

1 **Discount Rate Selection for Investments in Climate Change Adaptation and Flood**
2 **Risk Reduction Projects**

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10 **Abstract**

11 The selection of an appropriate discount rate is a critical yet often neglected aspect of
12 valuing flood protection and climate adaptation projects. We examine current discounting
13 practices in flood risk management and climate adaptation literature and compare these practices
14 with a proposed fair market value (FMV) discounting approach, in which we discount project cash
15 flows based on their levels of systematic market risk, consistent with practice in the private sector.
16 Using publicly available data from the National Flood Insurance Program (NFIP), we demonstrate
17 minimal correlation between flood risk-related cash flows and overall market returns, suggesting
18 flooding-related cash flows are most appropriately discounted near the risk-free rate. We consider
19 the example of a proposed storm surge barrier in Boston Harbor and compare prevailing public
20 sector discounting approaches to this FMV approach. Our results suggest prevailing discounting
21 approaches systematically undervalue flood risk reduction benefits and climate adaptation
22 investments relative to the private sector valuation implied via the FMV approach.

23 **Introduction**

24 Valuation is a critical component of the decision-making process for any project or
25 investment. The selection of a discount rate is an important step in valuation, as it can greatly
26 influence the perceived profitability of an investment or project. This is particularly true when
27 valuing infrastructure investments, as they often entail large upfront costs and future benefits that
28 accrue over the lifespan of the asset (Chen & Bartle, 2022). In such situations, the present value
29 of future benefits is highly sensitive to changes in the discount rate. As the discount rate increases,
30 larger future benefits are required to offset initial project costs and justify investment in a given
31 project. Climate change adaptation projects represent a rapidly expanding class of infrastructure
32 and are especially sensitive to discount rate selection, as the expected benefits tend to increase with
33 time, proportional to increases in the frequency and/or intensity of climate-related hazards (i.e.,
34 coastal flooding associated with storm surge and sea level rise; Strauss et al., 2021; IPCC, 2022).
35 Despite this notable sensitivity and the prevalence of a varied range of discounting practices,
36 discount rate selection is often neglected or misunderstood within the climate adaptation and
37 coastal flood risk management literature. The existing literature also lacks substantial guidance on
38 discount rate selection relevant to the private sector, despite the growing need for investments in
39 adaptation across industries within the private sector and rapidly increasing property and casualty
40 insurance premiums (ECAWG, 2009; Kousky et al., 2021).

41 In order to address this gap in the literature, we first examine relevant discounting
42 rationales employed in the literature for valuation of public sector adaptation investments,
43 inclusive of relevant U.S. federal policy guidance. Next, relying on basic modern financial
44 principles, we outline a private sector oriented fair market value (FMV) approach to discount rate

45 selection for adaptation investments, wherein discount rates for project cash flows are informed
46 by their underlying systematic risk and correlation to the broader market. This stands in contrast
47 to prevailing public sector discounting rationales applied to adaptation investments, which
48 routinely neglect such systematic market risks. Operationalizing the proposed FMV approach, we
49 first estimate the systematic (i.e., market correlated risk) for flood risk-related cash flows,
50 examining the correlation between U.S. flood insurance claims and broader market returns via a
51 copula-based approach. Informed by these results, we apply the FMV approach to value a real-
52 world climate adaptation case study, a proposed Boston Harbor Barrier, and underscore the general
53 implications of the proposed FMV approach for climate adaptation projects. We conclude with
54 actionable policy and best practice recommendations for appropriate discounting practices specific
55 to coastal flood protection and climate adaptation projects.

56 **Prevailing Discounting Practices Presented in Climate Adaptation Literature**

57 In adaptation planning literature and practice, discount rate selection is often treated in a
58 summary fashion, using a rate provided or extracted from policy guidance with little justification
59 (e.g., Kirshen et al., 2020; Melo et al., 2020, DHS, 2020; Ha et al., 2021; Stewart & Rosowsky,
60 2022). The rather cavalier and straightforward discount rate selection process can be considered
61 pragmatic, as alignment of economic analyses with public policy guidance lends a nominal degree
62 of legitimacy to the findings and conclusions associated with a given study, particularly if
63 conclusions are intended to support (in)action within prevailing institutional frameworks or federal
64 funding programs (e.g., justification of funding under the Federal Emergency Management;
65 FEMA Building Resilient Infrastructure and Communities; BRIC program; DHS, 2020).
66 Furthermore, since public policy within most western countries specifies the use of a single

67 discount rate for public sector investments (Bastidas-Arteaga & Stewart, 2019; Gollier, 2021), the
68 selection of any alternative rate implies a direct challenge to the assumptions underlying prevailing
69 public policy, though such contradictions are not necessarily irreconcilable in practice. Relevant
70 case law in the United States suggests federal agencies in fact have some latitude in the selection
71 of discount rate, insofar as they provide substantive justification for a selected discount rate
72 (Viscusi, 2022).

73 Rather than select and justify a single discount rate, some authors instead apply a more
74 agnostic approach and present results based on a range of rates, choosing to conduct sensitivity
75 analyses that illustrate the impact of discount rate on the present value of an adaptation investment
76 or study recommendations (e.g., Lincke & Hinkel, 2018; Oddo et al., 2020). Nominally aligning
77 with OMB (1992) best practice recommendations, such analyses can be useful, particularly when
78 the resulting investment decision is insensitive to changes in the discount rate (i.e., if the results
79 are robust to changes in discount rate). Rather than consider a range of discount rates, other authors
80 sidestep the selection of discount rate and investment valuation entirely, by presenting
81 undiscounted future values (e.g., Hallegatte et al., 2013; Neumann et al., 2021) intending for these
82 to serve as the basis of policy and investment recommendations. Still others formulate adaptation
83 planning frameworks that avoid financial analysis and discount rates entirely (e.g., Buchanan, et
84 al., 2016; Rasmussen et al., 2020). While studies that avoid the selection of a discount rate can
85 provide useful information for decision making, they ultimately fall short of valuing adaptation
86 investments, thereby disallowing direct economic comparisons of adaptation projects to other
87 alternative public investments, all of which are ultimately competing for a limited availability of
88 funds.

