Future flood losses in major coastal cities

Stephane Hallegatte^{1,2}*, Colin Green³, Robert J. Nicholls⁴ and Jan Corfee-Morlot⁵

Flood exposure is increasing in coastal cities^{1,2} owing to growing populations and assets, the changing climate³, and subsidence⁴⁻⁶. Here we provide a quantification of present and future flood losses in the 136 largest coastal cities. Using a new database of urban protection and different assumptions on adaptation, we account for existing and future flood defences. Average global flood losses in 2005 are estimated to be approximately US\$6 billion per year, increasing to US\$52 billion by 2050 with projected socio-economic change alone. With climate change and subsidence, present protection will need to be upgraded to avoid unacceptable losses of US\$1 trillion or more per year. Even if adaptation investments maintain constant flood probability, subsidence and sea-level rise will increase global flood losses to US\$60-63 billion per year in 2050. To maintain present flood risk, adaptation will need to reduce flood probabilities below present values. In this case, the magnitude of losses when floods do occur would increase, often by more than 50%, making it critical to also prepare for larger disasters than we experience today. The analysis identifies the cities that seem most vulnerable to these trends, that is, where the largest increase in losses can be expected.

A first screening study¹ provided a global overview of flood exposure in world coastal cities. The exposure metric can be viewed as a worst case scenario, but it does not estimate average annual losses, which is a standard metric in disaster risk management planning. To do so, it is necessary to take into account infrastructure-based adaptation (for example, dykes) and the vulnerability of populations and assets. Here, we assess economic average annual losses (AAL) in 136 coastal port cities, using a method developed for assessing city-level flood risk⁷ and a new database of urban coastal protection (Methods).

Present aggregated average annual flood losses in the 136 cities are estimated at approximately US\$6 billion per year. Table 1 ranks the most vulnerable cities in 2005 using two different metrics of vulnerability. In the left column, the table shows a ranking in terms of AAL, taking into account all potential floods and existing protection. The AAL estimates can be compared to more sophisticated approaches. For instance, the annual losses in New Orleans are estimated at US\$600 million, close to the US\$650 million estimates from the Interagency Performance Evaluation Taskforce⁸. In the right column, cities are ranked according to relative vulnerability, namely the ratio of AAL to the city's gross domestic product (GDP). This value can be understood as the share of the city's economic output that should be saved annually to pay for future flood losses. The 20 cities most vulnerable according to this last indicator are also presented in Fig. 1.

The ranking in terms of exposure includes mainly rich-country cities (Supplementary Table S4). On average, however, rich cities

are better protected than poorer ones, and the ranking in terms of absolute flood losses contains more cities from developing countries. In relative terms, developing-country cities are even more vulnerable, with only three cities from developed countries in the top 20 (New Orleans, Miami and Tampa—Saint-Petersburg). Moreover the ranking in absolute terms (left column) includes mainly capital cities, whereas secondary cities are more often represented in the ranking in relative terms (right column). This difference suggests that risk management efforts may be lower in secondary cities.

Table 1 shows the importance of existing flood defences: in a city such as Amsterdam, exposure is extremely high (US\$83 billion of assets exposed to the 100-year flood), but AAL do not exceed US\$3 million, because estimated defence standards are the highest that exist globally. On the other hand, a city such as Ho Chi Minh, in Vietnam, has a 100 year exposure of only US\$18 billion, but the lower level of protection means that the city is affected by small floods on a frequent basis, resulting in large estimated average costs. In relative terms, Ho Chi Minh City has one of the largest vulnerabilities, with AAL reaching 0.74% of local GDP. The ratio of AAL to local GDP exceeds 1% for two cities, Guangzhou and New Orleans. The vulnerability of New Orleans has been reduced however by recent post-Hurricane Katrina investments and is likely to be reduced further in the near future9.

Another conclusion from Table 1 is the concentration of losses in only a few cities. Only 13 cities have average losses in excess of US\$100 million, and three American cities (Miami, New York City and New Orleans) explain 31% of the global aggregate losses in the 136 cities, because of their high wealth and low protection level. Adding Guangzhou, the four top cities explain 43% of global losses. Also, the US seems particularly vulnerable, with 6 American cities in exposure ranking, 8 in the ranking by absolute AAL, and 3 in the ranking by relative AAL. As coastal flood risks are highly concentrated, flood reduction actions in a few locations could be very cost-effective.

To develop possible future patterns of drivers of risk to 2070, our analysis introduces three scenarios for socio-economic changes and six for environmental change. From there, we retain four main scenarios: SEC assumes only socio-economic changes, derived from OECD and UN scenarios; SEC-S adds subsidence to scenario SEC (40 cm in 2050 in the cities subjected to subsidence); and SLR-1 and SLR-2 add optimistic and pessimistic sea-level rise scenarios to SEC-S, respectively (with 20 cm and 40 cm in 2050). Here, we report results for 2050, but results for 2030 and 2070 are available in the Supplementary Information.

With no adaptation, the projected increase in average losses by 2050 is huge, with aggregate losses increasing to more than US\$1 trillion per year in scenarios SLR-1 and SLR-2 (Supplementary Table S6). All cities experience a similar increase

1

¹The World Bank, Sustainable Development Network, Washington DC 20433, USA, ²Centre International de Recherche sur l'Environnement et le Développement (CIRED), Nogent-sur-Marne 94736, France, ³Flood Hazard Research Centre, Middlesex University, London NW4 4BT, UK, ⁴Faculty of Engineering and the Environment, University of Southampton, Southampton SO17 1BJ, UK, ⁵Organisation for Economic Co-operation and Development, Paris 75775, France. *e-mail: shallegatte@worldbank.org

| Table 1 | City ranking b | y risk (AAL) and relative risk (| (AAL in percentage of GDP) for 2005. |
|---------|----------------|----------------------------------|--------------------------------------|
|---------|----------------|----------------------------------|--------------------------------------|

| | Ranking | g by AAL (US | \$\$ million) | | Ranking by relative AAL (percentage of city GDP) | | | | | | |
|----|------------------------|----------------------|-------------------------------------|---|--|------------------------|----------------------|---|---|--|--|
| | Urban agglomeration | 100 year exposure | AAL, with protection (US\$ million) | AAL, with protection (percentage of GDP) | | Urban agglomeration | 100 year exposure | AAL, with protection (US\$ million) | AAL, with protection (percentage of GDP) | | |
| 1 | Guangzhou | 38,508 | 687 | 1.32% | 1 | Guangzhou | 38,508 | 687 | 1.32% | | |
| 2 | Miami | 366,421 | 672 | 0.30% | 2 | New Orleans | 143,963 | 507 | 1.21% | | |
| 3 | New York—Newark | 236,530 | 628 | 0.08% | 3 | Guayaquil | 3,687 | 98 | 0.95% | | |
| 4 | New Orleans | 143,963 | 507 | 1.21% | 4 | Ho Chi Minh City | 18,708 | 104 | 0.74% | | |
| 5 | Mumbai | 23,188 | 284 | 0.47% | 5 | Abidjan | 1,786 | 38 | 0.72% | | |
| 6 | Nagoya | 77,988 | 260 | 0.26% | 6 | Zhanjiang | 2,780 | 46 | 0.50% | | |
| 7 | Tampa—St. Petersburg | 49,593 | 244 | 0.26% | 7 | Mumbai | 23,188 | 284 | 0.47% | | |
| 8 | Boston | 55,445 | 237 | 0.13% | 8 | Khulna | 2,073 | 13 | 0.43% | | |
| 9 | Shenzen | 11,338 | 169 | 0.38% | 9 | Palembang | 1,161 | 27 | 0.39% | | |
| 10 | Osaka—Kobe | 149,935 | 120 | 0.03% | 10 | Shenzen | 11,338 | 169 | 0.38% | | |
| 11 | Vancouver | 33,456 | 107 | 0.14% | 11 | Hai Phòng | 6,348 | 19 | 0.37% | | |
| 12 | Tianjin | 11,408 | 104 | 0.24% | 12 | N'ampo | 507 | 6 | 0.31% | | |
| 13 | Ho Chi Minh City | 18,708 | 104 | 0.74% | 13 | Miami | 366,421 | 672 | 0.30% | | |
| 14 | Kolkata | 14,769 | 99 | 0.21% | 14 | Kochi | 855 | 14 | 0.29% | | |
| 15 | Guayaquil | 3,687 | 98 | 0.95% | 15 | Tampa—St. Petersburg | 49,593 | 244 | 0.26% | | |
| 16 | Philadelphia | 22,132 | 89 | 0.04% | 16 | Nagoya | 77,988 | 260 | 0.26% | | |
| 17 | Virginia Beach | 61,507 | 89 | 0.15% | 17 | Surat | 3,288 | 30 | 0.25% | | |
| 18 | Fukuoka—Kitakyushu | 39,096 | 82 | 0.09% | 18 | Tianjin | 11,408 | 104 | 0.24% | | |
| 19 | Baltimore | 14,042 | 76 | 0.08% | 19 | Grande_Vitória | 6,738 | 32 | 0.23% | | |
| 20 | Jakarta | 4,256 | 73 | 0.14% | 20 | Xiamen | 4,486 | 33 | 0.22% | | |

A comparison with a ranking by exposure is proposed in the Supplementary Information.

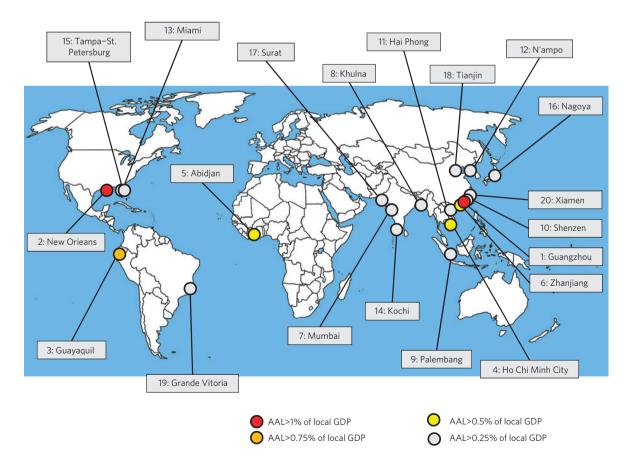


Figure 1 | The 20 cities where the relative risk is larger in 2005, that is, where the ratio of AAL with respect to local GDP is the largest. More information in Table 1.

Table 2 | The 20 cities with the highest loss in 2050, assuming scenario SLR-1 and adaptation option that maintains flood probability (option PD).

| | Scenarios with soci change alone (SEC) | | Scenarios with socio-economic change, subsidence, sea-level rise and adaptation to maintain flood probability (scenarios SLR-1, and adaptation option PD) | | | | | |
|----------------------|---|--------------------------------------|---|--|------------------------------|--|--|--|
| Urban agglomeration | AAL (US\$ million) | AAL (per- centage of city GDP) | AAL (US\$ million) | Increase in AAL compared with 2005 (%) | AAL (percentage of city GDP) | | | |
| Guangzhou (S) | 11,928 | 1.32% | 13,200 | 11% | 1.46% | | | |
| Mumbai | 6,109 | 0.47% | 6,414 | 5% | 0.49% | | | |
| Kolkata (S) | 2,704 | 0.21% | 3,350 | 24% | 0.26% | | | |
| Guayaquil (S) | 2,813 | 0.95% | 3,189 | 13% | 1.08% | | | |
| Shenzen | 2,929 | 0.38% | 3,136 | 7% | 0.40% | | | |
| Miami | 2,099 | 0.30% | 2,549 | 21% | 0.36% | | | |
| Tianjin (S) | 1,810 | 0.24% | 2,276 | 26% | 0.30% | | | |
| New York—Newark | 1,960 | 0.08% | 2,056 | 5% | 0.08% | | | |
| Ho Chi Minh City (S) | 1,743 | 0.74% | 1,953 | 12% | 0.83% | | | |
| New Orleans (S) | 1,583 | 1.21% | 1,864 | 18% | 1.42% | | | |
| Jakarta (S) | 1,139 | 0.14% | 1,750 | 54% | 0.22% | | | |
| Abidjan | 826 | 0.72% | 1,023 | 24% | 0.89% | | | |
| Chennai (Madras) | 825 | 0.12% | 939 | 14% | 0.14% | | | |
| Surat | 905 | 0.25% | 928 | 3% | 0.26% | | | |
| Zhanjiang (S) | 806 | 0.50% | 891 | 11% | 0.55% | | | |
| Tampa—St. Petersburg | 763 | 0.26% | 859 | 13% | 0.29% | | | |
| Boston | 741 | 0.13% | 793 | 7% | 0.14% | | | |
| Bangkok (S) | 596 | 0.07% | 734 | 23% | 0.09% | | | |
| Xiamen (S) | 572 | 0.22% | 729 | 27% | 0.29% | | | |
| Nagoya (S) | 564 | 0.26% | 644 | 14% | 0.30% | | | |

in risk. In the absence of adaptation, the impact of environmental change is much larger than the effect of socioeconomic change. These numbers should not be considered as predictions, but they demonstrate the need for adaptation, because inaction would result in unacceptably high losses.

We then consider adaptation and how it will alter losses. We assume first that adaptation action increases coastal flood defences to maintain a constant probability of flooding (adaptation option: present design, PD). The increase in aggregate AAL is much lower in this case. Owing to socio-economic change, there is still a ninefold increase in aggregate losses, from US\$6 to US\$52 billion per year, but this is made more manageable by the fact that these cities are also much richer. However, rising water levels still increase AAL: subsidence by 12% and sea-level rise by an additional 2–8%, reaching between US\$60 and 63 billion per year.

