The Chennai Water Crisis: Insufficient rainwater or suboptimal harnessing of runoff?

Sumant Nigam*, Alfredo Ruiz-Barradas and Agniv Sengupta

Chennai experienced acute water shortage during 2019 summer, and four years prior, an early-winter deluge. Analysis of 116 years (1901–2016) of rainfall in Chennai Sub-basin shows a weak climate change signal: Winter monsoon rainfall, has slightly increased, especially in December. The much larger Cauvery basin to the south also exhibits a nondescript climate change signal in winter rainfall. Late summer (September) rainfall in the Cauvery Basin has, however, precipitously declined in recent years (1987–2016). We show that this decline, as well as the mid-20th century increase, are attributable to natural multidecadal climate variability (Atlantic Multidecadal Oscillation) – cautioning against cavalier attributions of recent-period trends and the Chennai Water Crisis to climate change. Analysis of runoff – the rainwater leftover after its hydrologic and atmospheric processing – shows that harnessing even half of the winter monsoon runoff in the Chennai Sub-basin can satiate the city’s water demand for about seven months; and without needing new reservoir facilities. The present analysis suggests that Chennai’s water woes arise not from insufficient rainwater, but from the suboptimal harnessing of related runoff.

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monsoon, and precipitously down to $\sim 0.5 \times 10^6 \text{ m}^3$ (<1% average-fill) in July 2019 in the peak phase of the Chennai Water Crisis, which followed the delayed onset of the SW monsoon. The reservoirs are primed by summer monsoon and filled during winter monsoon, attaining peak levels in December and January, with fill levels dependent on fluctuant monsoons. El Niño – the warm phase of the El Niño Southern Oscillation – which suppresses summer rainfall over most of the Indian subcontinent$^{11,12}$ – increases winter monsoon rainfall in the Coromandel, especially in November$^7$.

Chennai’s reservoir-based water supply was recently augmented by water from the Krishna and Cauvery rivers. Water from the Krishna (in the north) comes via Telugu Ganga Canal – a 405 km canal linking Srisailam reservoir in Andhra Pradesh to the Poondi reservoir. Krishna water arrivals began in 2004, but the delivery has been unsteady because of multi-state claims on river water, and because of farmer resistance and diversions along the canal$^13$, leading to significantly smaller receipts than the annually permitted July–October uptake ($\sim 425 \times 10^6 \text{ m}^3$); for example, only a tenth was received by early 2019 (ref. 14). Water from the Cauvery (in the south) was tapped via Veeranam Lake, located 225 km south of Chennai and just upstream of the Cauvery Delta. Piped water delivery from Veeranam Lake also commenced in 2004 but has been erratic, being dependent on the release of surplus Cauvery water through the Mettur Dam; for example, in April 2019, $0.17 \times 10^6 \text{ m}^3$ of water was supplied each day to Chennai$^{15}$. Two desalination plants along the coast at Minjur and Nemmeli, operative since 2010 and 2013 respectively, provide another $\sim 0.2 \times 10^6 \text{ m}^3$ of water each day to Chennai$^{16}$. A growing movement to harvest rainwater also helps via groundwater replenishment$^7$.

The Chennai Water Crisis came to a head during mid-summer of 2019, drawing wide media attention$^{4,9,18,19}$ and spurring causal analyses. Increased demand and loss of precious wetlands from urban sprawl and auto industry growth, upstream deforestation, water resources mismanagement, and droughts and climate change$^9$ have all been implicated.

The present study targets the hydroclimate of the Coromandel region. Regional rainfall is analysed to identify both long-term (i.e. century-long) and short-term (over recent decades) trends. The long-term trends are indicative of climate change, while short-term trends of multidecadal climate variability. Rainfall trends are analysed for the Chennai Sub-basin and the neighbouring Cauvery basin. Subsequent analysis of runoff – the rainwater leftover after hydrologic and atmospheric processing of rainfall – completes the study providing insights on the surface-water potential of the region. The analysis provides the hydrometeorological context for Chennai’s water woes, including whether they arise from insufficient rainwater or suboptimal harnessing of runoff.