89 Another approach is to apply a discount rate equivalent to the cost of borrowing for
90 government institutions, in effect recommending use of the risk-free rate for valuation of projects.
91 For instance, NIBS (2019) recommends discounting adaptation investments using the real yield of
92 a 3-month U.S. Treasury bond. The current US Army Corps of Engineers (USACE) discount rate
93 selection policy takes a similar approach (notwithstanding the archaic peculiarities specific to their
94 discount rate selection formula) specifying that the discount rate for water resources projects shall
95 be informed by the nominal yield of long-term (i.e., >10 years to maturity) U.S. Treasury bonds
96 (Water Resources Development Act, 1974; USACE, 2019; Discount Rate Rule WRC, 2020).
97 Other authors (most notably Arrow & Lind, 1970) provide justification for discounting public
98 sector investments at the risk-free rate, insofar as it is suggested that the (federal) government is
99 capable of diversifying away the contributions towards aggregate (i.e., systematic or market) risk
100 associated with a given investment (Arrow & Lind, 1970; Gollier, 2021).

101 A separate contingent of authors suggest that adaptation projects represent
102 intergenerational wealth transfers over exceptionally long-lived horizons (similar to climate
103 change mitigation efforts) and hence, advocate for low or negative discount rates (Markanday et
104 al., 2019; Keenan, 2019) despite the presence of substantial near-term benefits associated with the
105 majority of such projects, and the lack of consensus on the expectations of relative wealth
106 inequality between generations or the subsequent implications for intergenerational discounting
107 (Viscusi, 2022). A larger subset of authors consider a social time preference (discount) rate
108 (STPR), applying Ramsey-style discount rate (Ramsey, 1928) that is informed by a time preference
109 (typically inclusive of an allowance for catastrophic systemic risk) and a wealth effect informed
110 by expectations of future consumption growth (HM Treasury, 2020; Vousdoukas et al., 2020; Lee
111 & Ellingwood, 2015; Rennert et al., 2022). A tangential discounting rationale, rooted in

112 implementation of the consumption capital asset pricing model (CCAPM) suggests the discount
113 rate for public sector investments follows a declining term structure (Lowe, 2008; Gollier, 2014;
114 Lee & Ellingwood, 2015), particularly in the valuation of projects with lifespans surpassing a
115 certain threshold (e.g., 30-50 years; Bastidas-Arteaga & Stewart, 2019). As noted by Gollier &
116 Hammitt (2014), this declining term structure is valid for investments whose associated risk
117 premium under the CCAPM is less than a certain value (the relative inequality aversion constant),
118 whereas if its risk premium is greater than this value, then discount rates would instead increase
119 with time, reflecting greater uncertainty in systematic risk.

120 **Methodology: A fair-value approach to discount rate selection**

121 Rather than rely on any given policy or public sector-oriented discounting philosophy at
122 face value, we instead aim to estimate the systematic (i.e., market-correlated) risk and resultant
123 discount rates for flood risk related cash flows and adaptation investment costs, given available
124 data. Taking a market driven approach, we assume the discount rate for any given (public or private
125 sector) investment should reflect its inherent level of systematic risk, as such risk is in practice
126 non-diversifiable, irrespective of the investor (Lucas, 2014a; Gollier, 2021). While this contrasts
127 with the majority of previously outlined discounting practices, such a fair market value (FMV)
128 approach to discount rate selection is consistent with basic principles of corporate finance and best
129 practices in the private sector. Fundamentally, such a valuation approach aims to determine the
130 value of a (public or private) investment on the open market, rather than its taxpayer subsidized
131 value (an issue which is particularly relevant during the negotiations between a public agency and
132 a private consortium prior to the establishment of a Public-Private Partnership, PPP; Lucas &
133 Montesinos, 2020). Such discounting approaches minimize the possibility of “budgetary arbitrage”

134 wherein government investments in risky assets (e.g., high-risk mortgages) only appear profitable
135 if discounted at the cost of government borrowing, and inherent market risk is neglected (Lucas,
136 2014b). Further, discounting at the cost of borrowing would violate the Modigliani-Miller theorem,
137 a basic principle of modern finance, in effect subsidizing riskier investments at the taxpayers' (i.e.,
138 equity-holders') expense (Lucas, 2014b).

139 As such, from a FMV perspective, market prices are the best available proxy for social
140 value, and the discount rate, or more precisely, the cost of capital (COC), should therefore reflect
141 the expected return an investor would make from a typical asset with similar level of systematic
142 risk and return as the asset (or cash flow) of interest (Geltner & de Neufville, 2018). Rephrased,
143 the discount rate is the minimum acceptable rate of return an investor or management is willing to
144 accept for a given project or cash flow (Martland, 2012). Such discount rates incorporate the effects
145 of both time preference and risk aversion on value.

146 Here, we determine the appropriate discount rate for adaptation project costs and benefits
147 (i.e., flood-related cash flows) via the Capital Asset Pricing Model (CAPM). We choose to employ
148 the CAPM rather than the consumption-based capital asset pricing model (CCAPM) as in practice,
149 the CCAPM has not well-characterized observed market risk premia (Mankiw & Shapiro, 1984;
150 Gollier & Hammitt, 2014). The CAPM is widely recognized and often employed by institutional
151 investors when assessing expected returns of assets when constructing investment portfolios
152 (Brealey et al., 2011). The CAPM defines the expected return of an asset or cash flow ($E[R_a]$)
153 based on the present risk-free rate (R_f , i.e., the rate of return on a risk-less asset) the expected
154 return of the broader market, $E[R_m]$, and the level (or correlation) of risk relative to the market,
155 the asset beta, β_a , (Brealey et al., 2011):

156
$$E[R_a] = R_f + \beta_a(E[R_m] - R_f) \tag{1}$$

157 Estimates of the level of systematic risk relative to the market (i.e., the risk premium and
158 related asset beta) are typically found by compiling historical data on the unlevered returns of the
159 asset of interest or economic sector relative to the market (e.g., the correlation of stock performance
160 to the overall stock market) or through correlations of comparable assets or industries to the
161 market, when information on the specific asset or cash flow of interest is unavailable or incomplete
162 (e.g., see Damodaran, 2021). Mathematically, the asset beta is defined as the covariance of asset
163 returns (R_a) and market returns (R_m), divided by the variance of the market returns:

164
$$\beta_a = \frac{cov(R_a, R_m)}{var(R_m)} \tag{2}$$

165 *Market Rate of Return and Risk-Free Rate*

166 Selection of an appropriate expected market rate of return and risk-free rate is an often-
167 overlooked aspect of operationalizing the CAPM yet is central to the determination of expected
168 returns and the appropriate opportunity cost of capital. A recent survey of financial analysts,
169 managers, and professors across 88 countries suggests that the market rate or return, market risk
170 premium, and risk-free rate are highly dependent on geographical market, and can vary widely
171 with a given market (Fernandez et al., 2021). In the U.S., the survey found an average expected
172 market return of 7.3%, reflecting investor sentiment on the expected future returns of the broader
173 U.S. stock market, rather than historical, required, or implied market returns (Fernandez et al.,
174 2021). (For reference, relying on data collected by Shiller (2022), we estimate the S&P500
175 returned 7.34% over the 50 years prior to October 2022.) Here, we adjust this value for inflation,
176 (which based on the CPI has averaged 2.5% over the past 10 years as of the time of writing; Shiller,
177 2022) and consider the real expected market return to be 4.8%.