Table 2 shows the top 20 cities in terms of AAL in 2050 in the scenario with subsidence and optimistic sea-level rise (SLR-1), with adaptation to maintain present flood probability. Guangzhou remains the most vulnerable city, with AAL exceeding US\$13 billion. With socio-economic change alone, AAL in Guangzhou would be around US\$12 billion per year in 2050 (a 17-fold increase in absolute terms). Subsidence and sea-level rise are thus responsible for an additional 10% increase, that is, a 10% increase in the AAL-to-GDP ratio. Indeed, even if the probability of coastal flooding is unchanged thanks to upgraded coastal defence infrastructure, the fact that a larger share of existing assets is protected by these defences means that annual losses will rise relative to local GDP. For instance, sea-level rise and subsidence increase the AAL-to-GDP ratio by 54% in Jakarta and by 24% in Abidjan even if present flood probabilities are maintained thanks to better defences. In other words, the world sees no more floods, but each flood is more destructive owing to sea-level rise and subsidence, even with better defences. This effect reinforces a trend that can be expected from socio-economic change alone, even in the absence of environmental change.¹⁰

Figure 2 shows the 20 cities where the increase in average annual losses between 2005 and 2050 is greatest in relative terms; detailed numerical values are provided in Supplementary Table S7. In Alexandria, for instance, maintaining flood probability leads to an increase by 154% in AAL. These most vulnerable cities are distributed all over the world, with a concentration in the Mediterranean Basin, the Gulf of Mexico and East Asia. Even though absolute levels of risk are sometimes low in these cities, they can be considered as adaptation hotspots because this is where flood risks are likely to increase the most in relative terms.

To avoid any increase in risk, an adaptation policy needs to do more than maintain present flood probability. Rather, maintaining present levels of risk (relative to local GDP) in the context of rising sea levels, subsidence and socio-economic changes requires adaptation policy that reduces flood probability over time. In the adaptation option termed present losses, an upgrade in defence is thus calibrated to cancel the impact of environmental changes and to maintain present losses on average relative to local wealth, keeping aggregate losses at US\$52 billion.

For each city, we estimate the increase in defence standard that would maintain the relative risk level (that is, keep constant the ratio of average annual losses to local GDP). The required increase in protection is larger than local sea-level rise. For instance, in Alexandria, protection needs to be raised by 67 cm, for a 60 cm rise in local sea level; this corresponds to moving from a 100-year design standard to a 270-year design standard (that is, a division by 2.7 of the probability of flooding). In other cities (Supplementary Table S7) the increase in dyke height is between 2 and 8 cm larger than sea-level rise. This increase corresponds to a significant increase in the standard of protection, that is, to a large decrease

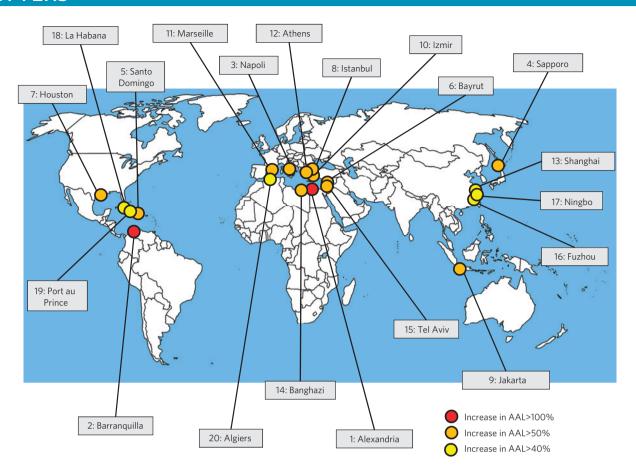


Figure 2 | The 20 cities where AAL increase most (in relative terms in 2050 compared with 2005) in the case of optimistic sea-level rise, if adaptation only maintains present defence standards or flood probability (PD). More information in Supplementary Table S7.

in the probability of flooding (for example, a division by 1.6 in Jakarta and Shanghai).

Even if the relative risk level is maintained, flood consequences would be much higher in a world with sea-level rise and better protection. In Alexandria in 2050, for instance, the probability of a flood may decrease by 2.7 owing to better defences, but if a flood occurs the total losses triple from US\$17 billion (with socio-economic change alone) to US\$51 billion, a tripling due to environmental changes alone. Results for other cities are presented in Supplementary Table S7. Many cities would experience losses that are more than 50% larger if an event exceeds the post-adaptation protection level. Hence, the world's coastal cities become more dependent on flood defences, but also more vulnerable once failure or overtopping occurs.

Finally, Table 3 looks at present city characteristics that influence the vulnerability in 2050. It shows that cities that grow rapidly, have large populations, are poor, exposed to tropical storms, and prone to subsidence are over-represented in the top 20 for absolute AAL (the cities from Table 2). However, these drivers are not as relevant for the relative increase in annual average loss with adaptation option PD (the cities from Fig. 2). Only subsidence seems to be a consistently good determinant of vulnerability for both absolute and relative measures of change, with twice as many cities with subsidence in the top 20 based on the two indicators (absolute and relative losses). The cities most vulnerable in relative terms may thus not be the ones suggested by the present situation and historic floods, nor are they the ones that necessarily attract the most research and analysis today with respect to managing risk.

While recognizing the limitations and uncertainties in this analysis, three important policy conclusions can be drawn that are robust across a range of plausible assumptions. First, failing

to adapt is not a viable option in coastal cities. It is difficult to estimate the cost of these adaptation options, as it depends on the specific context for each city and on selected approaches and technologies. On the basis of anecdotal evidence from a few cities¹¹⁻¹⁴, a few billion US dollars per city in initial investment—plus approximately 2% of the initial investment cost in annual operation and maintenance costs—is the possible order of magnitude for adaptation costs¹⁵. Hence, indicative annualized values with 5% interest rates are about US\$350 million per year per city, or approximately US\$50 billion per year for our 136-city sample. These estimated aggregate adaptation costs are far below our estimate of aggregate damage losses per year in the absence of adaptation, and of the same order of magnitude as residual losses with adaptation (Supplementary Table S6). These estimates include only flood risks and do not encompass all weather risks that these cities face; see, for instance, ref. 16 for a global analysis of wind damage from tropical cyclones.

Second, managing coastal flood risk requires doing more than maintaining today's standard of protection (and present probability of flooding). In practice, probability of flooding will need to be reduced to maintain flood risks at today's levels.

Last, improving standards of protection could maintain or reduce risk levels and decrease the number of floods, but the magnitude of losses when floods do occur will still increase. This result points to the limitations of what infrastructure-based adaptation can achieve. As illustrated by the recent landfall of hurricane Sandy on the east coast of the United States, there is a need to prepare at the local, national and international level for larger floods and the disasters that ensue. Such preparations can include strengthening disaster planning measures, including early warning and evacuation systems, more comprehensive insurance

Table 3 | The fraction of cities in the five categories (fast growing, large population, low income, subject to tropical storms and subject to subsidence) in the full sample of 136 cities, and in two top 20 categories using different indicators.

| | Fast-growing city (local GDP growth in 2005 larger than 5% per year) | Large population (larger than 5 million in 2005) | Low income (GDP per capita in 2005 less than US\$5,000) | Subject to hurricane and tropical storms | Prone to significant subsidence |
|---|---|---|--|--|---------------------------------------|
| All 136 cities Top 20 in terms of AAL in 2050 (adaptation option PD; Table 2) | 40% 65% | 19% 55% | 27% 40% | 51% 95% | 27% 55% |
| Top 20 in terms of increase in AAL (adaptation option PD; Fig. 2) | 40% | 15% | 20% | 45% | 40% |

schemes and other forms of post-disaster response to quickly rebuild affected communities.

Methods

Flood risks are analysed following ref. 7. The population exposure is taken from a previous analysis¹. Exposed population was translated into exposed assets using an estimate of produced capital per inhabitant drawing on recent work from the World Bank¹⁷. The DIVA database provides information about extreme water levels¹⁸. We create a first database for coastal defences; this is based on collected evidence on existing defences where possible, and the authors' expert estimates to complete the defence database (Supplementary Information).

Then, we calculate the probability of different flood levels in each city (within the flood defences) using three simple models for defence failure or overtopping. Even though absolute risk levels depend on the failure model, the relative effect of sea-level rise and subsidence on losses is relatively robust to this uncertainty.

From the probability of the various flood levels in the city, and from data on assets that are exposed at different flood levels, flood asset losses are estimated using depth–damage functions for 6 categories of assets.

To assess future losses, we use two socio-economic scenarios¹⁹: an OECD-based growth scenario in which urban populations grow at the same rate in all cities, following an extrapolation of UN urbanization scenarios; and the OECD-based growth scenario in which city population is capped at 35 million inhabitants. We assume that future assets in the city have the same elevation distribution as existing assets.

The six scenarios of environmental change combine: two assumptions on subsidence (no subsidence, or a 40 cm subsidence in 2050 in all cities subject to it); and three assumptions about sea-level rise (none, 20 cm, or 40 cm in 2050). Combined with socio-economic scenarios and adaptation options, these lead to 108 scenario combinations. Results for the 108 scenarios and input data and model codes are available in the Supplementary Information. The analysis presented here considers only four scenarios of the total number of scenarios (see text). All scenarios use the simplest defence failure model, the maximum protection level, and constrain cities to no more than 35 million inhabitants.

Received 4 December 2012; accepted 16 July 2013; published online 18 August 2013

References

- Hanson, S. et al. A global ranking of port cities with high exposure to climate extremes. Climatic Change 104, 89–111 (2011).
- De Sherbinin, A., Schiller, A. & Pulsipher, A. The vulnerability of global cities to climate hazards. Environ. Urban. 19, 39–64 (2007).
- Nicholls, R. J. et al. in IPCC Climate Change 2007: İmpacts, Adaptation and Vulnerability (eds Parry, M. L., Canziani, O. F., Palutikof, J. P., van der Linden, P. J. & Hanson, C. E.) 315–356 (Cambridge Univ. Press, 2007).
- Nicholls, R. J. Coastal megacities and climate change. GeoJournal 37, 369–379 (1995).
- Dixon, T. H. et al. Space geodesy: Subsidence and flooding in New Orleans. Nature 441, 587–588 (2006).
- Climate Risks and Adaptation in Asian Coastal Megacities (The World Bank, 2010).

- Hallegatte, S. et al. Assessing climate change impacts, sea level rise and storm surge risk in port cities: A case study on Copenhagen. Climatic Change 104, 113–137 (2011).
- 8. http://nolarisk.usace.army.mil/.
- Link, L. E. The anatomy of a disaster, an overview of Hurricane Katrina and New Orleans. Ocean Eng. 37, 4–12 (2010).
- Hallegatte, S. An Exploration of the Link between Development, Economic Growth, and Natural Risk Policy Research Working Paper No. 6216 (The World Bank, 2012).
- 11. Evans, E. et al. Proc. Inst. Civil Eng. 159, 53-61 (2006).
- 12. Kabat, P. et al. Dutch coasts in transition. Nature Geosci. 2, 450–452 (2009).
- Kates, R. W., Colten, C. E., Laska, S. & Leatherman, S. P. Reconstruction of New Orleans after Hurricane Katrina: A research perspective. *Proc. Natl Acad. Sci. USA* 103, 14653–14660 (2006).
- Ammerman, A. J. & McClennen, C. E. Saving Venice. Science 289, 1301–1302 (2000).
- 15. Nicholls, R., Brown, S., Hanson, S. & Hinkel, J. Economics of Coastal Zone Adaptation to Climate Change (The World Bank, 2010).
- Peduzzi, P. et al. Global trends in tropical cyclone risk. Nature Clim. Change 2, 289–294 (2012).
- 17. The Changing Wealth of Nations: Measuring Sustainable Development in the New Millennium (The World Bank, 2010).
- Vafeidis, A. T. et al. A new global coastal database for impact and vulnerability analysis to sea-level rise. J. Coast. Res. 917–924 (2008).
- Chateau, J., Rebolledo, C. & Dellink, R. An Economic Projection to 2050: The OECD 'ENV-Linkages' Model Baseline No. 41 (OECD, 2011).

Acknowledgements

These are early results from an ongoing OECD research project. The authors acknowledge support from the OECD for this research and support from J. Chateau to provide the socio-economic scenario data from the OECD's ENV-Linkages model. C.G. acknowledges support from the AVOID project. The views shared in this article represent those of the authors and are not intended to represent the views of the OECD or of its Member countries, or the view of World Bank, its executive directors, or the countries they represent.

Author contributions

The four authors designed the study, interpreted results and authored the paper. S.H. developed and ran the models. R.N. and C.G. provided expert input on depth–damage curves and coastal protection.

Additional information

Supplementary information is available in the online version of the paper. Reprints and permissions information is available online at www.nature.com/reprints. Correspondence and requests for materials should be addressed to S.H.

Competing financial interests

The authors declare no competing financial interests.

DOI: 10.1038/NCLIMATE1979

1

Future flood losses in major coastal cities

- ¹ The World Bank, Sustainable Development Network, Washington DC, USA.
- ² Centre International de Recherche sur l'Environnement et le Développement (CIRED), Paris, France.
- ³ Flood Hazard Research Centre, Middlesex University, London, UK.
- ⁴ Faculty of Engineering and the Environment, University of Southampton, Southampton, UK.
- ⁵ Organisation for Economic Co-operation and Development, Paris, France.
- *Correspondence to: shallegatte@worldbank.org

Supplementary Methods

A first screening study by Hanson and colleagues¹ provided a global overview of coastal flood exposure in world coastal cities, including rankings. That study considered several drivers of floods including demographic and socio-economic changes (including urbanization), climate-induced sea-level rise, and human-induced subsidence where appropriate.

The methodology was based on determining the numbers of people and the value of assets that would be exposed to extreme water levels in the absence of coastal defenses and protection. The reference extreme water level was the 100-year coastal flood event. This metric of exposure reveals much about the risks faced in each city, because people in the flood plain rely on formal or informal flood defenses, and thus will be at some level of risk. This risk could arise from a failure of existing flood defenses due to breaching, or from a high return-period event which exceeds the height of existing protection and overtops the defense. In other words the exposure metric can be viewed as a worst case scenario, and exposure can translate into major losses during extreme events (e.g. New Orleans and Hurricane Katrina in 2005).

While the first screen exercise¹ considered how exposure will change in response to socio-economic drivers of economic and population growth, and in response to environmental changes (e.g., sea-level rise, subsidence), a more interesting and useful question is how losses would evolve. To look at economic losses, it necessary to take into account infrastructure-based adaptation (e.g., upgrading dikes and sea walls) and consider how these actions might be taken over time to mitigate flood risk and reduce losses from a city to a global scale. This is what is done by the present analysis.