Data and methods

**Physiography data**

Topography and bathymetry were obtained from the National Oceanic and Atmospheric Administration (NOAA’s) National Oceanographic Data Center’s ETOPO1 Global Relief Model$^{20}$, a 1 arc-min resolution relief model of the Earth’s surface. River shapefiles were obtained from the Global Runoff Data Centre’s Major River Basins of the World$^{21}$ and the AQUAMAPS rivers of the world$^{22}$, while watershed boundaries were from the World Wildlife Fund’s Conservation Science Program’s HydroSHEDS 30 arc-sec resolution georeferenced data$^{23}$. The basins were discretized using MeteoInfo software$^{24}$, whose output is compatible with the Grid Analysis and Display System (GrADS)$^{25}$, a data processing and visualization program used extensively in this analysis.

Figure 1 shows the physio-geography of southern Peninsular India. Cauvery, the largest river in the region,
has its headwaters in the Western Ghats, while Ponnaiyar, Palar and Penner rivers have their source in the Deccan Plateau.

Precipitation and runoff

Rain gauge-based monthly precipitation on a 0.25° continental grid for the period 1901–2016 was from the German Meteorology Service’s (DWD) Global Precipitation Climatology Centre (GPCC, ver. 8)36. Rain gauge-based data from India Meteorological Department (IMD, ver. 4)37 was also used in select inter-comparisons. This daily data (1901–present) is based on ~7800 stations and gridded on a 0.25°-grid.

Runoff was obtained from the Global Runoff (GRUN) reconstruction dataset (ver. 1)35, where in situ streamflow observations (from Global Streamflow Indices and Metadata archive) train a machine learning algorithm that predicts monthly runoff rates based on antecedent precipitation and temperature (from Global Soil Wetness Project Phase-3 meteorological forcing dataset). The ensemble mean of 50 GRUN reconstructions obtained from training with different data subsets has been used here; monthly runoff is available on a 0.5° grid during 1902–2014 from https://figshare.com/articles/GRUN_Global_Runoff_Reconstruction/9228176.

Basin rainwater receipts

Basin rainwater receipts were obtained by multiplying rainfall on the 0.25° continental grid by the grid-cell area, followed by integration over the discretized basin using the GrADS asum function, which implicitly accounts for the variation of the grid-cell area with latitude.

Multidecadal climate variability

The attribution of recent period rainfall trends requires factoring for multidecadal climate variability. Atlantic Multidecadal Oscillation (AMO) and Pacific Decadal Oscillation are two major modes of sea surface temperature (SST) variability whose influence extends to the Indian subcontinent; their influence on summer monsoon rainfall is well documented29–31, but not the influence on winter monsoon. The AMO was tracked using the NOAA-AMO index32, defined as the linearly detrended, area-averaged SST anomaly in the North Atlantic (75°–5°W, EQ-60°N). The index was smoothed with the LOESS filter using a 15% span window (LOESS-15)33; the span is ~17 years for the 1901–2016 record, which is approximately one-fourth of AMO’s estimated period (50–70 years)34–36. By suppressing the subseasonal-to-interannual-to-decadal variabilities, the LOESS-15 filter highlights the multidecadal fluctuations in the 1901–2016 record.

Rainfall analysis

Spatiotemporal distribution of rainfall

Harmonic analysis of climatological monthly rainfall yields the annual mean (contoured) and annually varying (arrows) components of rainfall37 (Figure 2, top). Annual-mean rainfall is notably large (7–8 mm/day) in the northeast and along the west coast (Konkan and Malabar), but modest (2–4 mm/day) over southeastern Peninsular India. The arrows succinctly portray the monsoon nature (i.e. monsoonal nature) of rainfall, with July being the peak rainfall month across most of the subcontinent, reflecting the pervasive influence of the SW summer monsoon. The red-arrow region in Figure 2 indicates rainfall peaks in October–December. This early-winter rainfall is from the NE monsoon which brings moisture from the Bay of Bengal via low-level northeasterlies7. The red-arrow region effectively outlines the Coromandel Coast, as noted in the introduction. The winter monsoon also prevails over Sri Lanka, except for its southwest corner.

The rainfall distribution in the pre-summer monsoon, summer monsoon and winter monsoon periods is overlaid on the river basins of Peninsular India (Figure 2, lower panels). The Coromandel receives rainfall from the winter monsoon (6–8 mm/day) and from the summer monsoon (2–4 mm/day). The Chennai Sub-basin gets twice as much rainfall in early winter as summer. The premonsoon period (January–May) is the dry period over the Eastern Basin, including the Chennai Sub-basin. The Cauvery basin gets more rainfall during the winter monsoon, but the Penner basin receives more during the summer monsoon.