178 Fernandez et al. (2021) find the average risk-free rate across participants in the US was 1.8%, (i.e.,
179 less than the US dollar long-term inflation rate) and was largely informed by nominal short- and
180 medium-term US treasury yields. Alternatively, the yield of a duration-matched US treasury bond
181 can also be employed as an appropriate risk-free rate, consistent with USACE policy (USACE,
182 2021). Here, we apply a risk-free rate equal to the 5-year average of the real yield of long-term
183 (i.e., >10 years to maturity) U.S. Treasury bonds (U.S. Treasury, 2022), calculated to be 0.43% as
184 of October, 2022.

185 *Risk Premium of Adaptation Costs*

186 Recognizing the different levels of systematic risk associated with climate adaptation
187 investment costs and benefits, we apply separate discount rates to project costs and benefits. In
188 contrast to flood-risk related cash flows (which we explore further in the following subsection),
189 the costs associated with construction of flood protection measures and related infrastructure (e.g.,
190 costs of elevating critical infrastructure assets) are correlated with the market (i.e., subject to
191 systematic market risk) and therefore carry an associated risk premium under the CAPM pricing
192 framework. As such, the appropriate discount rate for adaptation costs varies by industry (Lucas
193 & Montesinos, 2020). For example, an adaptation project undertaken by a transit agency could
194 reasonably discount project costs (and other benefits associated with regular operations and
195 maintenance) using the expected rate of return for the transportation sector as determined by
196 equation (1) and the appropriate asset beta. **Table 1** provides unlevered asset betas (Damodaran,
197 2021) and associated discount rates (using the inflation-adjusted expected market return and real
198 risk-free rate established in the prior subsection) for various U.S. industries likely to undertake
199 climate adaptation investments.

200 **Table 1.** Asset betas and CAPM derived (real) discount rates associated with industries likely to
 201 undertake climate adaptation or flood risk reduction investments

Industry Name	β_{α}^*	r_{real}
Transportation	0.91	4.4%
Air Transport	0.78	3.8%
Utility (General)	0.28	1.6%
Utility (Energy)	0.38	2.1%
Utility (Water)	0.42	2.3%
Telecom. Services	0.63	3.2%
Green & Renewable Energy	0.68	3.4%
Environmental & Waste Services	0.83	4.1%
Insurance (Prop/Cas.)	0.67	3.3%
Reinsurance	0.80	3.9%
Real Estate (Development)	0.72	3.6%
Real Estate (Operations & Services)	0.80	3.9%
Engineering/Construction	1.05	5.0%
Hospitals/Healthcare	0.56	2.9%

*average unlevered asset beta corrected for cash
 (2016-2021); Damodaran (2021)

202

203

204 *Risk Premium of Flood Risk-Related Cash Flows*

205 Despite a lack of empirical evidence in the literature, it is generally accepted that localized
 206 catastrophes (e.g., floods) have a negligible influence on the overall economy (Merton, 1976).
 207 Related literature focused on catastrophe bond pricing rests on the resulting premise that insured
 208 catastrophe events (and resultant cash flows) are uncorrelated with the market (Lee & Yu, 2002;
 209 Froot & Posner, 2003; Ma & Ma, 2013; Nowak & Romaniuk, 2013). Attempting to provide an
 210 empirical basis for this assumption, we investigate the level of systematic market risk inherent in

211 flood risk-related cash flows, as characterized by the FEMA National Flood Insurance Program
212 (NFIP) redacted claims dataset (FEMA, 2021), which provides detailed record of insured flood
213 damage policy payouts within the United States since the inception of the program. Although NFIP
214 claims do not represent the sum total of flood-related damage costs in the U.S. (given NFIP policies
215 only offer limited coverage and only represent a subset of insured losses; Smith & Katz, 2013),
216 they are nonetheless a robust flood risk-related cash flow with sufficient data available to perform
217 a reliable longitudinal analysis.

218 The NFIP redacted claims dataset contains over 2,500,000 flood insurance policy claims
219 from the inception of the program in 1970 and is updated regularly (Dombrowski et al., 2020).
220 The dataset is directly available from FEMA, either as raw data or via the OpenFEMA API
221 (FEMA, 2021). Using the OpenFEMA API and a data scraping algorithm (written in JavaScript),
222 we aggregate monthly NFIP claims based on the date of claim (for both structures and contents
223 claims) paid to policyholders. Adjusting monthly claims data for inflation using the U.S. Bureau
224 of Labor and Statistics (BLS) consumer price index (CPI; Shiller, 2022) we determine the real
225 value of monthly claims from 1970 to July 2021 (expressed in USD at 2021 Q2 price levels). Here,
226 we choose to adjust for inflation using the CPI, as this index considers price changes in both
227 contents (i.e., consumer goods) and structure (i.e., shelter) from the perspective of a policyholder
228 (i.e., consumer). Further, this data is readily and publicly available over the time period of interest,
229 thereby increasing the transparency and replicability of the proposed methodology. Alternatively,
230 a construction cost index (e.g., the paywalled Engineering News Record; ENR; construction cost
231 index) could be applied instead, though such an index is best applied to structures damages only.

232 Here, we investigate the relationship between the monthly change in cumulative NFIP
 233 claims and S&P 500 returns (as a proxy for overall market returns) for the period of January 1980
 234 to July 2021, modelling the relationship between these two assets via a Frank copula. Copulas are
 235 multivariate joint probability distributions that relate the probability of two marginal distributions
 236 given a (typically univariate) measure of dependence between the marginals (Joe, 2014). While
 237 there are a variety of potential bivariate Archimedean copulas that could be used to relate the
 238 returns of these two assets, we employ a Frank copula as it allows for characterization of negative
 239 dependence between marginal distributions (Joe, 2014). Creating the copula first requires the
 240 specification of a marginal (probability) distribution for each variable (i.e., asset) of interest. Here,
 241 we fit a general form beta distribution to the sample data for each asset (Guthrie, 2020). Next, we
 242 characterize the correlation between these two assets using the historical sample data, first via the
 243 asset beta (as defined above), and via the Kendall rank correlation (i.e., the relative concordance
 244 of data pair rankings between the sets). Next, we use this Kendall rank correlation, τ_{sample} , to
 245 directly estimate the dependence parameter, θ , for the Frank copula via the following closed form
 246 solution (Joe, 2014; Weber, 2015):

$$247 \quad \tau_{sample} = 1 + \frac{4}{\theta}(D_1(\theta) - 1) \quad (3)$$

248 where $D_1(\theta)$ denotes the 1st order Debye function.