1. Assessing current flood risks

The investigation took the form of an elevated-based GIS (Geographical Information Systems) analysis. ^{2,3}

1.1. Current population and population exposure

Population exposure is taken from Hanson and colleagues, following the methodology used in previous studies.^{1,4} In each 50 cm "elevation layer" from current mean sea level (e.g., the area located between 0.5 and 1 m above normal sea level), exposed population is computed using topographic and population data.

Topographic data is the 90m resolution data from the Shuttle Radar Topography Mission (SRTM), except in the USA where 30m SRTM data is available, and in the UK, where a 10m Digital Elevation Model (provided by Infoterra) was used. Population data for the selected cities were taken from Landscan 2002 and constrained using city extents from post code data. Postcodes were largely taken from Risk Management Solutions (RMS) geocoding data and, in the USA, Metropolitan Statistical Areas (MSAs) from Census. Where postcode data were unavailable, internet-based city maps were used. The 1km resolution Landscan 2002 data was resampled to 100m for all cities, with the exception of those in the USA and UK, which were resampled to 30m. From this process, we obtain the number of inhabitants who would be flooded by various water levels, assuming no protection and uniform inundation.

At the pixel level, the SRTM elevation data can have errors of up to 10 m, which is large compared with sea level changes we are considering. These errors, however, are much lower in flat areas, where flood risks are concentrated, and have a large long-wavelength component (at the continent scale) that is not a problem when investigating local elevation differences; complete analyses are available in the literature. However, this dataset is not adequate for the engineering design of sea walls and dykes. Nevertheless, when aggregated over larger areas (e.g., neighbourhoods), this data is able to provide a fair estimate of the elevation and can, therefore, be used to estimate the exposed population and assets and to rank cities according to their risk level to identify where detailed analyses are most necessary.

1.2. Asset exposure

The exposed population was translated into exposed assets using an estimate of the amount of capital per inhabitant. This capital per inhabitant was computed from the GDP per capita in each country and an estimate of the ratio of "produced capital" to GDP.

The ratio of produced capital to GDP is calculated using the World Bank dataset published with the "Changing Wealth of the Nations" report⁶. As shown in Figure S1, there is almost a linear relationship between the two. To calculate the average ratio, we averaged the ratio of produced capital to GDP for all countries, with a weight calculated on the basis of each country's population. The resulting ratio is equal to 2.8 and is applied to all countries. This ratio is significantly lower than the value of 5 used in the previous exposure analysis.¹

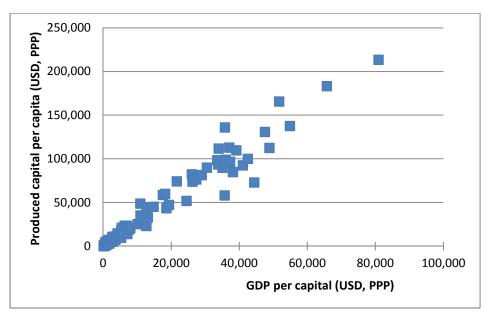


Figure S1. Relationship between GDP per capital and produced capital per capita (in USD, purchase power parity (PPP) exchange rate). Data from the World Bank.

1.3. Data on current defense levels in coastal cities

There is no global database of defense level in coastal cities, but patchy evidence is available on many of them with a bias towards richer countries and cities. It could be assumed that optimal defenses are present in all coastlines, designed using cost-benefit analyses as a decision framework.

However, it is observed that optimal defenses are rather exceptional and this assumption appears more useful as a baseline than a realistic description of existing protection. The recent landfall of Sandy illustrates for instance that Greater New York, despite having a larger GDP than London, Tokyo and Amsterdam, is currently only protected to a standard of roughly a 1 in 100 year flood with little formal flood defenses compared to those that exist for many European and Asian cities, and even New Orleans. The emphasis is on flood warning and evacuation as it is in most of the USA. Shanghai, a developing country city with a lower GDP than New York City and European cities, has a relatively high protection level similar to London. These examples highlight that protection levels are also strongly influenced by cultural, political and historical issues.

Here, we collected evidence on existing defenses starting from a previous analysis⁷, and we completed the defense database with estimates from the authors. Because of the uncertainty in some cases, we built two data sets, one with maximum protection level and minimum protection levels.

This defense database should not be considered as complete or exhaustive. Instead, it is a starting point created from limited information. We invite knowledgeable people to correct and improve the database using more detailed information as appropriate.

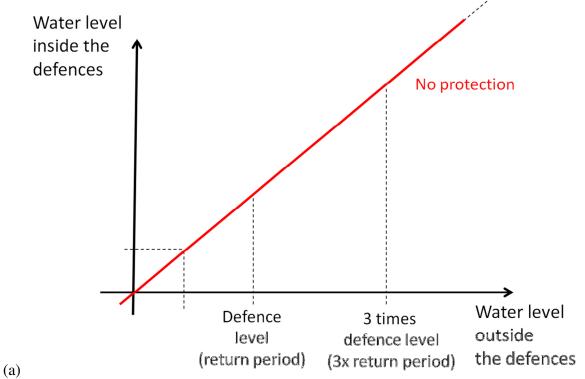
1.4. Water level in the city, accounting for defenses

The DIVA database provides information about the 10-yr, 100-yr and 1000-yr water levels on 12,148 segments around the world coasts. For all cities, these values were translated into water level probability distribution functions, assuming that these functions are in logarithm form.

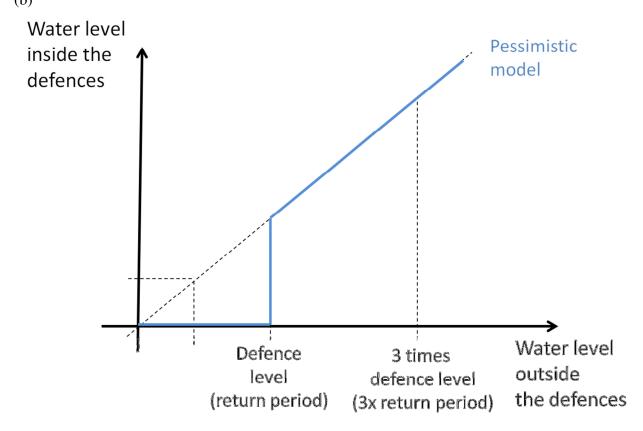
To assess flood losses, however, what matters is the water level within the defense system. To assess the probability distribution function of water levels within the defenses, assumptions are required on defense failure probabilities. In this analysis, we assume that defenses are designed to resist to a given standard of protection, expressed in terms of return period, and this standard of protection corresponds to a given defense water level.⁹

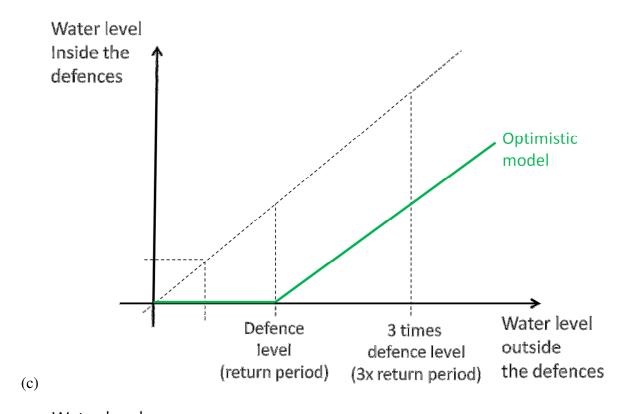
We assume that when the water level is below the designed defense level, failure probability is zero. Several simple assumptions can be made on how defenses behave when the defense level is exceeded, since this resistance depends on the protection type and characteristics (e.g., dikes vs. seawalls).

Here, we consider three distinct failure models, which in simple terms describe the range of possible behaviors (see also, Figure S2):



(b)





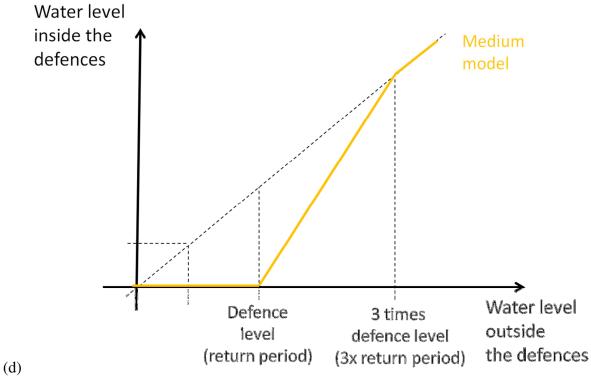


Figure S2. Representation of the relationship between the water level outside the city defenses and the water level inside the defenses, in the absence of protection (panel a), and with the three defenses failure models, i.e. the simplest and pessimistic model (panel b), the optimistic model (panel c), and the medium model (panel d).

The simplest assumption is that the defenses breaches when the design level is exceeded; in that case, there is no difference between the failed defenses and no defenses for events above the design standard. This model is also the most pessimistic.

The most optimistic assumption is that when defenses are first overtopped, this does not rapidly lead to breaching and they continue to provide some residual protection. Here, we assume that when defenses are exceeded by x cm, then the water level within the protection system is equal to x cm. In other terms, if a 2m protection experiences a 3m water level, the water level inside the protection will be reduced to 1m. This is a very simplistic assumption, but going beyond this assumption is beyond the scope of this study, and would require more detailed flood modeling more appropriate in an individual city assessment.

An intermediate assumption in which defenses progressively collapse as the water level increases until it reaches three times the design level (also expressed in return period), when total breaching is assumed. Hence, flooding increases linearly between the overtopping level and the collapse level. Above the collapse level, there is no difference between the failed defenses and no defenses for events above the design standard.

1.5. Flood losses, as a function of water level

Exposed assets as a function of water level were then translated into asset losses. To do this, assets in each elevation layer were first distributed into six categories: (1) lightweight-timber-framed dwellings; (2) masonry dwellings; (3) low-income-country dwellings; (4) dwelling contents; (5) non-residential structures; and (6) non-residential content.

Following Linham and colleagues⁷, we first distribute assets at risk in different broad categories, as shown in Table S1. The distribution of residential buildings then depends on the countries. For instance, North America and Australia are assumed to have mainly lightweight-timber-framed dwellings, while Europe and Asia have mainly masonry dwellings.

Table S1. Share of asset categories.

| Asset type | % total value of net fixed assets |
|---------------------------------------|-----------------------------------|
| Non-domestic buildings and structures | 42 |
| Residential buildings | 36 |
| Equipment | 14 |
| Domestic durables | 9 |

Then, assets are assumed to be homogenous distributed in each 50-cm elevation layer. And for each elevation (practically this is calculated using 0.5cm layers), we calculate the local water level (i.e. the water level measured against normal sea level minus the elevation, also measured against normal sea level) and the density of assets for each asset type.

Then depth-damage functions are used to calculate losses for each elevation and each asset type. Six depth-damage curves linking flood depth to the ratio of damage have been used for lightweight timber buildings, brick or concrete buildings, low-income country buildings, dwelling content, non-residential structures, and non-residential contents.⁷ The depth-damage curves for the six categories of assets are reproduced in Table S2.

Table S2. Depth-damage curves for the six asset types used in this study. The function provides the share of losses (total loss divided by the total value) as a function of flood depth (depth is zero in this table when flood depth is larger than zero, i.e. when there is a flood).

| Zero in tins tuete when | | Proportion of damage by depth and asset category (%) | | | | | | | | |
|--|-----|--|------|------|------|------|------|------|--|--|
| Depth (meters) | 0 | 0.25 | 0.50 | 1.00 | 1.50 | 2.00 | 2.50 | 3.00 | | |
| Dwellings: lightweight timber framed Source: Dale (2009) ¹⁰ | 1.6 | 16.9 | 44.5 | 61.8 | 62.9 | 63.3 | 69.4 | 69.4 | | |
| Dwellings: Masonry Source: Yan (2005) ¹¹ | 0.0 | 1.0 | 2.0 | 5.0 | 8.0 | 11.0 | 22.0 | 35.0 | | |
| Dwellings: structural, low income countries Source: Islam (1997) ¹² | 0.0 | 3.4 | 5.0 | 7.7 | 10.6 | 13.4 | 15.1 | 15.1 | | |
| Dwellings: contents Penning-Rowsell et al. (2003) ¹³ | 0.0 | 31.0 | 34.0 | 36.0 | 38.0 | 38.0 | 38.0 | 38.0 | | |
| Non-residential: structures Source: Penning-Rowsell et al. (2003) ¹³ | 4.0 | 18.3 | 39.1 | 50.4 | 59.7 | 64.6 | 66.5 | 68.1 | | |
| Non-residential: contents Source: Penning-Rowsell et al. (2003) ¹³ | 0.6 | 17.0 | 37.4 | 55.0 | 63.2 | 68.9 | 73.3 | 76.8 | | |

1.6. Calculation of mean annual flood losses.

Using the flood losses for each water level, and the probability of each water level, we can estimate the mean annual flood losses in each city, taking into account the estimated level of protection following the methodology from Hallegatte and colleagues³. Results differ largely for the different models of defense failure discussed above.

This analysis provides an estimate of aggregated average annual flood losses in the 136 coastal cities, and world average losses are shown in Table S3. Results are highly dependent on the defense overtopping model. The most optimistic overtopping model gives an aggregate annual flood loss worth \$46 million with the optimistic estimate of protection levels and \$90 million with the pessimistic estimate of protection level. These numbers are clearly overoptimistic: assuming that the return period of Katrina is 400 years, as in the Interagency Performance Evaluation Taskforce report¹⁴, average annual losses for New Orleans alone would be around

\$50 million, which is more than half of our assessment for all 136 port cities with the optimistic failure model.

Table S3. Aggregated global mean annual losses due to coastal floods in the 136 port cities in 2005, depending on the protection failure model and protection standard.

| | Aggregated global mean annual losses (million USD) | | | | |
|---------------------------------|--|---------|--|--|--|
| Protection failure model | Protection standard | | | | |
| 1 1000010011 11111010 1110 1101 | Minimum | Maximum | | | |
| Pessimistic | 8,823 | 5,744 | | | |
| Medium | 5,153 | 3,375 | | | |
| Optimistic | 90 | 46 | | | |

The losses are much larger with the pessimistic and the medium overtopping models, with aggregate losses ranging from \$3 billion to \$9 billion per year, depending on the protection failure model. Using these two models thus provide some bounds for aggregate losses. Investigating how sea level rise affects these losses with these different models provide an idea of the result robustness.