Climatological basin rainfall and rainwater receipts

Figure 3 shows the monthly variation of average rainfall and rainwater volume received in each basin. Table 1 notes the period accumulations. Basin-averaged rainfall generally increases from late spring to early winter, with only the Cauvery basin exhibiting a secondary peak in July. Rainfall in the Penner basin – the northernmost – peaks in September, i.e. at the tail end of the SW monsoon. Peak rainfall in the other basins occurs during the winter monsoon: in October in the Cauvery and Eastern basins, and in November in the Chennai Sub-basin. Basin rainfall is largest in the Chennai Sub-basin, as it is most exposed to the coastally focused NE winter monsoon.

The volume of rainwater received in each basin – the basin rainwater receipts – depends both on
Figure 2. Climatological rainfall over the Indian subcontinent. (Upper panel) Arrows represent the annual cycle (first harmonic) and contours the annual mean (mm/day) of rainfall. The arrow scale and phase convention are shown in the lower-right corner; arrows pointing north indicate July as the peak rainfall month, and so on. Arrows in red mark regions receiving peak rainfall during October–December, i.e. during the northeast winter monsoon. Arrows are plotted when their amplitude is ≥0.75 mm/day. Contouring and shading interval as in the colour bar on the left. (Lower panel) Climatological rainfall over southern Peninsular India, with blue and red lines denoting the main rivers and their basin boundaries, following Figure 1. The 1901–2016 GPCC rainfall climatology is analysed. The fields are shown after four applications of $s_{mth9}$ in GrADS.

basin-averaged rainfall and basin area. Cauvery, with the largest area (93,814 km²), is the clear frontrunner in all calendar months (Figure 3, lower panel). In early summer, its rainwater receipts are more than twice as large as the Penner and Eastern basins, and many times larger than the Chennai Sub-basin, which because of its small size (6,773 km²) ranks the lowest even in early winter when it has the largest rainfall. Other notable features include (also from Table 1):

- Cauvery basin is unique because of exposure to both monsoons.
- Large basins with deep extensions into the interior (e.g. Cauvery, Penner, Palar and Ponnatayar) receive more rainwater during summer monsoon.
- Smaller, coastally confined basins (e.g. Naidupet, Chennai and Varahanadhi), on the other hand, receive significantly more rainwater during winter monsoon. For example, Chennai Sub-basin receives ~38% of its annual water from the summer monsoon, but as much as ~51% from the winter monsoon.

The amount of rainwater received in the Chennai Sub-basin is compared with the average-fill volume of
Table 1. Basin-averaged rainfall (mm/day) and basin-integrated rainwater volume (m³/day) in river basins of Peninsular India based on 1901–2016 climatology; seasons are based on monsoons. Seasonal contribution to annual rainwater accumulation (%) is noted. River basins are listed in descending order of their area. GPCC precipitation on a 0.25° grid is analysed.

| River Basin (area)  | Annual average (January–December) | Pre-monsoon average (January–May) | Summer Monsoon average (June–September) | Winter Monsoon average (October–December) |
|---------------------|-----------------------------------|-----------------------------------|-----------------------------------------|------------------------------------------|
|                     | Rainfall (mm/day) | Annual rainwater accumulation (10⁶ m³) | Rainfall (mm/day) | Basin rainwater volume (10⁶ m³/day) | Percentage of annual rainwater accumulation | Rainfall (mm/day) | Basin rainwater volume (10⁶ m³/day) | Percentage of annual rainwater accumulation | Rainfall (mm/day) | Basin rainwater volume (10⁶ m³/day) | Percentage of annual rainwater accumulation |
| Cauvery (93,814 km²) | 2.78               | 95.103                            | 1.25                      | 117.30                                | 18.6                                      | 3.89               | 364.89                              | 46.8                                      | 3.81               | 357.31                              | 34.6                                      |
| Penner (54,624 km²)  | 1.79               | 35.743                            | 0.61                      | 33.25                                 | 14.0                                      | 2.92               | 159.6                               | 54.5                                      | 2.24               | 122.29                              | 31.5                                      |
| Palar (18,830 km²)   | 2.52               | 17.305                            | 0.88                      | 16.53                                 | 14.4                                      | 3.43               | 64.55                               | 45.5                                      | 4.00               | 75.38                               | 40.1                                      |
| Ponnaiyar (15.093 km²) | 2.44           | 13.419                            | 1.02                      | 15.46                                 | 17.4                                      | 3.24               | 48.84                               | 44.4                                      | 3.69               | 55.70                               | 38.2                                      |
| Naidupet (10,500 km²) | 2.87              | 11.012                            | 0.72                      | 7.58                                  | 10.4                                      | 2.73               | 28.63                               | 31.7                                      | 6.60               | 69.28                               | 57.9                                      |
| Chennai (6,773 km²)  | 3.01               | 7.433                             | 0.77                      | 5.22                                  | 10.6                                      | 3.46               | 23.45                               | 38.5                                      | 6.08               | 41.18                               | 50.9                                      |
| Varahanadhi (4,531 km²) | 3.11             | 5.142                             | 0.85                      | 3.83                                  | 11.3                                      | 3.57               | 16.18                               | 38.4                                      | 6.21               | 28.14                               | 50.3                                      |
Chennai’s four rainfed reservoirs, annual recruitment of the Cauvery and Krishna waters, and, above all, Chennai city’s potential water needs and current supply in the runoff analysis section.