249 Given this dependence variable, the marginal cumulative distribution functions
 250 characterizing monthly changes in cumulative NFIP claims, $F_1(u)$, and S&P 500 returns, $F_2(v)$,
 251 we can characterize the observed dependence between the two marginal distributions via a Frank
 252 copula (Joe, 2014):

253
$$C(u, v, \theta) = \frac{-1}{\theta} \ln \left(\frac{(1-e^{-\theta} - (1-e^{-\theta}u)(1-e^{-\theta}v))}{1-e^{-\theta}} \right) \quad (4)$$

254
$$c(u, v, \theta) = \frac{\theta(1-e^{-\theta})e^{-\theta(u+v)}}{[1-e^{-\theta} - (1-e^{-\theta}u)(1-e^{-\theta}v)]^2} \quad (5)$$

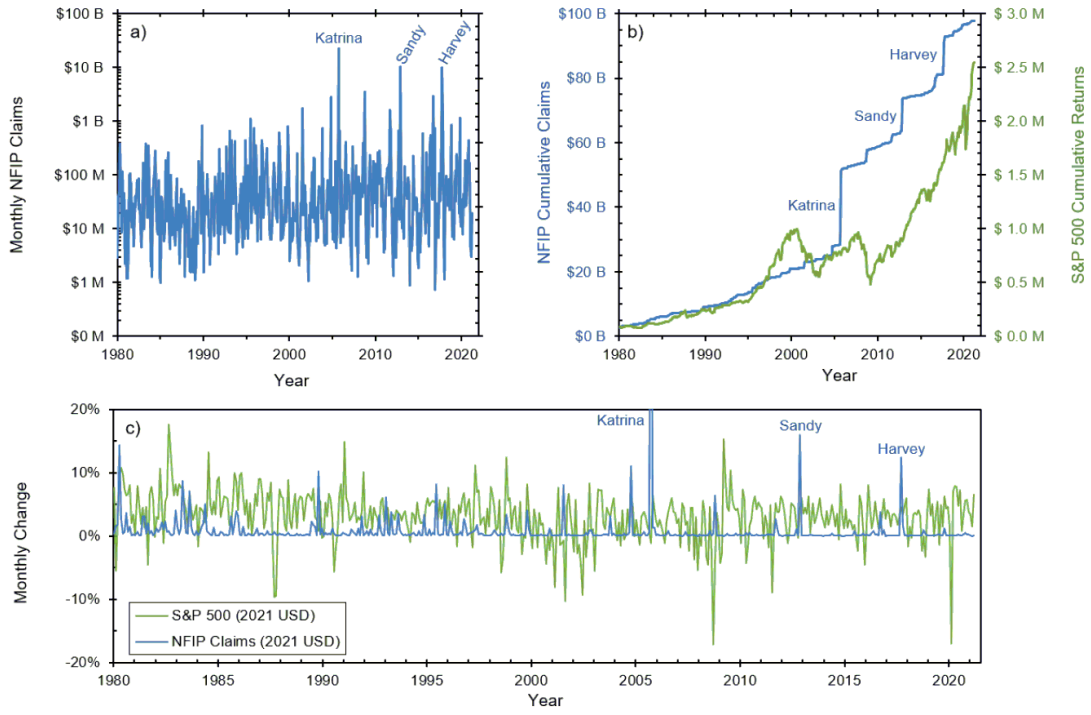
255 where $C(u, v, \theta)$ denotes the joint cumulative distribution function (CDF) and $c(u, v, \theta)$ denotes
 256 the joint probability distribution function (PDF).

257 Given this copula, we develop a full range of possible market risk premia (i.e., CAPM beta
 258 values using equation 1) via Monte Carlo Simulation (MCS), simulating the sample period (i.e.,
 259 1980-2021; $t = 41$ years) for $n=10,000$ trials. Some research suggests there is empirical evidence
 260 that certain market sectors (i.e., insurance) cannot fully diversify natural disaster risks (inclusive
 261 of flood risk), leading to statistically significant increases (or decreases) in post-disaster returns
 262 (Wang & Kutan, 2013; Bourdeau-Brien & Kryzanowski, 2017). Considering this possibility of
 263 increasing (or decreasing) dependence, we further interrogate the influence of correlation on
 264 CAPM beta values, and investigate how the market risk premium varies with increasing correlation
 265 to market returns (as measured by Kendall's Tau), by repeating this MCS for a range of Kendall's
 266 Tau values (-0.5, 0.5) for $n = 1,000$ trials.

267 **Results**

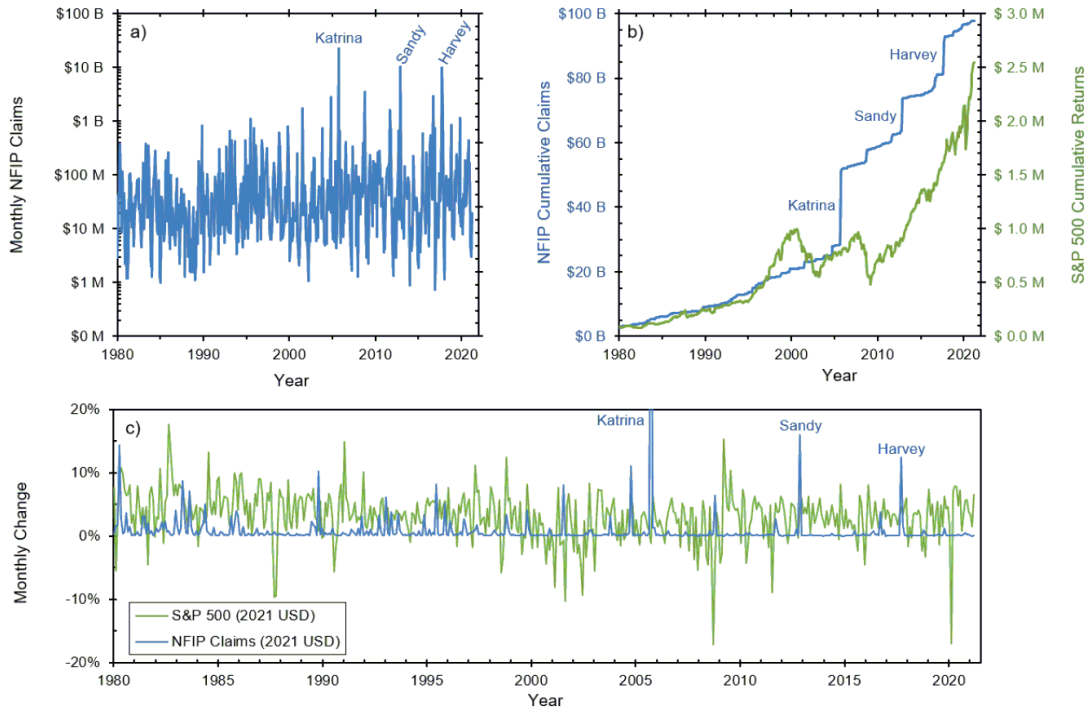
268 Compiling NFIP redacted claims across the U.S. on a monthly basis, we see monthly claims
 269 over the study period ranged from \$1M to \$1B, with the exception of three notable outliers
 270 concurrent with the arrivals of Hurricanes Katrina (August, 2005), Sandy (October, 2012), and