In the main text, all assessments are made using the simplest (pessimistic) model that assumes that defenses fail when their protection design is exceeded. We also use the maximum protection standard (corresponding to average flood losses of \$5.7 billion in the current situation).

Table S4 ranks the most vulnerable cities in 2005 using three different metrics of vulnerability. In the left column, the table provides a ranking that is comparable to the previous exposure analysis¹, based on exposure to the 100-yr flood, i.e. the assets below the 100-yr flood irrespective of defence standard. In the central column, the table shows a ranking in terms of absolute average annual losses (AAL in million USD), taking into account all potential floods and existing protection. Some of these estimates can be compared with more sophisticated approaches. For instance, the annual losses in New Orleans are estimated at \$600 million, close to the \$650 million estimates from the Interagency Performance Evaluation Taskforce. In the right column, cities are ranked according to relative vulnerability, namely the ratio of AAL to the city's GDP. This value can be understood as the share of the city's economic output that should be saved annually to pay for future flood losses.

| | | Ranking | by exposure | | | Ran | king by A | AL (million USI |)) | | Ranking b | y relative | AAL (% of city | GDP) |
|----|------------------------|--------------------|---|---------------------------------|----|------------------------|--------------------|---|---------------------------------------|----|------------------------|--------------------|-------------------------------------|---------------------------------|
| A | Urban gglomeration | 100-yr exposure | AAL, with protection (millions USD) | AAL, with protection (% of GDP) | A | Urban gglomeration | 100-yr exposure | AAL, with protection (millions USD) | AAL, with protection (% of GDP) | 1 | Urban Agglomeration | 100-yr exposure | ALL, with protection (millions USD) | AAL, with protection (% of GDP) |
| 1 | Miami | 366 421 | 672 | 0.30% | 1 | Guangzhou | 38,508 | 687 | 1.32% | 1 | Guangzhou | 38,508 | 687 | 1.32% |
| 2 | New York- Newark | 236 530 | 628 | 0.08% | 2 | Miami | 366,421 | 672 | 0.30% | 2 | New Orleans | 143,963 | 507 | 1.21% |
| 3 | Osaka-Kobe | 149 935 | 120 | 0.03% | 3 | New York- Newark | 236,530 | 628 | 0.08% | 3 | Guayaquil | 3,687 | 98 | 0.95% |
| 4 | New Orleans | 143 963 | 507 | 1.21% | 4 | New Orleans | 143,963 | 507 | 1.21% | 4 | Ho Chi Minh City | 18,708 | 104 | 0.74% |
| 5 | Tokyo | 122 910 | 27 | 0.00% | 5 | Mumbai | 23,188 | 284 | 0.47% | 5 | Abidjan | 1,786 | 38 | 0.72% |
| 6 | Amsterdam | 83 182 | 3 | 0.01% | 6 | Nagoya | 77,988 | 260 | 0.26% | 6 | Zhanjiang | 2,780 | 46 | 0.50% |
| 7 | Nagoya | 77 988 | 260 | 0.26% | 7 | Tampa-St Petersburg | 49,593 | 244 | 0.26% | 7 | Mumbai | 23,188 | 284 | 0.47% |
| 8 | Rotterdam | 76 585 | 2 | 0.01% | 8 | Boston | 55,445 | 237 | 0.13% | 8 | Khulna | 2,073 | 13 | 0.43% |
| 9 | Virginia Beach | 61 507 | 89 | 0.15% | 9 | Shenzen | 11,338 | 169 | 0.38% | 9 | Palembang | 1,161 | 27 | 0.39% |
| 10 | Boston | 55 445 | 237 | 0.13% | 10 | Osaka-Kobe | 149,935 | 120 | 0.03% | 10 | Shenzen | 11,338 | 169 | 0.38% |
| 11 | Tampa-St Petersburg | 49 593 | 244 | 0.26% | 11 | Vancouver | 33,456 | 107 | 0.14% | 11 | Hai Phòng | 6,348 | 19 | 0.37% |
| 12 | London | 45 130 | 5 | 0.00% | 12 | | 11,408 | 104 | 0.24% | 12 | N'ampo | 507 | 6 | 0.31% |
| 13 | Fukuoka- Kitakyushu | 39 096 | 82 | 0.09% | 13 | Ho Chi Minh City | 18,708 | 104 | 0.74% | 13 | Miami | 366,421 | 672 | 0.30% |
| 14 | Guangzhou | 38 508 | 687 | 1.32% | 14 | Kolkata | 14,769 | 99 | 0.21% | 14 | 1100111 | 855 | 14 | 0.29% |
| 15 | Shanghai | 34 306 | 4 | 0.00% | 15 | Guayaquil | 3,687 | 98 | 0.95% | 15 | Tampa-St Petersburg | 49,593 | 244 | 0.26% |
| 16 | Vancouver | 33 456 | 107 | 0.14% | 16 | I | 22,132 | 89 | 0.04% | 16 | Nagoya | 77,988 | 260 | 0.26% |
| 17 | Hong Kong | 26 988 | 8 | 0.00% | 17 | Virginia Beach | 61,507 | 89 | 0.15% | 17 | Surat | 3,288 | 30 | 0.25% |
| 18 | Hamburg | 26 260 | 6 | 0.01% | 18 | Fukuoka- Kitakyushu | 39,096 | 82 | 0.09% | 18 | Tianjin | 11,408 | 104 | 0.24% |
| 19 | St Peterburg | 23 384 | 3 | 0.00% | 19 | Baltimore | 14,042 | 76 | 0.08% | 19 | Grande_Vitória | 6,738 | 32 | 0.23% |
| 20 | Hiroshima | 23 331 | 37 | 0.06% | 20 | Jakarta | 4,256 | 73 | 0.14% | 20 | Xiamen | 4,486 | 33 | 0.22% |

Table S4. City ranking by estimated exposure and risk for 2005. All monetary values are in million USD.

2. Scenarios for the future

2.1. Future cities: population, income, assets

To develop future scenarios, our analysis combines three scenarios for socio-economic changes. The first scenario (labelled NC) assumes an unchanged (or baseline) population and wealth.

The second and third scenarios start from the OECD long-term scenarios for population and GDP, and use extrapolations of UN scenarios for urbanization rate to project future city population. The OECD socio-economic baseline scenario used here are updated from those used in the previous analysis¹ and an extension of those recently published the OECD as part of the OECD Environmental Outlook to 2050. 15,16

The scenario is constructed using the OECD ENV-Linkages model – a recursive dynamic neoclassical general equilibrium model (GE). It is a global economic model built primarily on a database of national economies. In its current form, the model represents the world economy in 15 countries/regions, each with 22 economic sectors. The scenario creates a consistent projection of economic activity for the coming decades to 2070, applying the general framework of "conditional convergence." The assumption of "conditional convergence" assumes that there will be some gradual convergence of income levels towards those of the most developed economies. The methodology used to derive per capita GDP trend pathways on a country basis relies on a conditional convergence hypothesis for the key drivers of per capita economic growth in the long run, i.e. for population, total factor productivity, physical capital, employment and human capital.

Urbanization scenarios are similar to the ones used in the previous analysis¹ and are based on an extrapolation to 2070 of UN urbanization scenarios.

The second scenario (labelled S) assumes that all cities in a given country grow at the same rate leading by the 2070s to several coastal cities with populations exceeding 50 million people¹; the third scenario (labelled L) assumes that no city can exceed 35 million inhabitants.

The assessment is based on the assumption that the future assets (infrastructure, housing, productive capital) that will be built in coastal cities will have the elevation distribution than the assets that are already installed. Under this assumption, future assets in one elevation layer increase linearly with total assets in the city. As a result, future exposures (and losses) are proportional to current exposures (and losses) and to the increase in capital in the cities.

2.2. Taking into account climate-induced sea level rise

Considering the uncertainty on future sea level¹⁷, we make simple assumptions with optimistic and pessimistic scenarios. We assume that climate-induced sea level rise is homogeneous globally and that climate change and sea level rise do not change storm surge likelihood. The analysis combines three scenarios on sea level rise. The first one (labelled "s") assumes a stable sea level over the 21st century; the second scenario (labelled "o") is optimistic and assumes that sea level rise will reach 10 cm in 2030, 20 cm in 2050, and 30 cm in 2070; the last – and most pessimistic – scenario (labelled "p") assumes that sea level rise reaches 20 cm in 2030, 40 cm in 2050, and 70 cm in 2070.

A significant difference with the previous analysis¹ is that storminess is assumed to remain unchanged. This change makes results more conservative, considering the uncertainty on change in storm intensity.¹⁷

2.3. Taking into account local subsidence

Small magnitudes of land uplift and subsidence are almost universal, contributing to local sealevel change 17. However, the magnitudes are generally small and not considered here. In addition, in susceptible locations such as deltaic areas, human-induced subsidence due to groundwater withdrawal and drainage can be significant, especially in cities built on deltas. Maximum subsidence during the 20th Century has been up to 5 metres 18-22 and subsidence is seen as a major threat comparable to climate change in many coastal Asian cites 21. The mean subsidence in these cities is less well measured and further, future subsidence is uncertain as it depends on human action. Hence reasonable high-end mean scenarios of human-induced subsidence are developed following Nicholls and colleagues 4, and applied to the entire flood prone area of the cities where such subsidence may occur.

Two scenarios on subsidence are considered. The first scenario assumes no subsidence (natural or artificial) and is labelled "0"; in the second scenario, labelled "1", natural and artificial subsidence affects susceptible cities that are mainly located wholly or partly on a delta, making local sea level rise by 20 cm more in 2030, 40 cm more in 2050, and 50 cm more in 2070. This would largely constitute human-induced subsidence. In cities potentially affected by human-induced subsidence, the local sea level can thus rise by up to 120 cm by the 2070s in the most pessimistic scenario when global sea-level rise and subsidence are combined.

2.4. Combining scenarios

Combining all these assumptions leads to 3x3x2x3x2=108 scenarios. Because it would be too complex to present all results, the main paper focuses on five future scenarios summarized in Table S5. Supplementary data provide results for all scenarios.

| | | | | Scenarios | | |
|-----------------------|---------------------------------------|------------------------|---|--|--|---|
| | | NC (present situation) | SEC (socio- economic change, no SLR, no subsidence) | SEC-S (socio- economic change, no SLR, with subsidence) | SLR-1 (socio- economic change, optimistic SLR, subsidence) | SLR-2 (socio- economic change, pessimistic SLR, subsidence) |
| | Constant (NC) | X | | | | |
| Socio-economic trend | Scenario with no city limit (S) | | | | | |
| | Scenario with city limit (L) | | X | X | X | X |
| | Stable (s) | X | X | X | | |
| Sea Level Rise | Optimistic rise (o) | | | | X | |
| | Pessimistic rise (p) | | | | | X |
| | No (0) | X | X | | | |
| Subsidence | Yes (1) | | | X | X | X |
| | Optimistic | | | | | |
| Defense failure model | Medium | | | | | |
| model | Pessimistic | X | X | X | X | X |
| Protection | Min | | | | | |
| levels | Max | X | X | X | X | X |

Table S5. Characteristics of the five scenarios analyzed in the main text.

3. Taking into account adaptation.

Changes in sea levels will trigger investments in new and/or reinforced coastal defenses. However, there is a large uncertainty on how adaptation will be implemented. Here, we tested three assumptions about adaptation, termed adaptation options:

- No upgrade: The most pessimistic, which is an absence of defense upgrade (Option NA).
- Maintain defense standards: The defenses will be improved to maintain coastal flooding likelihood. In practical terms, this is equivalent to assuming that dikes and seawalls will be raised by the same magnitude as relative sea level rise in each city, including subsidence as appropriate (Option PD).
- Maintain relative risk: The most optimistic assumption considered is that flood risk (i.e. the relative mean annual losses) will be maintained unchanged by raising protection by more than the relative sea level rise. In practical terms, this adaptation scenario assumes that the Standard of Protection rises appropriately to maintain risk levels and the probability of mean annual losses remains unchanged, allowing for the effect of socioeconomic change. (Option PL).

To maintain constant absolute levels of risk would require that the Standard of Protection rises even more than the PL scenario, to compensate for the increase in value at risk due to economic and population growth.