Interannual variability and trends in basin rainfall

To provide context for the Chennai Water Crisis, interannual variability and long-term trends in basin rainfall were analysed (Figure 4). While rainfall is key, its disposal—soil moisture recharge, evapotranspiration, infiltration, and runoff—can also modulate regional hydroclimate episodes. The large rainfall (7–8 mm/day) in October–November in the Chennai Sub-basin (top panel) and its impressive year-to-year variability (SD of 4–6 mm/day is two-thirds of climatology) stand out; the larger variability of winter monsoon was noted earlier. Such large variations, if superposed, on long-term trends, can generate extreme episodes. (Attribution of the 2015 early-winter Chennai deluge however requires a more elaborate analysis framework, one extending beyond monthly timescales to the weekly/daily ones.)

The centennial rainfall trend in the Chennai Sub-basin (thick red line) is near-zero, but positive during September–November; notably positive in December (+1.3 mm/day/century, vis-à-vis climatology of ~3.3 mm/day); and negative in January (~1.1 mm/day/century, against climatology of ~0.8 mm/day). As centennial trends—reflecting more of the climate change signal—are positive in each month but January, global climate change cannot be implicated in the Chennai Water Crisis, at least from the perspective of rainfall, the controlling element of atmospheric and terrestrial water cycles. The rainfall decline in January—significant, as it is a dry month (<1 mm/day) to begin with—cannot be overlooked, but its hydroclimate impact is likely ameliorated as it follows December, a positive-trend month. Climate change, on the other hand, can perhaps be implicated in December being more wet in the Chennai Sub-basin, but not necessarily in the generation of intense weekly episodes such as the early-winter deluge in 2015.

The recent 30-yr trends are larger than centennial ones from exposure to multidecadal climate variability (50–70-yr timescale) and, possibly, accelerated climate change; these trends also exhibit notably less month-to-month coherence than their centennial counterparts. Recent trends in the Chennai sub-basin are strongly positive in October and December (+3.76 mm/day/century) and negative in November. The December trend is the strongest of all months.

Centennial rainfall trends over the Cauvery basin are near-zero but for July, August and January when they are negative; e.g. the trend is ~0.6 mm/day/century in July when climatological rainfall is ~4.5 mm/day, indicating a...
more than a 10% reduction in July rainfall in the 20th century. January is also a challenging month for this basin from the superposition of a significant negative trend on weak climatological rainfall. The recent period trends are again larger and more volatile, e.g. September’s large negative trend vis-à-vis modest trend in the adjacent months.

**Figure 5.** Rainfall trends over Southern Peninsular India. Linear trends are shown in December (first row) (top) and September (second row); recent trends – most positive in the Chennai Sub-basin in December and most negative in the Cauvery basin in September (cf. Figure 4) – motivate the month choice. Trends are shown over 1901–2016 (left) and 1987–2016 (middle, right), the latter from both GPCCv8 and IMDv4 data; all in units of mm/day/century following the brown–green shading scale. Basin-averaged rainfall anomalies (i.e. departures from normal) during 1901–2016 are plotted for the Chennai Sub-basin in December (third row) and the Cauvery basin in September (last row). The thick black curve shows smoothed rainfall anomalies (from LOESS-15% filtering of the 116-yr yearly record) in the lower panels. Both long-period (black dashed line) and short-period (thin red line) rainfall trends in the unsmoothed record are marked and noted.