271 Harvey (August, 2017), as shown in



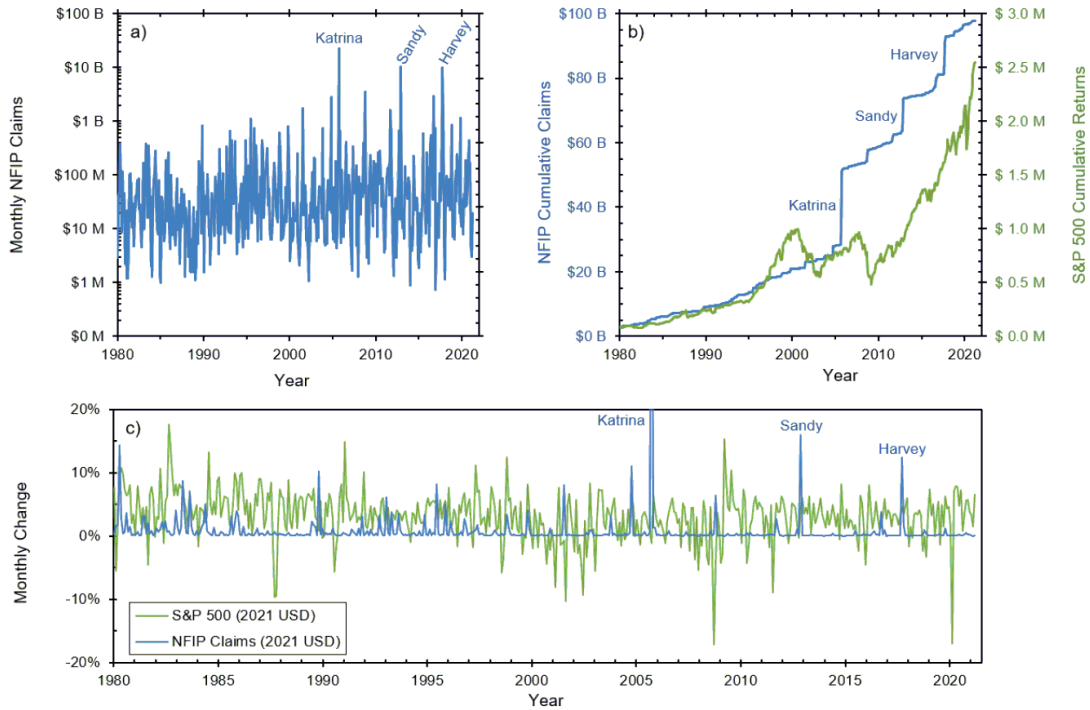
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Fig. 1a. Cumulative NFIP claims and S&P 500 cumulative returns are shown in



275

Fig. 1b, and monthly changes in cumulative NFIP claims and S&P 500 returns are shown

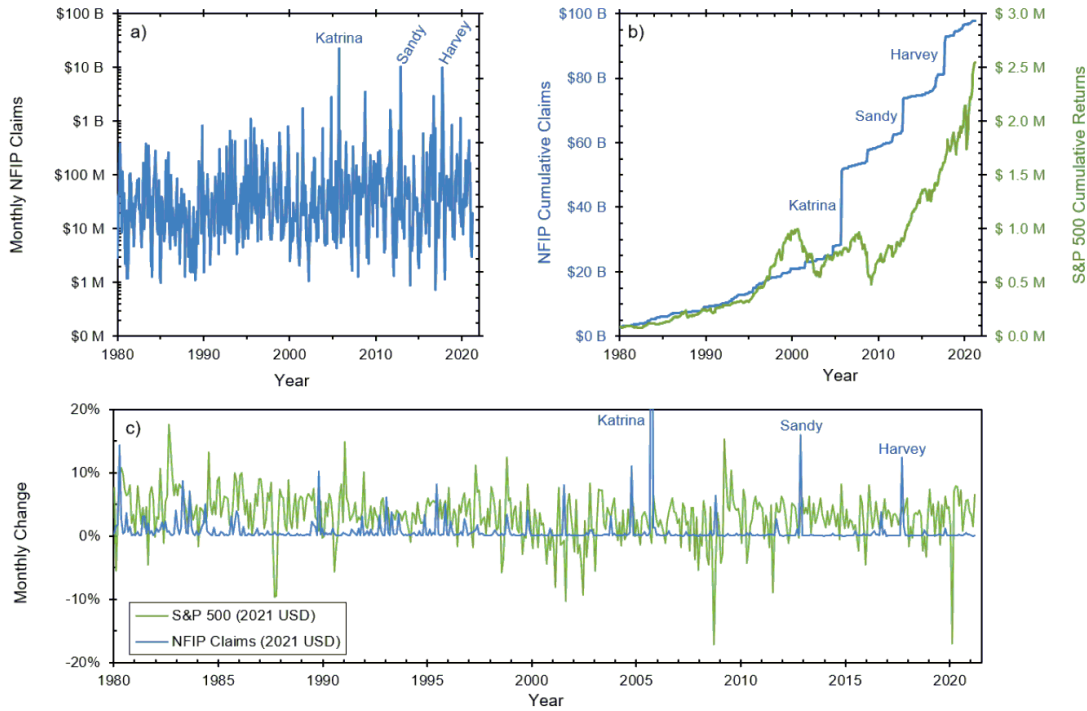


276 in

277 Fig. 1c, with the arrivals of Hurricanes Katrina, Sandy, and Harvey labeled appropriately.

278 Given this sample data, we estimate a market risk premium, $\beta_{sample} = -0.027$, and a Kendall

279 rank correlation coefficient, $\tau_{sample} = 0.0977$.