Table S6 presents aggregate losses for the 136 cities, for different scenarios and different adaptation scenarios. It shows the strong increase in the absence of adaptation, with total losses largely exceeding \$1 trillion per year. It also show the increase in risk when adaptation only maintain the probability of occurrence of a flood.

| | Mean annual losses (million USD) in 2050 | | | | | | |
|--|--|---|---|--|--|--|--|
| | | Adaptation Options | | | | | |
| | No adaptation (NA) | Maintain present defences & constant flood probability (PD) | Maintain present average losses relative to local wealth (PL) | | | | |
| Scenario NC (no change – the current situation) | 5,744 | 5,744 | 5,744 | | | | |
| Scenario SEC (only socio-economic changes) | 52,015 | 52,015 | 52,015 | | | | |
| Scenario SEC-S (adds subsidence) | 687,186 | 58,579 | 52,015 | | | | |
| Scenario SLR-1 (adds optimistic sea-level rise) | 1,192,785 | 59,767 | 52,015 | | | | |
| Scenario SLR-2 (same as Scenario 1 with pessimistic sea- level rise) | 1,566,856 | 63,273 | 52,015 | | | | |

Table S6. Change in aggregated annual losses in 2050 in the 136 cities, due to different driver Scenarios and possible adaptation options.

| City | mainta prob | aptation ins flood pability ion PD) | Adaptation needs to maintain mean annual losses (Option PL) | | | | Flood losses in event of defence failure (million USD – total losses for single event) (Option PL) | | | |
|-----------------------------------|----------------|--|---|--------------------------------------|--|--|---|-------------------|--------------|--|
| | AAL (M\$) | Increase (%) (Fig. 1) | Local sea- level rise (cm) | Rise in dike height (cm) | Protection standard in 2005 (Return period (yrs)) | Required Protection standard in 2050 (Return period (yrs)) | Scenario SEC | Scenario SLR-1 | Increase (%) | |
| Alexandria | 504 | 154% | 60 | 67 | 100 | 268 | 16,533 | 50,551 | 206% | |
| Barranquilla | 10 | 116% | 60 | 66 | 10 | 23 | 39 | 100 | 156% | |
| Napoli | 2 | 82% | 20 | 24 | 50 | 97 | 39 | 80 | 105% | |
| Sapporo | 4 | 76% | 60 | 64 | 100 | 179 | 218 | 417 | 91% | |
| Santo Domingo | 34 | 65% | 20 | 23 | 10 | 17 | 179 | 325 | 82% | |
| Bayrut | 2 | 63% | 20 | 23 | 100 | 172 | 91 | 162 | 78% | |
| Houston | 190 | 60% | 60 | 68 | 50 | 84 | 5,110 | 9,053 | 77% | |
| Istanbul | 21 | 57% | 20 | 23 | 100 | 167 | 1,120 | 1,972 | 76% | |
| Jakarta | 1,750 | 54% | 60 | 63 | 10 | 16 | 10,513 | 17,276 | 64% | |
| Izmir | 11 | 51% | 20 | 22 | 100 | 157 | 625 | 1,023 | 64% | |
| Marseille- Aix-en- Provence | 5 | 51% | 20 | 24 | 100 | 160 | 266 | 453 | 70% | |
| Athens | 1 | 50% | 20 | 23 | 100 | 156 | 48 | 78 | 63% | |
| Shanghai | 93 | 48% | 60 | 66 | 1000 | 1,519 | 57,646 | 89,545 | 55% | |
| Banghazi | 22 | 46% | 20 | 23 | 100 | 154 | 1,223 | 2,019 | 65% | |
| Tel Aviv- Yafo | 0 | 45% | 20 | 23 | 100 | 152 | 7 | 11 | 57% | |
| Fuzhou_Fujia n | 199 | 45% | 60 | 62 | 20 | 29 | 2,697 | 3,873 | 44% | |
| Ningbo | 256 | 45% | 60 | 62 | 20 | 29 | 3,397 | 5,001 | 47% | |
| La Habana | 0 | 42% | 20 | 22 | 100 | 146 | 14 | 21 | 50% | |
| Port-au- Prince | 1 | 41% | 20 | 22 | 10 | 14 | 6 | 10 | 67% | |
| Algiers | 9 | 41% | 20 | 22 | 50 | 72 | 291 | 439 | 51% | |

Table S7. The twenty cities with the largest increase in average annual losses (from 2005 to 2050) (scenario SLR-1, adaptation option PD), if adaptation only maintains present defence standards. It provides the increase in defence height needed to maintain flood risk, and corresponding increase in protection standard. The three last columns describe the consequence of a storm that exceeds protection standards, in the case without and with sea-level rise and subsidence.

Results for all scenarios, all defense level estimates, the three overtopping model, and all time horizons (2030-2050-2070) are provided in Supplementary Data, together with the MatLab files used to produce them.

These data are organized as follows.

The directory "CODES" includes all MatLab codes. They reproduce all results. Running "City_Loop.m" produces all the information by running the risk assessment for all cities and storing results as MatLab variables.

Then, running "write_results_subs.m" writes all results in Excel format.

The main input file is located in the directory "INPUT CITY SCENARIOS": The file "City_Scenarios_Input.xls" provides the input data, namely population and economic data and scenarios, water extreme level data, protection level data, and a marker to indicate which city is subject to subsidence. This file also includes all socio-economic and urbanization scenarios.

GIS Inputs are located in the directory "GIS DATA". There is one file per city, and it provides the population in each 50cm elevation layer.

The results files are the following.

In the directory "RESULTS MAIN PAPER", there are:

- The file "All_scenarios_aggregated.xls" provides the global aggregated results for all scenarios, and the five ones used in the main text are highlighted in yellow.
- The five scenarios that are described in details in the main text of the letter: "SEC Scenario.xls", "SEC-S Scenario.xls", "SLR-1 Scenario.xls" and "SLR-2 Scenario.xls" provides the results for these five scenarios, for the 136 cities, the two assumptions on the current protection level, and the three adaptation options. Each file has one tab for each defense failure model and for each time horizon (2005, 2030, 2050, and 2070). There is no file for the NC scenario, because it corresponds to the 2005 situation in all other scenarios.

All other scenarios (a total of 108 scenarios) are in the folder "OTHER SCENARIOS", and they are named according to the scenario terminology (see Table S5):

First letter:

N for no socio-economic change,

S for socio-economic change without city limit and,

L for socio-economic change with city limit;

Second letter:

s for stable sea level

o for optimistic sea level rise

p for pessimistic sea level rise

Third letter:

0 for no subsidence

1 for the scenario with subsidence

Each file has then different tabs for different time horizons (2005, 2030, 2050 and 2070) and different defense failure models. And in each tab, one can find the three adaptation options.

Supplementary references

- 1. Hanson, S., Nicholls, R., Ranger, N., Hallegatte, S., Corfee-Morlot, J., Herweijer, C., Chateau, J., 2011. A global ranking of port cities with high exposure to climate extremes. Climatic change 104, 89–111.
- 2. McGranahan, G., Balk, D., Anderson, B., 2007. The rising tide: assessing the risks of climate change and human settlements in low elevation coastal zones. Environment and Urbanization 19, 17–37.
- 3. Hallegatte, S., Ranger, N., Mestre, O., Dumas, P., Corfee-Morlot, J., Herweijer, C., Muir Wood, R., 2011. Assessing climate change impacts, sea level rise and storm surge risk in port cities: a case study on Copenhagen. Climatic change 104, 113–137.
- 4. Nicholls, R.J., Hanson, S., Herweijer, C., Patmore, N., Hallegatte, S., Corfee-Morlot, J., Château, J., Muir-Wood, R., 2008. "Ranking Port Cities with High Exposure and Vulnerability to Climate Extremes: Exposure Estimates", OECD Environment Working Papers, No. 1, OECD Publishing, Paris. doi: 10.1787/011766488208
- 5. Rodriguez, E., C.S. Morris, J.E. Belz, E.C. Chapin, J.M. Martin, W. Daffer, S. Hensley, 2005, An assessment of the SRTM topographic products, Technical Report JPL D-31639, Jet Propulsion Laboratory, Pasadena, California, 143 pp.
- 6. World Bank, 2010. The Changing Wealth of Nations: Measuring Sustainable Development in the New Millennium, The World Bank, Washington DC, USA.
- 7. Linham, M., Green, C., Nicholls, R. J. (2010) Costs of adaptation to the effects of climate change in the world's large port cities. AVOID Avoiding dangerous climate change Report WS2D1R14 London, UK; Department of Energy and Climate Change (DECC) and Department for Environment Food and Rural Affairs (DEFRA). http://www.avoid.uk.net/ http://www.metoffice.gov.uk/avoid/files/resources-researchers/AVOID_WS2_D1_14_20100701.pdf.
- 8. Vafeidis, A.T., Nicholls, R.J., McFadden, L., Tol, R.S.J., Hinkel, J., Spencer, T., Grashoff, P.S., Boot, G., Klein, R.J.T., 2008. A new global coastal database for impact and vulnerability analysis to sea-level rise. Journal of Coastal Research 917–924.
- 9. Hall, J.W., Dawson, R.J., Sayers, P.B., Rosu, C., Chatterton, J.B. and Deakin, R., 2003. A methodology for national-scale flood risk assessment. Proceedings of ICE. Water and Maritime Engineering, 156, 235-247.
- 10. Dale, K.W., Edwards, M.R., Middelmann, M.H. and Zoppou, C., 2009. Structural flood vulnerability and the Australianisation of Black's Curves. Australia: Geoscience Australia.
- 11. Yan, H., 2005. Appropriate Modeling for Integrated Flood Risk Assessment. PhD Thesis, University of Twente.

- 12. Islam, K.M.N., 1997. The Impacts of Flooding and Methods of Assessment in Urban Areas of Bangladesh. PhD Thesis, Middlesex University
- 13. Penning-Rowsell, E.C., Johnson, C., Tunstall, S.M., Tapsell, S.M., Morris, J., Chatterton, J.B., Coker, A. and Green, C., 2003. The benefits of flood and coastal defence: techniques and data for 2003. Flood Hazard Research Centre, Middlesex University, London.
- 14. Interagency Performance Evaluation Taskforce, US Army Corps of Engineers, www.nolarisk.usace.army.mil.
- 15. OECD, 2012. OECD Environmental Outlook to 2050: The Consequences of Inaction. OECD Publishing: Paris.
- 16. Chateau, J., C. Rebolledo and R. Dellink (2011), "An Economic Projection to 2050: The OECD "ENV-Linkages" Model Baseline", OECD Environment Working Papers, No. 41, OECD Publishing. http://dx.doi.org/10.1787/5kg0ndkjvfhf-en.
- 17. IPCC, 2012: Managing the Risks of Extreme Events and Disasters to Advance Climate Change Adaptation. A Special Report of Working Groups I and II of the Intergovernmental Panel on Climate Change [Field, C.B., V. Barros, T.F. Stocker, D. Qin, D.J. Dokken, K.L. Ebi, M.D. Mastrandrea, K.J. Mach, G.-K. Plattner, S.K. Allen, M. Tignor, and P.M. Midgley (eds.)]. Cambridge University Press, Cambridge, UK, and New York, NY, USA, 582 pp.
- Emery, K. O. & Aubrey, D. G. 1991. Sea Levels, Land Levels, and Tide Gauges. xiv +
 237pp. Berlin, Heidelberg, New York, Paris, London, Tokyo, Hong Kong: Springer-Verlag. ISBN 3 540 97449 0.
- 19. Nicholls, R.J., 1995. Coastal megacities and climate change. GeoJournal 37, 369–379.
- 20. Nicholls, Robert J. (2010) Impacts of and responses to sea-level rise. In, Church, John A., Woodworth, Philip L., Aarup, Thorkild and Wilson, W. Stanley (eds.) Understanding Sea-Level Rise and Variability. Chichester, GB, Wiley-Blackwell, 17-51.
- 21. Nicholls, R.J., 2011. Planning for the impacts of sea level rise. Oceanography [Special issue: Sea Level], 24, (2), 142-155.
- 22. World Bank, 2010. Climate Risks and Adaptation in Asian Coastal Megacities: A Synthesis Report. World Bank, Washington DC, 120 pp.

| The protection level is taken at its optimistic bound (maximi | ani protection). |
|---|------------------------|
| The protection failure model is the simplest one (model #1 | nanel a in Figure \$2\ |