**Spatiotemporal distribution of rainfall trends**

The centennial trends are positive over the Chennai (and Naidupet) Sub-basin in December but nondescript in September (Figure 5, left column). In neighbouring basins, trends are nondescript in December but modestly negative in September, especially in central Penner and eastern
Cauvery basins. The recent period trends (middle–right columns) are far more extensive and impressive: They are positive over the Chennai Sub-basin and stronger in December than September. The December trends are more fractured along the coast, with Chennai in the bull’s eye of regional wet trends. The surrounding sea of browns indicates significant rainfall decline over the Cauvery and Penner, and Ponnamar and Palar basins in recent decades, especially in September when the GPCC and IMD trends are in good agreement. It is intriguing why the Chennai Sub-basin and its coastal vicinity are carved out region of increasing rainfall in both century-long and recent multidecadal periods, especially the latter, when this region is surrounded by steeply declining rainfall trends.

A more granular temporal view of rainfall variability is shown in the plots of basin-averaged rainfall over 1901–2016 in Figure 5 (lower panels). Departures from normal (the 116-yr monthly average) are plotted for the Chennai Sub-basin in December (a wet trend month) and Cauvery basin in September (a drying trend month). The former shows large positive departures, more frequent in the latter half of the record. The 2015 December rainfall departure, while notably positive, is not as historic as the 2015 Chennai deluge, as it is exceeded by three previous December months (1946, 1996, 1978) and nearly equalled by two others (1903, 2005); indicating that the days-to-weeks long intense rainfall episodes are not always reflected in monthly departures. No less interesting are the multiyear (3–5) periods of strongly deficient December rainfall, e.g. 1920–1924, 1947–1951, 1974–1977 and 2002–2004, when negative rainfall anomalies approach the climatological value (~3.3 mm/day), reflecting little/no rainfall in these periods. Unlike large positive departures that appear sporadic, the large negative ones exhibit multiyear coherence.

Multidecadal variability of rainfall: Atlantic Multidecadal Oscillation and the climate change signal

Figure 5 (lower panels) highlights the modulation of rainfall on longer timescales; the smooth black curve captures both the multidecadal decline in early 20th century and the increasing trend in early 21st century in December rainfall in the Chennai Sub-basin. The 116-yr trend (sloped dashed black line) is +1.20 mm/day/century; the recent 30-yr trend (red line) is also positive but larger (+3.76 mm/day/century); both consistent with trends in Figure 4. Rainfall trends in two non-overlapping 30-yr sub-records, 1901–1930 (early century) and 1945–1974 (mid-century), are strongly negative (~12.67 mm/day/century) and near zero (+0.19 mm/day/century) respectively, showing that December rainfall in the Chennai Sub-basin has not risen monotonically, as may be suggested by its centennial or recent period trends. The rainfall record instead consists of multidecadal periods of varying trends, including opposite signed ones.

The Cauvery basin rainfall in September (when recent period trends are most negative, Figure 4) also exhibits significant multidecadal variability against the backdrop of a weakly negative centennial trend (~0.30 mm/day/century; dashed black line). The recent period trend is strongly negative, but it follows a mid-century period (1945–1974) of strongly positive trend (+5.60 mm/day/century), which, in turn, follows an early-century period of negative trend (~2.22 mm/day/century). In short, multidecadal trends of alternate sign underlie the weakly negative centennial trend.

Influence of Atlantic Multidecadal Oscillation

AMO was tracked using the NOAA–AMO index (blue line in Figure 6). Its recent variations include cooling of the North Atlantic basin since the 1950s and its warming since the mid-1970s. The AMO’s influence on Peninsular rainfall is apparent from Figure 6 (right panels). The December rainfall regressions, consisting of positive anomalies over the Chennai and Naidupet basins and negative ones elsewhere, including the Cauvery delta, broadly resemble the recent period trends in GPCC rainfall (Figure 5, top–middle). Likewise, regressions of September rainfall on the AMO index, comprising of large deficits over Penner and Cauvery basins, closely resemble the recent period September trends (Figure 5, second row). These resemblances indicate that the influence of AMO on Peninsular rainfall is significant and likely dominant in recent-period trends. The 116-yr correlation of the smoothed AMO index with December rainfall in the Chennai Sub-basin (+0.23) and September rainfall in the Cauvery basin (~0.59) support this assessment.