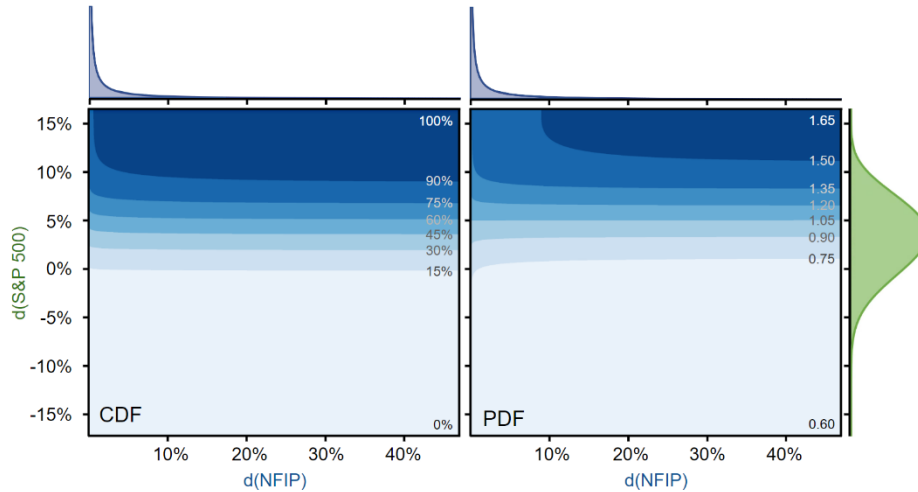


280

281 **Fig. 1.** FEMA National Flood Insurance Program (NFIP) a) monthly claims over time; b)
 282 cumulative claims over time and S&P 500 cumulative returns; c) monthly change in cumulative
 283 NFIP claims and S&P 500 monthly returns over time. All values shown adjusted to 2021 Q2
 284 price levels. (NFIP data from FEMA, 2021; S&P 500 data from Shiller, 2022).

285

286 Relying on equation 3, we estimate a Frank copula dependence parameter, $\theta = 0.93$,
 287 indicative of minimal dependence between the two variables. Using this dependence parameter
 288 and the fitted marginal (general form beta; β) distributions ($\beta(\alpha, \beta, min, max)$): $d(\text{S\&P 500}) =$
 289 $\beta(11.43, 8.00, -0.17, 0.36)$; $d(\text{NFIP}) = \beta(0.032, 3.12, 0, 0.77)$, we construct the Frank
 290 copula shown in Fig. 2 to characterize the joint probability distribution of monthly changes in S&P
 291 500 returns and cumulative NFIP claims.

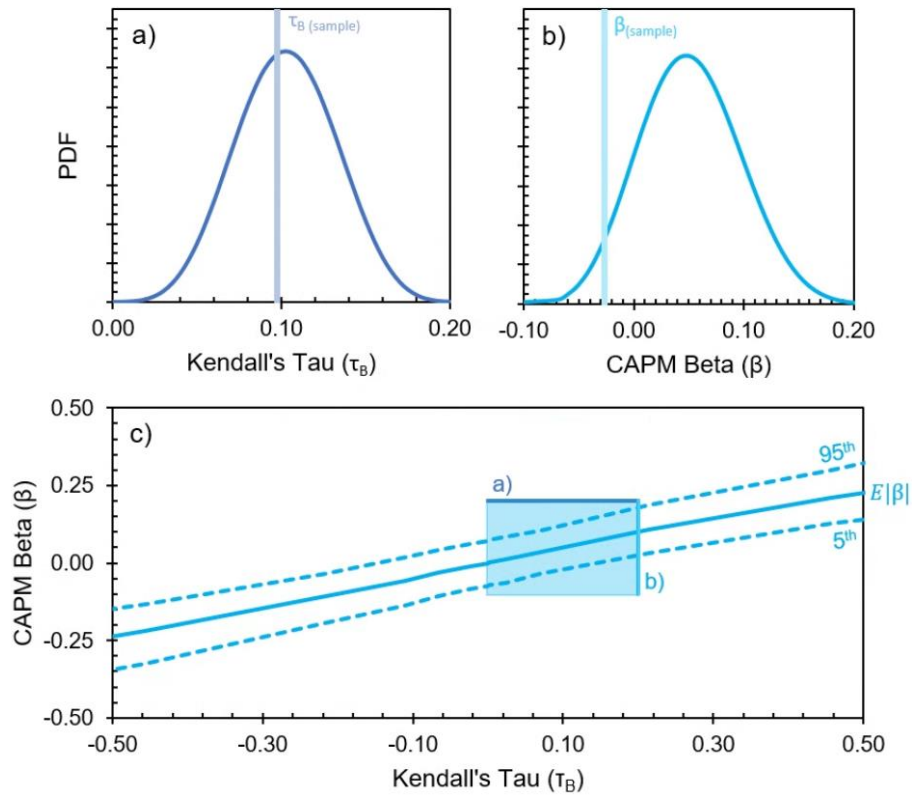


292

293 **Fig. 2.** Joint Cumulative Distribution Function (CDF) and Joint Probability Density Function
 294 (PDF) of monthly S&P 500 returns and monthly change in cumulative NFIP claims as
 295 characterized by a Frank copula ($\theta = 0.93$).

296

297 Given this characterization of joint probability, simulating the 41-year study period (1980-
 298 2021) for $n=10,000$ trials, we develop a wider range of estimates for the Kendall rank correlation
 299 coefficient and market risk premium (CAPM beta). As shown in Fig. 3a, given these marginal
 300 distributions and copula construction, we estimate a Kendall rank correlation coefficient, $\tau =$
 301 0.10 ± 0.11 .