| | | No ch | ange in sea lev | vel | 20cm sea le subsidence (r | | (adaptation | evel rise and at constant ation scenar | probability, | 20cm sea level rise and subsidence (adaptation at constant relative risk, adaptation scenario PL) | | | |
|--|-------------------------|---|--|---------------------------|----------------------------------|------------------------|---|--|-------------------------------------|---|---|---|--------------------------------------|
| Urban Agglomeration | 100-yr exposition (M\$) | 100-yr losses with no protection (M\$) | Protection level in 2005 (return period in years) | Mean annual loss (M\$) | Mean annual loss (% city GDP) | Mean annual loss (M\$) | Mean increase due to SLR and subsidence | Mean annual loss (M\$) | Mean annual loss (% city GDP) | Mean increase due to SLR and subsidence | Increase in dike height (for 20cm SLR + subsidence) | New protection level to maintain relative risk (return period in years) | New loss if protection is overtopped |
| Abidjan | 4,017 | 1,328 | 10 | 826 | 0.72% | 8,968 | 986% | 1,023 | 0.89% | 24% | 21.3 | 13 | 9,839 |
| Accra Adelaide | 190 1,278 | 57 643 | 10 100 | 36 4 | 0.03% 0.00% | 393 19 | 993% 368% | 43 | 0.04% | 20% 7% | 20.9 | | 418 417 |
| Al Kuwayt (Kuwait City) | 1,579 | 549 | | 9 | 0.00% | 235 | 2627% | 9 | | 5% | 20.93 | | 896 |
| Al-Iskandariyah (Alexandria) | 27,675 | 3,669 | 100 | 199 | 0.09% | 34,621 | 17300% | 504 | 0.22% | 154% | 67 | | 50,551 |
| Amsterdam | 187,159 | 103,598 | 10000 | 9 | 0.01% | 54,021 | | 10 | | 9% | 63.06 | | 93,122 |
| Athinai (Athens) | 487 | 113 | 100 | 1 | 0.00% | 18 | | 1 | | 50% | 23 | | 78 |
| Auckland | 1,119 | 563 | | 4 | 0.00% | 42 | | 4 | | 6% | 20.56 | | 435 |
| Baixada Santista (Santos) Baltimore | 788 31,595 | 218 15,406 | 50 50 | 9 238 | 0.01% 0.08% | 274 1,178 | 3041% 396% | 9 271 | 0.01% | 8% 14% | 20.45 21.76 | | 463 12,741 |
| Banghazi | 3,048 | 734 | | 15 | 0.01% | 1,178 | 1112% | 271 | | 46% | 21.70 | | 2,019 |
| Barcelona | 2,370 | 549 | 100 | 3 | 0.00% | 37 | 1195% | 4 | 0.00% | 40% | 22.93 | 146 | 372 |
| Barranquilla Bayrut_(Beirut) | 94 | 28 45 | | 5 | 0.00% | 87 71 | 1782% 6658% | 10 | | 116% 63% | 66 23 | | 100 162 |
| Belém | 2,007 | 542 | 50 | 12 | 0.00% | 93 | 698% | 13 | | 8% | 20.8 | | 609 |
| Boston | 124,752 | 69,675 | | 741 | 0.13% | 5,557 | 650% | 793 | 0.14% | 7% | 20.7 | 54 | 38,400 |
| Brisbane Buenos Aires | 12,803 6,847 | 5,486 1,868 | 100 50 | 29 44 | 0.02% 0.00% | 178 161 | 504% 268% | 33 49 | | 11% 13% | 21.21 22.07 | | 3,102 2,258 |
| Cape Town | 757 | 209 | 100 | 44 | 0.00% | 51 | | 5 | | 7% | 20.59 | | 471 |
| Chennai (Madras) | 3,457 | 966 | 10 | 825 | 0.12% | 7,392 | 796% | 939 | 0.14% | 14% | 21.24 | 11 | 9,001 |
| Chittagong | 4,751 | 1,186 | | 154 | 0.08% | 671 | 337% | 162 | 0.09% | 6% | 20.76 | | 3,138 |
| Ciudad de Panama (Panama City) | 468 | 200 | 10 | 42 | 0.06% | 431 | 916% | 44 | 0.06% | 4% | 20.23 | | 437 |
| Conakry Dakar | 449 177 | 166 48 | | 66 35 | 0.09% | 716 377 | 981% 973% | 78 41 | | 18% 16% | 20.87 20.78 | | 760 398 |
| Dalian | 1,973 | 750 | 20 | 143 | 0.04% | 1,223 | 757% | 155 | 0.05% | 9% | 20.8 | | 3,008 |
| Dar-el-Beida (Casablanca) | 940 | 301 | 100 | 12 | 0.01% | 191 | 1458% | 13 | | 8% | 20.56 | 108 | 1,288 |
| Dar-es-Salaam | 101 | 41 | | 62 5 | 0.07% | 624 | 910% | 65 | | 6% | 20.43 | | 642 |
| Davao Dhaka | 61 10,126 | 22 2,774 | 10 20 | 451 | 0.01% 0.10% | 56 6,287 | 921% 1294% | 506 | 0.01% 0.11% | 7% 12% | 20.4 62.72 | | 57 9,734 |
| Douala | 162 | 48 | | 24 | 0.03% | 284 | 1070% | 31 | 0.04% | 29% | 62.62 | 13 | 304 |
| Dubayy (Dubai) | 49,574 | 16,861 | 100 | 286 | 0.11% | 4,180 | 1362% | 305 | 0.12% | 7% | 20.49 | | 29,881 |
| Dublin Durban | 6,825 1,094 | 1,866 317 | 70 500 | 7 | 0.01% | 85 38 | 1046% 2444% | 8 | 0.01% | 11% 7% | 20.87 20.38 | | 554 774 |
| EI_Djazaïr (Algiers) | 363 | 88 | | 7 | 0.00% | 253 | 3755% | 9 | 0.00% | 41% | 22 | 72 | 439 |
| Fortaleza Fukuoka-Kitakyushu | 493 | 160 | 50 100 | 2 | 0.00% | 52 | 2762% | 2 | | 23% | 21.33 | | 105 |
| Fuzhou_Fujian | 87,966 2,651 | 42,913 919 | 20 | 177 137 | 0.09% 0.05% | 19,904 3,525 | 11134% 2471% | 221 199 | 0.12% 0.08% | 25% 45% | 61.47 62 | | 21,789 3,873 |
| Glasgow | 4,365 | 1,757 | | 4 | 0.00% | 95 | 2243% | 4 | | 6% | 20.36 | | 842 |
| Grande_Vitória Guangzhou_Guangdong | 15,160 86,644 | 4,712 37,105 | 50 20 | 190 11,928 | 0.23% 1.32% | 2,643 254,721 | 1289% 2036% | 211 13,200 | 0.26% 1.46% | 11% 11% | 20.82 60.88 | | 10,266 261,495 |
| Guayaquil | 8,296 | 3,133 | | 2,813 | 0.95% | 31,288 | 1012% | 3,189 | 1.08% | 13% | 60.64 | | 31,645 |
| Hai Phòng | 14,283 | 4,728 | | 320 | 0.37% | 6,209 | 941% | 383 | 0.44% | 30% | 64 | | 17,867 |
| Hamburg Hangzhou | 59,084 586 | 28,626 149 | | 14 6 | 0.01% | 61 125 | | 15 7 | | 6% 22% | 20.79 62.17 | | 14,225 134 |
| Helsinki | 2,482 | 936 | 200 | 3 | 0.00% | 11 | 245% | 3 | 0.00% | 8% | 21.35 | 217 | 645 |
| Hiroshima | 52,494 | 23,307 | | 80 | 0.06% | 9,456 | 11660% | 109 | 0.08% | 35% | 62.07 | | 10,696 |
| Hong Kong Houston | 60,722 29,147 | 31,300 12,337 | 900 50 | 140 119 | 0.00% 0.02% | 1,269 6,088 | 808% 5017% | 150 190 | 0.00% | 7% 60% | 20.63 | | 131,051 9,053 |
| Inchon | 22,825 | 9,493 | 100 | 29 | 0.01% | 191 | 553% | 30 | 0.02% | 2% | 20.24 | 102 | 2,953 |
| Istanbul Izmir | 3,162 1,415 | 419 341 | 100 100 | 13 | 0.00% | 327 314 | 2348% 4299% | 21 11 | | 57% 51% | 23 22 | | 1,972 1,023 |
| Jakarta | 9,577 | 2,553 | 100 | 1,139 | 0.14% | 16,354 | 1336% | 1,750 | 0.22% | 54% | 63 | 16 | 17,276 |
| Jiddah | 1,092 | 393 | 10 | 60 | 0.02% | 631 | 943% | 67 | 0.02% | 10% | 20.47 | | 651 |
| Karachi Khulna | 792 4,665 | 260 1,748 | 10 20 | 121 295 | 0.02% | 1,385 6,172 | 1040% 1989% | 145 409 | 0.02% | 20% 39% | 61.64 66.11 | 12 | 1,432 7,650 |
| København (Copenhagen) | 9,567 | 2,462 | 000 | 7 | 0.01% | 36 | 391% | 8 | 0.01% | 15% | 21.79 | | 1,573 |
| Kochi (Cochin) | 1,923 | 505 | 10 | 430 | 0.29% | 4,327 | 907% | 481 | 0.33% | 12% | 20.97 | 11 | 4,648 |
| Kolkata (Calcutta) | 33,231 | 12,649 | | 2,704 | 0.21% | 56,303 | | 3,350 | 0.26% | | 63.76 | | 64,760 |
| Krung_Thep (Bangkok) Kuala Lumpur | 49,065 19,437 | 16,462 6,696 | | 596 56 | 0.07% 0.03% | 20,778 253 | 3387% 355% | 734 63 | 0.09% | 23% 13% | 63.86 21.69 | | 34,145 5,897 |
| La Habana (Havana) | 31 | 8 | 100 | 0 | 0.00% | 9 | 5939% | 0 | 0.00% | 42% | 22 | 146 | 21 |
| Lagos Lima | 2,750 47 | 728 14 | | 287 | 0.15% 0.00% | 3,026 39 | 954% 1009% | 328 4 | | 14% 26% | 20.79 21.29 | | 3,196 43 |
| Lisboa (Lisbon) | 4,764 | 1,770 | | 13 | 0.00% | 100 | | 15 | 0.01% | 10% | 20.95 | | 1,406 |
| Lomé | 1,005 | 308 | 100 | 28 | 0.05% | 1,017 | 3521% | 33 | 0.06% | 16% | 20.86 | 117 | 3,169 |
| London | 101,543 | 36,751 | 1000 | 13 | 0.00% | 65 | 384% | 14 | | 9% | 21.09 | | 13,789 |
| Los Angeles-Long Beach Santa Ana Luanda | 35,833 21 | 15,517 7 | | 188 9 | 0.08% | 9,427 95 | 4912% 952% | 203 10 | 0.09% | 8% 12% | 20.37 | | 9,960 99 |
| Maceió | 518 | 188 | 50 | 6 | 0.01% | 54 | 887% | 6 | 0.01% | 10% | 20.83 | 55 | 290 |
| Manila Maputo | 2,339 | 872 | | 254 | 0.05% | 2,846 | 1019% | 329 | | 29% | 63.62 | | 3,122 |
| Maputo Maracaibo | 404 775 | 169 236 | | 24 6 | 0.03% 0.01% | 184 67 | | 26 7 | | 7% 22% | 20.63 21.71 | | 2,523 658 |
| Marseille-Aix-en-Provence | 3,403 | 788 | 100 | 3 | 0.00% | 43 | 1233% | 5 | 0.00% | 51% | 24 | | 453 |
| Melbourne | 682 | 328 | 100 | 2 | 0.00% | 10 | | 2 | | 9% | 21.23 | 109 | 229 |
| Miami | 824,448 | 387,360 | 100 | 2,099 | 0.30% | 7,340 | 250% | 2,549 | 0.36% | 23% | 23.57 | | 228,589 |
| Montevideo Montréal | 992 4,235 | 334 2,482 | 50 50 | 14 14 | 0.02% | 50 86 | | 15 15 | | 9% 4% | | | 716 729 |
| Mumbai (Bombay) | 52,173 | 19,827 | 20 | 6,109 | 0.47% | 107,285 | 1656% | 6,414 | 0.49% | 5% | 20.34 | 21 | 126,406 |
| Muqdisho (Mogadishu) Nagoya | 21 175,473 | 99,196 | | 13 564 | 0.04% | 128 57,954 | 920% 10183% | 13 644 | | 6% 14% | 20.37 61.44 | | 131 63,344 |
| N'ampo | 1/5,4/3 | 99,196 | | 24 | 0.26% | 57,954 87 | | | | | 20.47 | | 246 |
| Napoli (Naples) | 981 | 125 | 50 | 1 | 0.00% | 35 | 2895% | 2 | 0.00% | 82% | 24 | 97 | 80 |
| Natal | 647 323,917 | 250 180,646 | | 9 1,583 | 0.02% 1.21% | 150 161,141 | 1505% 10080% | 10 1,864 | 0.02% 1.42% | 7% 18% | 20.49 61.99 | | 489 182,592 |
| New Orleans | 343,317 | | | | 0.08% | 7,914 | 304% | 2,056 | 0.08% | 5% | 20.7 | | 198,885 |
| New Orleans New York-Newark | 532,194 | 291,729 | 100 | 1,960 | 0.0676 | 7,514 | 30470 | | | | | | |
| New York-Newark Ningbo | 6,272 | 1,373 | 20 | 176 | 0.09% | 4,548 | 2479% | 256 | 0.13% | 45% | 62 | 29 | 5,001 |
| New York-Newark | | | 20 100 | | | | 2479% | | | 14% | | 29 115 | 5,001 2,146 98,671 |