Climate change signal

The 116-yr-long rainfall record is clearly not sufficient to encompass multiple cycles of the AMO and as such, the trend in this century-plus period – referred to as centennial trend here – will include some aliasing of multidecadal variability, compromising the integrity of centennial trend as a marker of climate change. A more refined estimate of the climate change signal in rainfall is developed by removing the AMO influence from the rainfall record. The influence is removed by subtracting the product of AMO index (blue curve, Figure 6) and related rainfall regressions (right panel, Figure 6) from the smoothed rainfall record (thick black curve, Figure 6); the AMO-removed rainfall record is shown in Figure 6 by the thin black curve. The 116-yr trend in the AMO-removed December rainfall in the Chennai Sub-basin is +0.80 mm/day/century versus +1.20 mm/day/century when the
AMO influence is retained. The aliasing of AMO influence thus leads to significant overestimation of the climate change signal in this period. The exclusion of AMO influence flips the trend sign in September rainfall in the Cauvery basin, from \(-0.30\) to \(+0.04\) mm/day/century. Although AMO-removed trends can be further adjusted for potential aliasing of other multidecadal variabilities, this analysis strongly cautions against viewing century-long trends, let alone multidecadal ones, in observed data as markers of the climate change signal.

The AMO influence is even more impressive on short-era-period trends: December rainfall in the Chennai Sub-basin and September rainfall in the Cauvery basin exhibit steep trends during 1987–2016: \(+6.44\) and \(-3.38\) mm/day/century respectively (Figure 6). Sans AMO influence, the same trends are \(+3.20\) and \(-0.55\) mm/day/century respectively, i.e. smaller by a factor of 2–6; illustrating the fallacy of cavalier attributions of recent multidecadal trends to climate change.

Runoff analysis

Runoff in the Chennai Sub-basin: climatology and trends

The rainwater received is not all available for human needs. A significant portion is lost to soil moisture recharge (especially after dry periods), evapotranspiration (water- or energy-limited), infiltration (landscape-dependent, but leading potentially to groundwater recharge) and runoff, which can be harnessed for immediate or later use through storage.

The seasonal cycle of basin-averaged rainfall and runoff in the Chennai Sub-basin shows the May-onward increase of rainfall to mark the end of the dry period (January–May), but runoff remains minimal (~0.1 mm/day) until September (Figure 7). The modest June–August rainfall (~3 mm/day) is evidently insufficient for generating runoff, principally because of soil moisture capacitance. Runoff – the rainwater leftover after soil moisture recharge (and evapotranspiration) – is thus lagged vis-à-vis rainfall, but only during build-up of the rainy season. In the exit phase (December–February), runoff variations are near synchronous with rainfall, reflecting at-capacity soil moisture.

The runoff-to-rainfall ratio peaks in February (0.41), i.e. after winter monsoon rainfall when the soil moisture is at capacity. In the recent period (Figure 7, dashed red line), it has peaked earlier—in January—at 0.43 (=0.28/0.65), i.e. at a marginally higher value; the January ratio in the 113-yr climatology is 0.34 (=0.28/0.82), for reference. The higher runoff-to-rainfall ratio in recent January months is potentially interesting. It can arise from a more saturated soil state resulting from higher rainfall in the preceding three months in recent decades (Figure 4, top) and/or from the land-use land-cover change in recent decades that generates more runoff from the same
rainfall, i.e. a higher ratio. More hydrologic analysis and modelling are clearly needed for this attribution.

Surface rainwater in Chennai Sub-basin: limited availability or suboptimal management?

Approximately $41 \times 10^6$ m$^3$/day of rainwater is received in the Chennai Sub-basin, on average, during the winter monsoon (Table 1), but not all of it is available for human needs. An estimate of the amount remaining at the surface (surface water) after hydrologic and atmospheric processing of rainwater is provided by runoff – the available surface rainwater. The average runoff in the Chennai Sub-basin during the winter monsoon is ~0.97 mm/day (Figure 7), generating an ‘available’ surface rainwater volume of $\sim 6.57 \times 10^6$ m$^3$/day during October–December. Not all runoff can be diverted into reservoirs and lakes, nor should it be, as untamed runoff is vital for habitat and ecosystem functioning. The management of surface water is, in any case, not optimal. The estimation uncertainty and harnessing inefficiency notwithstanding, winter monsoon runoff is a pertinent water metric as harnessing regional runoff is the foremost goal of water resource management. In this context, it is both insightful and motivating to express the water stores/resources of the Chennai Sub-basin and the Chennai city’s water recruitment, demand and current supply in units of winter monsoon basin-runoff days, even if only a fraction of the basin runoff is being (can be) tapped for consumptive needs. This is motivating because it brings an immediate awareness of the possibilities.