302

303 **Fig. 3.** Estimated a) Kendall's Tau, and b) market risk premium (CAPM Beta) for flood risk-
 304 related cash flows, given observed dependence for $n= 10,000$ trials; c) market risk premium
 305 (CAPM Beta) estimates for flood risk-related cash flows at varying levels of correlation to the
 306 market, as measured by Kendall's Tau

307

308 As a result, we estimate a market risk premium (as defined by the CAPM beta), $\beta =$
 309 0.052 ± 0.20 , indicative of minimal positive (or negative) correlation with broader market
 310 performance. Further investigating the potential systematic market risk premium across a wider
 311 range of potential dependence levels via MCS (recalibrating the Frank copula for a range of
 312 Kendall's rank correlation values), we observe a linear relationship between market risk premia

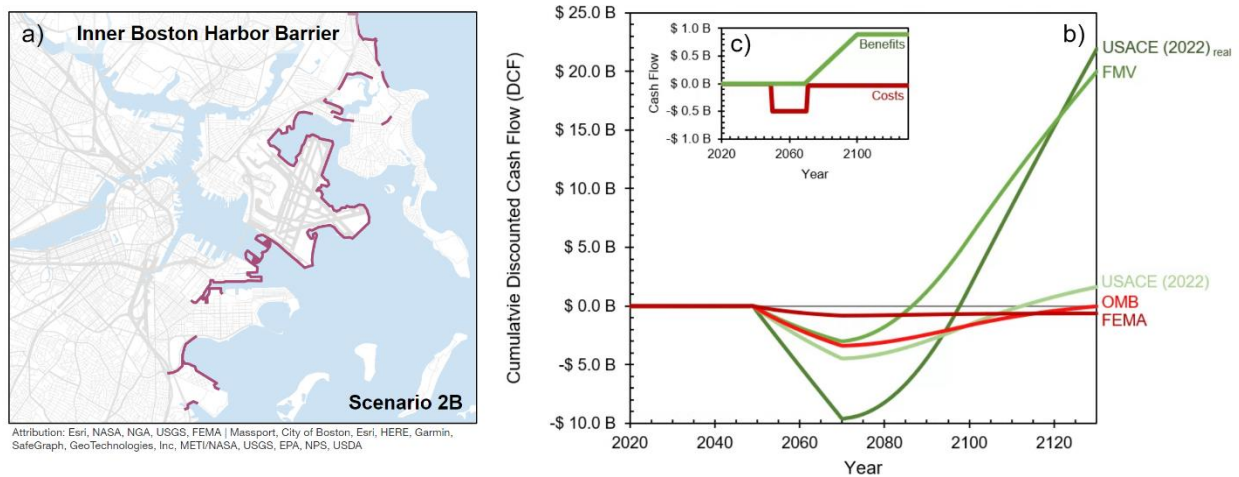
313 and correlation, as measured by Kendall’s rank correlation. We note that given prevailing
314 correlations between global mean sea level (GMSL) and market returns, there is at present little
315 evidence to suggest the market risk premium of flood risk-related cash flows will change over
316 time. Changes in cumulative NFIP claims exhibit little correlation with monthly changes in GMSL
317 over the period for which data is available (1993-2021; NOAA, 2021), $r = -0.02$; $\tau = 0.096$,
318 and monthly changes in GMSL are similarly uncorrelated with monthly S&P500 returns, $r =$
319 0.06 ; $\tau = 0.053$.

320 These results demonstrate the risk associated with NFIP claims, and by proxy flood risk-
321 related cash flows is nearly fully diversifiable under a CAPM pricing framework, and therefore
322 commands a very small market risk premium, $\beta = 0.052$, suggesting an appropriate discount rate
323 of 0.7%, when specified in real terms, given prevailing interest rates at the time of writing. These
324 results largely support the established consensus in the catastrophe bond market (i.e., that
325 catastrophe-related cash flows are uncorrelated with the market), yet the implications for FMV
326 discounting and valuation of climate adaptation projects contrast sharply with prevailing
327 discounting practices presented in the literature, including current U.S. federal policy.

328 *Case Study: Boston Harbor Barrier*

329 To further illustrate this contrast between a FMV approach and prevailing discounting
330 approaches, we investigate the valuation implications for a real-world coastal adaptation case
331 study, a Boston Harbor barrier design proposed by Kirshen et al. (2018). While Kirshen et al.
332 (2018) outline and assess several different potential harbor barrier configurations, construction
333 timelines, and performance assumptions, here we focus on a single example, the ‘inner harbor
334 barrier’ configuration (Scenario 2B, Kirshen et al., 2018) as shown in Fig. 4a. Under this particular

335 scenario, the proposed inner harbor barrier would be constructed over a 20-year period, reaching
 336 completion in 2070 at a present value cost of \$10.0B (2021 Q2 USD), after which operations and
 337 maintenance costs are expected to be \$30M annually. After completion in 2070, the barrier is
 338 anticipated to provide flood risk reduction benefits for coastal flood events exceeding an elevation
 339 El. +4.3 m (El. +14 ft) NAVD88 (it is assumed shore-based measures will be built up to this
 340 elevation by 2070, regardless of whether the barrier is constructed). These benefits are estimated
 341 to increase linearly from 2070 to 2100, reaching an expected annualized value of \$888M,
 342 plateauing thereafter until the end of the analysis period in 2130. Fig. 4c summarizes these
 343 undiscounted cash flows.



344
 345 **Fig. 4.** a) Inner Boston Harbor barrier configuration (Scenario 2B proposed in Kirshen et al.,
 346 2018; Esri, 2023); b) cumulative discounted cash flow (DCF) over time for the Boston Harbor
 347 Barrier example, under several discounting approaches; and c) the undiscounted stream of direct
 348 adaptation costs and flood protection benefits.

349

350 We assess the value of this Boston Harbor barrier configuration under several relevant
351 discounting approaches embedded within U.S. federal policy, contrasting these with a FMV
352 approach and a modified version of current USACE policy, adjusted for inflation. Here, we adjust
353 the nominal discount rate specified by WRDA (1974) by 2% to account for inflation (i.e., to
354 determine the equivalent real discount rate) as to properly discount the future costs and benefits
355 which are specified in real terms. (We note that current USACE valuation guidance nonetheless
356 specifies that real future cash flows be discounted using the nominal cost of borrowing; USACE,
357 2019.) Under the FMV approach, we discount the benefits of this adaptation project (i.e., flood
358 risk reduction) at the 0.7% rate outlined previously and apply a discount rate of 3.3% to project
359 costs, reflecting the expected market risk associated with property and casualty insurance industry
360 investments. Fig. 4b provides the cumulative discounted cash flow (DCF) over the study period
361 for the harbor barrier under these discounting approaches, and **Table 2** summarizes performance
362 across discounting approaches.