| Perth | 4,220 | 2,145 | 100 | 16 | 0.01% | 65 | | 17 | 0.01% | 8% | 21.18 | 109 | 1,626 |
|--|---------|-----------|-------|-------|-------|--------|---------|-------|-------|-----|-------|--------|--------|
| Philadelphia | 49,797 | 27,432 | 50 | 279 | 0.04% | 1,017 | 265% | 294 | 0.04% | 5% | 20.83 | 53 | 14,127 |
| Port-au-Prince | 9 | 3 | 10 | 1 | 0.00% | 8 | | 1 | 0.00% | 41% | 22 | 14 | 10 |
| Portland | 3,753 | 1,716 | 50 | 3 | 0.00% | 114 | | 3 | 0.00% | 12% | 20.62 | 56 | 152 |
| Porto | 1,756 | 722 | 100 | 6 | 0.01% | 34 | | 7 | 0.01% | 7% | 20.86 | 107 | 669 |
| Porto Alegre | 0 | 0 | 50 | 10 | 0.00% | 71 | | 10 | 0.01% | 8% | 20.8 | 54 | 497 |
| Providence | 17,857 | 9,479 | 50 | 118 | 0.07% | 525 | | 127 | 0.08% | 7% | 20.96 | 54 | 6,084 |
| Pusan | 16,852 | 4,523 | 100 | 30 | 0.01% | 384 | 1165% | 34 | 0.01% | 10% | 20.8 | 111 | 3,262 |
| Qingdao | 3,055 | 958 | 20 | 163 | 0.05% | 3,260 | 1904% | 177 | 0.06% | 9% | 20.56 | 22 | 3,461 |
| Rabat | 248 | 71 | 100 | 2 | 0.00% | 60 | 2930% | 2 | 0.00% | 15% | 20.84 | 115 | 220 |
| Rangoon | 4,177 | 1,387 | 10 | 163 | 0.17% | 1,818 | 1017% | 202 | 0.21% | 24% | 63.34 | 12 | 1,962 |
| Recife | 1,218 | 477 | 50 | 19 | 0.01% | 259 | 1279% | 20 | 0.01% | 7% | 20.49 | 53 | 977 |
| Rio_de_Janeiro | 4,132 | 1,010 | 50 | 35 | 0.01% | 411 | | 38 | 0.01% | 9% | 20.72 | 55 | 1,839 |
| Rotterdam | 172,316 | 96,071 | 10000 | 7 | 0.01% | 256 | 3416% | 8 | 0.01% | 11% | 61.84 | 11,153 | 79,289 |
| Salvador | 377 | 143 | 50 | 5 | 0.00% | 245 | 4903% | 5 | 0.00% | 7% | 20.34 | 53 | 257 |
| San Diego | 1,369 | 731 | 50 | 12 | 0.00% | 410 | 3299% | 13 | 0.00% | 7% | 20.39 | 53 | 634 |
| San Francisco - Oakland | 34,156 | 13,240 | 50 | 149 | 0.03% | 1,703 | 1045% | 168 | 0.04% | 13% | 21.02 | 57 | 8,071 |
| San Jose | 2,826 | 932 | 50 | 2 | 0.00% | 10 | 551% | 2 | 0.00% | 11% | 21.13 | 56 | 83 |
| San Juan | 7,342 | 2,235 | 50 | 68 | 0.02% | 1,680 | 2365% | 89 | 0.03% | 31% | 21.78 | 66 | 4,239 |
| Sankt Peterburg (St. Petersburg) | 52,615 | 11,899 | 1000 | 11 | 0.00% | 187 | 1572% | 12 | 0.00% | 8% | 20.53 | 1,077 | 11,748 |
| Santo Domingo | 636 | 153 | 10 | 21 | 0.02% | 263 | 1166% | 34 | 0.03% | 65% | 23 | 17 | 325 |
| Sapporo | 1,339 | 591 | 100 | 2 | 0.00% | 346 | 14308% | 4 | 0.00% | 76% | 64 | 179 | 417 |
| Seattle | 10,235 | 6,100 | 50 | 85 | 0.02% | 545 | 542% | 87 | 0.02% | 3% | 20.33 | 52 | 4,304 |
| Shanghai | 77,189 | 28,576 | 1000 | 63 | 0.00% | 5,563 | 8721% | 93 | 0.01% | 48% | 66 | 1,519 | 89,545 |
| Shenzen | 25,510 | 10,362 | 20 | 2,929 | 0.38% | 17,553 | 499% | 3,136 | 0.40% | 7% | 20.79 | 21 | 60,550 |
| Singapore | 3,412 | 2,020 | 2000 | 2 | 0.00% | 27 | 1222% | 2 | 0.00% | 3% | 20.26 | 2,068 | 4,208 |
| Stockholm | 517 | 152 | 100 | 1 | 0.00% | 14 | 1613% | 1 | 0.00% | 26% | 21.73 | 128 | 98 |
| Surabaya | 727 | 163 | 10 | 80 | 0.04% | 1,052 | 1222% | 110 | 0.06% | 39% | 62.05 | 14 | 1,089 |
| Surat | 7,399 | 3,078 | 10 | 905 | 0.25% | 9,070 | 902% | 928 | 0.26% | 3% | 20.21 | 10 | 9,201 |
| Sydney | 8,293 | 4,314 | 100 | 28 | 0.01% | 172 | 509% | 31 | 0.01% | 9% | 21.03 | 110 | 2,943 |
| Taipei | 2,713 | 1,042 | 20 | 274 | 0.10% | 6,436 | 2250% | 344 | 0.12% | 26% | 61.25 | 25 | 6,757 |
| Tampa-St Petersburg | 111,585 | 56,155 | 50 | 763 | 0.26% | 2,997 | 293% | 859 | 0.29% | 13% | 21.87 | 57 | 40,022 |
| Tarabulus (Tripoli) | 267 | 64 | 100 | 2 | 0.00% | 19 | 1058% | 2 | 0.00% | 37% | 22.83 | 141 | 207 |
| Tel Aviv-Yafo (Tel Aviv-Jaffa) | 26 | 6 | 100 | 0 | 0.00% | 1 | 1255% | 0 | 0.00% | 45% | 23 | 152 | 11 |
| Thành-Pho-Ho-Chí-Minh (Ho Chi Minh City) | 42,093 | 20,216 | 50 | 1,743 | 0.74% | 7,335 | | 1,953 | 0.83% | 11% | 65.16 | 57 | 90,365 |
| Tianjin | 25,667 | 9,779 | 20 | 1,810 | 0.24% | 40,492 | | 2,276 | 0.30% | 26% | 62.85 | 25 | 44,463 |
| Tokyo | 276,547 | 140,711 | 1000 | 58 | 0.00% | 61,737 | 106550% | 77 | 0.00% | 33% | 62 | 1,346 | 75,503 |
| Ujung Pandang | 156 | 33 | 10 | 11 | 0.01% | 67 | | 12 | 0.02% | 12% | 21.29 | 11 | 116 |
| Ulsan | 1,209 | 368 | 100 | 2 | 0.00% | 53 | 2189% | 3 | 0.00% | 30% | 21.81 | 133 | 286 |
| Vancouver | 75,276 | 43,232 | 50 | 325 | 0.14% | 18,912 | | 423 | 0.18% | 30% | 61.64 | 65 | 20,674 |
| Virginia Beach | 138,390 | 62,671 | 140 | 278 | 0.15% | 1,520 | 446% | 303 | 0.16% | 10% | 21.05 | 153 | 40,549 |
| Visakhapatnam | 369 | 111 | 10 | 110 | 0.08% | 1,129 | 926% | 122 | 0.08% | 11% | 20.73 | 11 | 1,189 |
| Washington, D C | 12,326 | 5,972 | 50 | 74 | 0.01% | 1,045 | 1321% | 82 | 0.01% | 12% | 20.89 | 56 | 3,976 |
| Wenzhou | 3,187 | 921 | 20 | 152 | 0.06% | 3,367 | 2119% | 174 | 0.07% | 15% | 60.85 | 23 | 3,447 |
| Xiamen | 10,093 | 3,682 | 20 | 572 | 0.22% | 12,182 | 2029% | 729 | 0.29% | 27% | 63.67 | 26 | 13,902 |
| Yantai | 679 | 164 | 20 | 21 | 0.01% | 406 | 1878% | 24 | 0.01% | 18% | 21.16 | 24 | 461 |
| Zhanjiang | 6,256 | 2,865 | 20 | 806 | 0.50% | 16,709 | 1973% | 891 | 0.55% | 11% | 61.24 | 22 | 17,440 |
| | ., | , , , , , | | | | -, -, | | | | | | | |

| | No change in sea level | | | | | 40cm sea le subsidence (r | evel rise and no adaptation) | (adaptation | evel rise and at constant tation scenar | probability, | 40cm sea level rise and subsidence (adaptation at constant relative risk, adaptation scenario PL) | | | |
|--|-------------------------|---|--|---------------------------|----------------------------------|------------------------------|---|---------------------------|---|---|---|---|--------------------------------------|--|
| Urban Agglomeration | 100-yr exposition (M\$) | 100-yr losses with no protection (M\$) | Protection level in 2005 (return period in years) | Mean annual loss (M\$) | Mean annual loss (% city GDP) | Mean annual loss (M\$) | Mean increase due to SLR and subsidence | Mean annual loss (M\$) | Mean annual loss (% city GDP) | Mean increase due to SLR and subsidence | Increase in dike height (for 40cm SLR + subsidence) | New protection level to maintain relative risk (return period in years) | New loss if protection is overtopped | |
| Abidjan Accra | 4,017 190 | 1,328 57 | 10 10 | 826 36 | 0.72% 0.03% | 10,836 | 1213% | 6 1,187 6 49 | 1.03% | 36% | 42.18 41.52 | 15 14 | 11,623 480 | |
| Adelaide | 1,278 | 643 | 100 | 4 | 0.00% | 88 | 2086% | 5 | | 14% | 41.73 | 114 | 446 | |
| Al Kuwayt (Kuwait City) | 1,579 | 549 | 100 | 9 | 0.00% | 892 | 10242% | 6 9 | 0.00% | 10% | 40.55 | 110 | 936 | |
| Al-Iskandariyah (Alexandria) | 27,675 | 3,669 | 100 | 199 | 0.09% | 44,877 | 22455% | | 0.25% | 192% | 87.88 | 304 | 58,470 | |
| Amsterdam Athinai (Athens) | 187,159 487 | 103,598 113 | 10000 | 9 | 0.01% 0.00% | 102 73 | 1042% | 6 10 | 0.01% | 13% 98% | 84.13 44.18 | 11,340 207 | 96,240 107 | |
| Auckland | 1,119 | 563 | 100 | 4 | 0.00% | 419 | 9914% | 6 5 | | 13% | 41.1 | 114 | 463 | |
| Baixada Santista (Santos) | 788 | 218 | 50 | 9 | 0.01% | 467 | 5256% | 6 10 | 0.01% | 15% | 40.82 | 58 | 493 | |
| Baltimore | 31,595 | 15,406 | 50 | 238 | 0.08% | 5,841 | 2359% | 299 | 0.10% | 26% | 43.03 | 64 | 14,367 | |
| Banghazi Barcelona | 3,048 | 734 | 100 | 15 | 0.01% | 1,588 | 10595% | 28 | | 89% | 45.56 | 200 | 2,748 | |
| Barranquilla | 2,370 94 | 549 28 | 100 10 | 3 5 | 0.00% | 307 102 | 10731% 2106% | 6 5 6 11 | | 75% 143% | 44.71 86.94 | 183 25 | 484 113 | |
| Bayrut_(Beirut) | 336 | 45 | 100 | 1 | 0.00% | 164 | 15583% | 6 2 | 0.00% | 136% | 44.37 | 251 | 246 | |
| Belém Boston | 2,007 | 542 | 50 50 | 12 741 | 0.01% 0.13% | 586 | 4955% | 6 13 6 849 | | 15% 15% | 41.41 | 58 | 654 | |
| Brisbane | 124,752 12,803 | 69,675 5,486 | 100 | 741 | 0.13% | 37,206 1,076 | 4924% 3552% | 6 849 | | 22% | 41.4 42.35 | 58 124 | 41,249 3,459 | |
| Buenos Aires | 6,847 | 1,868 | 50 | 44 | 0.00% | 592 | 1257% | 56 | 0.00% | 28% | 44.13 | 65 | 2,602 | |
| Cape Town Chennai (Madras) | 757 3,457 | 209 966 | 100 10 | 4 825 | 0.00% 0.12% | 453 9,347 | 9998% | 6 5 6 1,028 | 0.00% 0.15% | 15% 25% | 41.15 42.07 | 115 13 | 501 10,024 | |
| Chittagong | 3,457 4,751 | 1,186 | 20 | 825 154 | 0.12% | 9,347 2,934 | 1033% | | | 25% 11% | 42.07 | 13 | 3,288 | |
| Ciudad de Panama (Panama City) | 468 | 200 | 10 | 42 | 0.06% | 451 | 962% | 6 47 | | 10% | 40.51 | 11 | 459 | |
| Conakry | 449 | 166 | 10 | 66 | 0.00% | 831 | 1155% | 6 89 | | 34% | 41.51 | 14 | 872 | |
| Dakar | 177 | 48 | 10 | 35 | 0.04% | 429 | 1121% | 45 | 0.05% | 29% | 41.32 | 13 | 446 | |
| Dalian | 1,973 | 750 | 20 | 143 | 0.04% | 3,012 | 2010% | 166 | | 16% | 41.43 | 23 | 3,241 | |
| Dar-el-Beida (Casablanca) Dar-es-Salaam | 940 | 301 41 | 100 | 12 62 | 0.01% | 1,256 660 | 10159% 969% | 6 14 6 69 | | 16% 11% | 41.13 40.8 | 117 11 | 1,395 677 | |
| Davao | 61 | 22 | 10 | 5 | 0.07% | 60 | 995% | 6 6 | | 16% | 40.8 | 12 | 62 | |
| Dhaka | 10,126 | 2,774 | 20 | 451 | 0.10% | 9,219 | 1944% | 524 | | 16% | 83.55 | 23 | 10,154 | |
| Douala Dubour (Duboi) | 162 49,574 | 48 16,861 | 10 100 | 24 286 | 0.03% 0.11% | 311 29,215 | 1179% 10114% | 6 34 6 326 | | 40% 14% | 83.43 41.01 | 14 115 | 333 31,900 | |
| Dubayy (Dubai) Dublin | 6,825 | 1,866 | 70 | 7 | 0.11% | 532 | 7114% | 6 9 | | 20% | 41.53 | 84 | 606 | |
| Durban | 1,094 | 317 | 500 | 1 | 0.00% | 746 | 27710% | 6 2 | | 14% | 40.81 | 569 | 826 | |
| EI_Djazaïr (Algiers) Fortaleza | 363 | 88 | 50 | 7 | 0.00% | 452 | 6805% | 6 12 6 3 | | 76% | 43.26 | 91 | 566 | |
| Fukuoka-Kitakyushu | 493 87,966 | 160 42,913 | 50 100 | 177 | 0.00% | 108 21,434 | 5814% 11998% | 6 235 | | 62% 33% | 43.15 81.87 | 85 133 | 141 23,251 | |
| Fuzhou_Fujian | 2,651 | 919 | 20 | 137 | 0.05% | 4,148 | 2926% | 6 237 | 0.09% | 73% | 82.89 | 35 | 4,637 | |
| Glasgow Grande_Vitória | 4,365 15,160 | 1,757 | 200.00 50 | 4 190 | 0.00% 0.23% | 824 10,096 | 20293% 5208% | 6 5 6 227 | 0.01% 0.28% | 12% 20% | 40.71 41.38 | 224 60 | 890 11,163 | |
| Guangzhou_Guangdong | 86,644 | 4,712 37,105 | 20 | 11,928 | 1.32% | 262,207 | 2098% | 13,537 | 1.49% | 13% | 81.1 | 23 | 268,090 | |
| Guayaquil | 8,296 | 3,133 | 10 | 2,813 | 0.95% | 32,267 | 1047% | 3,278 | | 17% | 80.79 | 12 | 32,555 | |
| Hai Phòng Hamburg | 14,283 59,084 | 4,728 28,626 | 50 1000 | 320 14 | 0.37% 0.01% | 16,029 268 | 2482% 1825% | 6 418 6 15 | | 40% 11% | 85.97 41.49 | 67 1,116 | 19,432 15,064 | |
| Hangzhou | 586 | 149 | 20 | 6 | 0.00% | 133 | 2252% | 6 7 | 0.00% | 27% | 82.68 | 26 | 141 | |
| Helsinki | 2,482 | 936 | 200 | 3 | 0.00% | 38 | 1093% | 6 4 | 0.00% | 17% | 42.76 | 237 | 703 | |
| Hiroshima Hong Kong | 52,494 60,722 | 23,307 31,300 | 100 900 | 80 140 | 0.06% | 10,421 11,514 | 12860% 6067% | 6 117 6 159 | | 46% 14% | 82.57 41.19 | 147 1,026 | 11,584 140,038 | |
| Houston | 29,147 | 12,337 | 50 | 119 | 0.02% | 7,246 | 5990% | 6 214 | | 80% | 89.45 | 95 | 10,389 | |
| Inchon | 22,825 | 9,493 | 100 | 29 | 0.01% | 1,248 | 4166% | 6 31 | 0.02% | 5% | 40.55 | 105 | 3,030 | |
| Istanbul | 3,162 1,415 | 419 341 | 100 100 | 13 | 0.00% | 1,746 997 | 12980% | 6 29 6 14 | | 119% 101% | 45.31 43.9 | 234 209 | 2,890 1,411 | |
| Jakarta | 9,577 | 2,553 | 100 | 1,139 | 0.14% | 17,852 | 1467% | 1,872 | | 64% | 83.32 | 17 | 18,540 | |
| Jiddah | 1,092 | 393 | 10 | 60 | 0.02% | 702 | 1062% | 75 | | 24% | 41.06 | 12 | 733 | |
| Karachi Khulna | 792 4,665 | 260 1,748 | 10 20 | 121 295 | 0.02% 0.43% | 1,451 6,851 | 1095% 2219% | 6 151 6 459 | | 24% 56% | 81.97 88.15 | 12 32 | 1,489 8,788 | |
| København (Copenhagen) | | | 200 | 7 | | | | | | | | | | |
| København (Copenhagen) Kochi (Cochin) | 9,567 1,923 | 2,462 505 | 10 | 430 | 0.01% 0.29% | 177 4,837 | 2308% 1026% | 6 9 6 521 | 0.01% | 28% 21% | 43.24 41.64 | 259 12 | 1,784 5,096 | |
| Kolkata (Calcutta) | 33,231 | 12,649 | 20 | 2,704 | 0.21% | 60,930 | 2154% | 3,515 | 0.27% | 30% | 84.56 | 26 | 68,537 | |
| Krung_Thep (Bangkok) Kuala Lumpur | 49,065 19,437 | 16,462 6,696 | 50 100 | 596 56 | 0.07% 0.03% | 30,827 1,153 | 5073% 1971% | 6 818 6 69 | | 37% 24% | 85.93 42.96 | 71 125 | 38,153 6,607 | |
| La Habana (Havana) | 19,437 | 8 | 100 | 0 | 0.03% | 21 | 13660% | 6 0 | | 80% | 42.96 | 185 | 28 | |
| Lagos | 2,750 | 728 | 10 | 287 | 0.15% | 3,407 | 1087% | 360 | 0.19% | 26% | 41.33 | 13 | 3,549 | |
| Lima Lisboa (Lisbon) | 47 4,764 | 14 1,770 | 10 100 | 4 13 | 0.00% 0.01% | 48 747 | 1254% 5521% | | 0.00% | 48% 18% | 42.15 41.7 | 15 119 | 51 1,530 | |
| Lomé | 1,005 | 308 | 100 | 28 | 0.01% | 3,148 | 11112% | | | 18% 29% | 41.7 | 119 | 3,562 | |
| London | 101,543 | 36,751 | 1000 | 13 | 0.00% | 313 | 2246% | | | 17% | 42.14 | 1,184 | 15,010 | |
| Los Angeles-Long Beach Santa Ana | 35,833 | 15,517 | 50 | 188 | 0.08% | 10,158 | 5301% | | | 15% | 40.72 | 58 | 10,697 | |
| Luanda Maceió | 21 | 7 | 10 | 9 | 0.00% | 105 | 1061% | 6 11 6 7 | | 21% | 41.04 | 12 | 108 | |
| Maceió Manila | 518 2,339 | 188 872 | 50 10 | 6 254 | 0.01% 0.05% | 283 3,123 | 5025% 1128% | | 0.01% | 22% 44% | 41.85 85.11 | 62 15 | 323 3,489 | |
| Maputo | 404 | 169 | 100 | 24 | 0.03% | 1,389 | 3289% | 6 27 | 0.03% | 13% | 41.15 | 112 | 2,672 | |
| Maracaibo | 775 | 236 | 100 | 6 | 0.01% | 588 | 10238% | 6 8 | 0.01% | 40% | 42.84 | 142 | 773 | |
| Marseille-Aix-en-Provence | 3,403 | 788 | 100 | 3 | 0.00% | 358 | 10971% | | | 100% | 45.86 | 214 | 637 | |
| Melbourne Miami | 682 824,448 | 328 387,360 | 100 100 | 2,099 | 0.00% | 42 25,674 | 1784% 1123% | | 0.00% | 18% 45% | 42.43 46.15 | 120 147 | 249 277,610 | |
| Montevideo | 992 | 334 | 50 | 2,099 | 0.02% | 180 | 11237 | | | 18% | 42.85 | 60 | 785 | |
| Montréal | 4,235 | 2,482 | 50 | 14 | 0.00% | 510 | 3451% | 6 15 | 0.00% | 8% | 40.84 | 54 | 757 | |
| Mumbai (Bombay) Muqdisho (Mogadishu) | 52,173 21 | 19,827 | 20 10 | 6,109 13 | 0.47% 0.04% | 128,081 136 | 1997% 982% | 6 6,664 6 14 | | 9% 13% | 40.61 40.77 | 22 11 | 131,802 140 | |
| Muqaisno (Mogadisnu) Nagoya | 175,473 | 99,196 | 100 | 564 | 0.04% | 60,738 | 10677% | | | 13% | 81.81 | 118 | 65,731 | |
| N'ampo | 1,141 | 383 | 10 | 24 | 0.31% | 245 | 906% | 6 26 | 0.33% | 6% | 40.88 | 11 | 253 | |
| Napoli (Naples) Natal | 981 647 | 125 250 | 50 50 | 1 9 | 0.00% 0.02% | 81 487 | 6216% 5100% | 6 3 6 11 | 0.00% | 188% 15% | 46.27 41.05 | 155 58 | 133 527 | |
| New Orleans | 323,917 | 180,646 | 100 | 1,583 | 1.21% | 171,210 | 10716% | | | 23% | 41.05 82.47 | 123 | 190,749 | |
| New York-Newark | 532,194 | 291,729 | 100 | 1,960 | 0.08% | 31,949 | 1530% | 2,159 | 0.09% | 10% | 41.43 | 110 | 209,043 | |
| Ningbo Odessa | 6,272 2,733 | 1,373 859 | 20 100 | 176 19 | 0.09% 0.04% | 5,137 2,030 | 2813% 10333% | | | 60% 26% | 83.08 41.75 | 32 126 | 5,561 2,391 | |
| Osaka-Kobe | 337,354 | 177,054 | 300 | 261 | 0.03% | 92,559 | 35430% | 6 366 | | 41% | 82.47 | 425 | 107,860 | |
| Palembang | 2,612 | 934 | 10 | 418 | 0.39% | 5,016 | 1100% | | | 26% | 82.49 | 13 | 5,183 | |

