lained b conce n abo he conce . enceCof becoming infec ed b COVID-19, o abo ooding, and leng h of eidence.¹⁰ The o i k e ce ion a iableCa e Ca iC call igni can and e lain a la ge o o ion of he ela ionChi han he leng h of eidence.

Footnote 8 (continued)

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⁹ We al e nd ha olde indi id alCa e leClikel o Ca e ha he in end o e ac a e if he e iCa manda o e ac a ion o de, b hiCiCno de o e ce ionC ela ed o COVID-19 and ooding. Ho e e, hiC ma be de o o he iCeClike oo mobili and heal h, o limi ed social ne o kC hich e do no ca - e in he C e .

¹⁰ Co ela ion anal eCCho ha a highe age iCaCcia ed i h highe conce n abo COVID-19 and leng h of eidence, b a lo e o abo ooding.

Table 5 Ordered probit model of evacuation intentions during Hurricane Eta (based on the November 2020 survey)

	Coefficient Model 1	Coefficient Model 2
<i>Socio-demographics</i>		
Age	0.017*** (0.00)	0.006* (0.00)
Gender (1 = female)	0.003 (0.10)	0.038 (0.12)
Education	0.018 (0.05)	0.075 (0.06)
Income	0.053 (0.04)	0.021 (0.04)
Length of residence		0.001 (0.01)
<i>Flood risk perceptions</i>		
Perceived flood damage		0.071* (0.04)
Work availability		0.324*** (0.05)
<i>COVID-19 related flood probability</i>		

*** $p < 0.01$; ** $p < 0.05$; * $p < 0.1$.

Standard errors shown in brackets below the coefficients. An ordered probit model is used to account for the ordinal nature of the dependent variable (1 = not at all likely to evacuate, 4 = extremely likely to evacuate).

3.3 Evacuation intentions during Hurricane Eta

When respondents were asked during the height of Hurricane Eta how likely they were going to evacuate to a safe place, 35% answered highly unlikely, 27% answered unlikely, 10% answered likely, and only 6% answered extremely likely. We evaluated the same analysis of evacuation intentions during the 2020 hurricane season (have been reported in Tables 3 and 4) for evacuation intentions during Hurricane Eta, which hit Florida at the end of the hurricane season in November 2020. These results are shown in Tables 5 and 6. The ordered probit model coefficients in Table 5 compare evacuation intentions during Hurricane Eta to negative evacuation intentions (model 1), of which the significant decline is marginally significant in model 2 when risk perception is added. These findings again show that evacuation intentions during Hurricane Eta are negatively affected by concerns of becoming infected by COVID-19, and of increased flood risk perception. However, the sign of the


coefficient of correlation of the length of residence is the same in Table 5 as Table 3, his coefficient of correlation in Table 5. When each of the selected countries is infected, the probability of infection can be determined by the relationship between the variables in the equation, which may be determined by the

andemic. Indeed, the early onset of COVID-19 in Florida coincided with the end of the 2020 hurricane season, which has historically been dominated by hurricanes. The onset of COVID-19, which is an obstacle for the state. Moreover, older people, who are more concerned about the consequences of becoming infected by COVID-19, are likely to be affected among a sample of the 2020 hurricane season, and continued by real-time COVID-19 cases at the end of the hurricane season during the height of Hurricane Eta. This should be taken into account by officials aimed at minimizing hurricane-related deaths and a disease for which older people are more vulnerable. The majority of the reported deaths and actions have not coincided with a pandemic did not observe a significant increase of age (Baker 1991; Solomon 2000).

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