363

364 **Table 2.** Applied discounting approaches and associated financial performance of the Boston
365 Harbor Barrier case study

Discounting Approach	Discount Rate, r	NPV [\$B]	BCR
USACE (2021) real	0.25%	21.9	2.99
Fair Market Valuation (FMV)	0.7%, 3.3%	20.0	7.26
USACE (2021)	2.25%	1.6	1.34
OMB (2003)	3.0%	-0.04	0.999
FEMA (2021)	7.0%	-0.63	0.25

366

367

368 We observe significant valuation discrepancies arising from the varied discounting
369 methods, with an inflation-adjusted USACE discounting approach (most consistent with Arrow &
370 Lind, 1970) yielding the highest NPV. The FMV approach yields the next highest NPV, though a
371 higher benefit-cost ratio (BCR) as adaptation investment costs are discounted more heavily, given
372 their higher levels of systematic market risk. Current USACE discounting policy for FY 2022
373 (USACE, 2021) specifies a 2.25% discount rate (based on the nominal yield of long-term U.S.
374 Treasury bonds) and yields a much smaller, though still positive NPV. By contrast, a 3% discount
375 rate (a STPR presented by OMB, 2003; primarily intended for analysis of proposed regulations)
376 yields a negative NPV, though suggests the investment would be very close to breaking even by
377 the end of the analysis period. Relying on the more general 7% discount rate (intended to reflect
378 the expected market rate of return for private capital; OMB, 1992), which is applicable for FEMA
379 funded projects (DHS, 2021), the project has a negative NPV and $BCR = 0.25$.

380 Should the City of Boston choose to advance this particular harbor barrier beyond
381 preliminary analysis, these valuation discrepancies have significant practical implications, as these
382 results suggest FEMA is unlikely to view this adaptation investment as worthy of funding.
383 However, USACE would likely consider the project as a worthwhile investment. Our FMV results
384 suggest that a private sector investor would place even greater value on the adaptation investment
385 opportunity, implying that there may be latent demand for PPP delivery, if an agreement on
386 sufficient payment structure could be reached between the municipality and a private consortium.

387 **Discussion and Conclusion**

388 As demonstrated by the case study, relative to a FMV approach, prevailing federal policy
389 and underlying discounting approaches significantly over-discount and subsequently undervalue

390 future adaptation benefits (i.e., avoided climate-related damage costs). By neglecting the (lack of)
391 systematic risk inherent in flood risk-related cash flows, prevailing discounting approaches
392 consider future flood damages to be less important (or less “painful”) than a FMV approach would
393 otherwise suggest. Consequently, the avoidance of future flood damages is comparatively less
394 valuable. Further, these results suggest the prevailing discounting approaches currently employed
395 by US federal agencies best positioned to finance climate change adaptation projects
396 systematically undervalue flood risk reduction benefits, relative to the private sector.

397 Such a systematic undervaluation of adaptation benefits is liable to result in a misallocation
398 of capital away from a subset of adaptation projects with a positive fair market value but nominally
399 negative net present value, potentially imposing a “deadweight loss” onto society (Gollier, 2021).
400 The resulting systematic bias towards underinvestment in climate change adaptation projects
401 favors *reactive* spending in response to climate-related disasters, rather than *proactive* investments
402 aimed at minimizing future climate-related damage costs. This bias towards underinvestment
403 extends beyond the U.S., insofar as prevailing discounting policies in other countries are founded
404 on similar philosophies and approaches that neglect systematic market risk associated with project
405 cash flows. Though unlikely, adoption of a FMV discounting approach at the federal level would
406 empower key federal agencies to properly consider systematic market risk when valuing climate
407 adaptation projects, while simultaneously restructuring broader federal discounting practices to
408 better align with the philosophy underlying prevailing OMB guidance (OMB, 1992).

409 While relevant portions of federal policy codified in public law are unlikely to change in
410 the near-term, current case law suggests federal agencies have the authority and latitude to diverge
411 from OMB guidance with sufficient justification (Viscusi, 2022), suggesting that there is already

412 enough latitude to implement a FMV discounting approach in certain situations within a U.S.
413 context. Further, nonfederal U.S. public agencies are likely to have even greater latitude in
414 determination of internal discounting policy. Even in instances where non-federal public agencies
415 are required to follow prevailing federal discounting policy to satisfy federal funding requirements,
416 a FMV approach can nonetheless still be employed to develop separate valuations for internal
417 purposes and decision-making processes. Public agencies typically have the latitude formulate
418 their own capital investment criteria (Chen & Bartle, 2022) and therefore have the capacity to
419 adopt a fair market valuation approach when planning, prioritizing, and budgeting for future capital
420 investments. Valuing projects in a manner that adequately reflects the associated systematic risks
421 (or lack thereof) enables public agencies to better estimate the societal value of a project and make
422 more informed capital allocation decisions, thereby more effectively utilizing public funds.

423 In principle, the application of the proposed FMV discounting approach can be readily
424 extended beyond the context of adaptation investments, provided sufficient data is available to
425 estimate the systematic (i.e., market-correlated) risk of underlying project cash flows (e.g., climate
426 mitigation projects, should stakeholders reach agreement on the ‘climate beta’; Dietz et al., 2018).
427 However, given that the current valuation framework as presented relies on traditional discounted
428 cash flow analysis to arrive at an NPV, the current valuation framework is unable to consider inter-
429 annual variability in systematic risk levels, or embedded project optionality (In et al., 2022). The
430 valuation framework presented could be extended to allow for certainty-equivalent discounting,
431 wherein future cash flows are first discounted to account for systematic risk (or probabilistically
432 adjusted to account for embedded optionality), then subsequently discounted at the risk free rate
433 (Brealey et al., 2011). Such extensions would allow for consideration of adaptation projects with
434 more complex cash flows and temporally dependent market correlations.

435 Beyond these limitations, we further note that operationalizing the proposed FMV
436 approach is sensitive to changes in both the expected market rate of return and the risk-free rate.
437 Consequently, as broader market conditions change and influence these factors over time, any
438 resulting discount rates will also change, implying sensitivity to the base year of analysis. More
439 generally, the sensitivity of the discount rate to these key input parameters suggests that reliable
440 and timely estimation of the expected market rate of return and risk-free rate is imperative for
441 accurately capturing prevailing market conditions and effectively operationalizing the proposed
442 FMV approach.

443 Extending beyond the public sector, the proposed FMV approach outlined in this paper is
444 notably directly applicable to private sector investments, and is to the authors' knowledge, the first
445 such approach proposed within the climate change adaptation literature. Ultimately, as Li et al.
446 (2014) note, clear communication and transparency surrounding discounting assumptions in cost-
447 benefit analyses is a key aspect of legitimizing the business case for a climate adaptation project,
448 irrespective of which discount rate(s) are applied for valuation of (public or private) climate change
449 adaptation investments.

450 **Declaration of Competing Interest**

451 The authors declare that they have no known competing financial interests or personal
452 relationships that could have appeared to influence the work reported in this paper.

453 **Data Availability Statement**

454 Some or all data, models, or code that support the findings of this study are available from the
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