| Perth | 4,220 | 2,145 | 100 | 16 | 0.01% | 265 | | 18 | 0.01% | 17% | 42.34 | 118 | 1,762 |
|--|----------|-----------|-------|-------|-------|--------|--------|-------|-------|------|-------|--------|--------|
| Philadelphia | 49,797 | 27,432 | 50 | 279 | 0.04% | 3,713 | | 309 | 0.04% | 11% | 41.66 | 56 | 14,908 |
| Port-au-Prince | 9 | 3 | 10 | 1 | 0.00% | 11 | | 1 | 0.00% | 77% | 42.81 | 18 | 12 |
| Portland | 3,753 | 1,716 | 50 | 3 | 0.00% | 156 | 5455% | 4 | 0.00% | 28% | 41.39 | 65 | 175 |
| Porto | 1,756 | 722 | 100 | 6 | 0.01% | 179 | 2667% | 7 | 0.01% | 13% | 41.57 | 114 | 715 |
| Porto Alegre | 0 | 0 | 50 | 10 | 0.00% | 483 | 4918% | 12 | 0.01% | 21% | 42 | 61 | 550 |
| Providence | 17,857 | 9,479 | 50 | 118 | 0.07% | 2,333 | 1872% | 135 | 0.08% | 14% | 41.82 | 57 | 6,515 |
| Pusan | 16,852 | 4,523 | 100 | 30 | 0.01% | 3,109 | 10127% | 36 | 0.01% | 18% | 41.36 | 119 | 3,529 |
| Qingdao | 3,055 | 958 | 20 | 163 | 0.05% | 3,546 | 2079% | 189 | 0.06% | 16% | 41 | 23 | 3,720 |
| Rabat | 248 | 71 | 100 | 2 | 0.00% | 217 | 10837% | 3 | 0.00% | 34% | 41.77 | 135 | 257 |
| Rangoon | 4,177 | 1,387 | 10 | 163 | 0.17% | 1,942 | 1093% | 211 | 0.22% | 30% | 84.05 | 13 | 2,062 |
| Recife | 1,218 | 477 | 50 | 19 | 0.01% | 970 | 5063% | 21 | 0.01% | 14% | 41.04 | 57 | 1,047 |
| Rio_de_Janeiro | 4,132 | 1,010 | 50 | 35 | 0.01% | 1,803 | 5108% | 40 | 0.01% | 17% | 41.28 | 59 | 1,971 |
| Rotterdam | 172,316 | 96,071 | 10000 | 7 | 0.01% | 840 | 11417% | 8 | 0.01% | 15% | 82.34 | 11,489 | 81,871 |
| Salvador | 377 | 143 | 50 | 5 | 0.00% | 262 | 5248% | 6 | 0.00% | 16% | 40.75 | 58 | 277 |
| San Diego | 1,369 | 731 | 50 | 12 | 0.00% | 641 | 5215% | 14 | 0.00% | 13% | 40.72 | 57 | 674 |
| San Francisco - Oakland | 34,156 | 13,240 | 50 | 149 | 0.03% | 7,834 | 5166% | 185 | 0.04% | 24% | 41.84 | 63 | 8,966 |
| San Jose | 2,826 | 932 | 50 | 2 | 0.00% | 67 | 4133% | 2 | 0.00% | 22% | 42.22 | 62 | 92 |
| San Juan | 7,342 | 2,235 | 50 | 68 | 0.02% | 4,238 | 6118% | 107 | 0.03% | 56% | 42.92 | 80 | 5,202 |
| Sankt Peterburg (St. Petersburg) | 52,615 | 11,899 | 1000 | 11 | 0.00% | 3,119 | 27857% | 13 | 0.01% | 17% | 41.13 | 1,173 | 12,643 |
| Santo Domingo | 636 | 153 | 10 | 21 | 0.02% | 410 | 1880% | 50 | 0.04% | 141% | 44.62 | 26 | 499 |
| Sapporo | 1,339 | 591 | 100 | 2 | 0.00% | 402 | 16647% | 5 | 0.00% | 95% | 84.19 | 198 | 464 |
| Seattle | 10,235 | 6,100 | 50 | 85 | 0.02% | 3,499 | 4026% | 90 | 0.02% | 6% | 40.66 | 53 | 4,439 |
| Shanghai | 77,189 | 28,576 | 1000 | 63 | 0.00% | 24,763 | 39168% | 102 | 0.01% | 62% | 86.82 | 1,664 | 99,775 |
| Shenzen | 25,510 | 10,362 | 20 | 2,929 | 0.38% | 59,948 | 1947% | 3,338 | 0.43% | 14% | 41.5 | 23 | 64,947 |
| Singapore | 3,412 | 2,020 | 2000 | 2 | 0.00% | 359 | 17364% | 2 | 0.00% | 7% | 40.48 | 2,128 | 4,339 |
| Stockholm | 517 | 152 | 100 | 1 | 0.00% | 90 | 10941% | 1 | 0.00% | 48% | 42.86 | 150 | 118 |
| Surabaya | 727 | 163 | 10 | 80 | 0.04% | 1,126 | 1316% | 119 | 0.06% | 50% | 82.55 | 15 | 1,170 |
| Surat | 7,399 | 3,078 | 10 | 905 | 0.25% | 9,298 | 927% | 949 | 0.27% | 5% | 40.4 | 10 | 9,403 |
| Sydney | 8,293 | 4,314 | 100 | 28 | 0.01% | 1,046 | 3609% | 34 | 0.01% | 20% | 42.11 | 121 | 3,240 |
| Taipei | 2,713 | 1,042 | 20 | 274 | 0.10% | 7,003 | 2458% | 375 | 0.13% | 37% | 81.74 | 28 | 7,390 |
| Tampa-St Petersburg | 111,585 | 56,155 | 50 | 763 | 0.26% | 11,777 | 1444% | 948 | 0.32% | 24% | 43.39 | 63 | 45,042 |
| Tarabulus (Tripoli) | 267 | 64 | 100 | 2 | 0.00% | 169 | 10365% | 3 | 0.00% | 68% | 44.56 | 175 | 265 |
| Tel Aviv-Yafo (Tel Aviv-Jaffa) | 26 | 6 | 100 | 0 | 0.00% | 9 | 10931% | 0 | 0.00% | 87% | 45.18 | 196 | 15 |
| Thành-Pho-Ho-Chí-Minh (Ho Chi Minh City) | 42,093 | 20,216 | 50 | 1,743 | 0.74% | 11,842 | | 2,032 | 0.86% | 15% | 86.96 | 59 | 94,788 |
| Tianjin | 25,667 | 9,779 | 20 | 1,810 | 0.24% | 43,488 | | 2,383 | 0.31% | 32% | 83.4 | 26 | 46,759 |
| Tokyo | 276,547 | 140,711 | 1000 | 58 | 0.00% | 67,958 | | 85 | 0.00% | 47% | 82.65 | 1,482 | 83,432 |
| Ujung Pandang | 156 | 33 | 10 | 11 | 0.01% | 118 | | 13 | 0.02% | 21% | 42.21 | 12 | 127 |
| Ulsan | 1,209 | 368 | 100 | 2 | 0.00% | 269 | | 4 | 0.00% | 56% | 42.98 | 159 | 350 |
| Vancouver | 75,276 | 43,232 | 50 | 325 | 0.14% | 20,765 | | 466 | 0.20% | 43% | 82.24 | 72 | 22,858 |
| Virginia Beach | 138,390 | 62,671 | 140 | 278 | 0.15% | 8,304 | | 328 | 0.17% | 19% | 42.02 | 166 | 44,202 |
| Visakhapatnam | 369 | 111 | 10 | 110 | 0.08% | 1,242 | 1029% | 131 | 0.09% | 19% | 41.23 | 12 | 1,289 |
| Washington, D C | 12,326 | 5,972 | 50 | 74 | 0.01% | 3,914 | 5225% | 91 | 0.02% | 23% | 41.63 | 62 | 4,410 |
| Wenzhou | 3,187 | 921 | 20 | 152 | 0.06% | 3,488 | 2199% | 180 | 0.08% | 19% | 81.06 | 24 | 3,557 |
| Xiamen | 10,093 | 3,682 | 20 | 572 | 0.22% | 13,216 | 2210% | 793 | 0.31% | 39% | 84.89 | 28 | 15,322 |
| Yantai | 679 | 164 | 20 | 21 | 0.01% | 484 | 2257% | 29 | 0.01% | 40% | 42.36 | 28 | 553 |
| Zhanjiang | 6,256 | 2,865 | 20 | 806 | 0.50% | 17,287 | 2045% | 926 | 0.57% | 15% | 81.72 | 23 | 18,128 |
| | -, -, -, | , , , , , | | | | | | | | | | | |