Table 2 takes a step in this direction. For example, the December and January average-fill of the four rainfed reservoirs supplying water to Chennai is $\sim 175 \times 10^6$ m$^3$ or $\sim 27$ basin-runoff days ($\sim 175 \times 10^6/(6.57 \times 10^6)$) (ref. 10). That is, assuming full harnessing of winter monsoon runoff in the Chennai Sub-basin ($6773$ km$^2$) and ignoring direct rain-feed and consumptive withdrawals, it would take $\sim 27$ days during winter monsoon to fill these reservoirs from an empty state to their average-fill state. In practice, it would take longer because of the suboptimal harnessing and management of surface water, and because of consumptive drawdowns. Even a three times longer estimate ($\sim 81$ days) allows attainment of the average-fill state within the 90 day-long winter monsoon period.

Table 2 shows that if all of the winter monsoon runoff in the Chennai Sub-basin is harnessed and stored, it would equal $604 \times 10^6$ m$^3$ ($\sim 92$ days $\times 6.57 \times 10^6$ m$^3$/day), or $\sim 3.45$ times the average-fill of the four reservoirs. Clearly, not all winter runoff in the Chennai Sub-basin can be (should be) harnessed or stored. This cursory analysis, however, suggests that there is potential here – to harness more and store more. Can more runoff be harnessed without disrupting ecosystems and habitats? While this question is beyond the purview of the present analysis, the question on the amount of water storage that would make Chennai resilient against monsoon-related water supply disruptions is tractable.

Lessons from the 2019 Chennai Water Crisis suggest that having seven months of the city’s water demand reservoired in January can help cope with monsoon disruptions, excluding multi-year ones. The seven month storage will cover both the January–May dry period and about half of the summer monsoon period (to cover disruptions from delayed onsets). At current demand level ($1,200$ MLD), seven month of demand equals $\sim 255 \times 10^6$ m$^3$, which is only about $\sim 80$% of the full capacity of the four reservoirs ($=11,257$ million cubic feet, or $319 \times 10^6$ m$^3$) (ref. 10).

Chennai’s seven month demand can thus be met by harvesting $\sim 42$% ($\sim 255/604$) of winter monsoon runoff.
Table 2. Water in the Chennai Sub-basin: Basin-integrated winter monsoon rainwater and runoff are compared with other water resources. Runoff is the rainwater leftover after hydrologic and atmospheric processing of rainfall. As harnessing runoff is an important goal of water resource management, the water stores/resources of the region are expressed in units of basin-runoff days.

| Water in the Chennai Sub-basin during winter | Chennai Sub-basin area: 6,713 km²; Chennai metro area: 173 km² | Equivalent basin-runoff days |
|---------------------------------------------|---------------------------------------------------------------|-----------------------------|
| Runoff Volume (sub-basin integrated)        | GRUNv1, 1901–2014; October–December climatology              | 6.57 × 10⁶ m³/day           | 1 |
| Rainwater Volume (sub-basin integrated; Table 1) | GPCCv8, 1901–2016; October–December climatology               | 41.18 × 10⁶ m³/day          | –6 |
| Average-fill of Chennai’s four reservoirs   | Poondi, Cholavaram, Redhills, and Chembarambakkam 2005–2019, December–January average¹ | –175 × 10⁶ m³              | –27 |
| Krishna Water²                              | Permitted uptake (via Telugu Ganga Canal)                     | –425 × 10⁶ m³               | –65 |
|                                            | 2019 Receipt                                                 | –45 × 10⁶ m³                | –7  |
|                                            | 2020 Receipt³                                                | –166 × 10⁶ m³               | –25 |
| Cauvery Water⁴                             | From Mettur Dam via Veeranam Lake                             | –0.17 × 10⁶ m³/day          | –0.03 |
| Desalination Yield⁵                        | From Minjur and Nemmeli plants                               | –0.2 × 10⁶ m³/day           | –0.03 |
| Chennai City Water Demand⁶                 | Median estimate: 1200 million litres/day (MLD)               | –1.20 × 10⁶ m³/day          | –0.18 |
| Chennai City Water Supply⁷                  | Peak supply estimate: 850 MLD                                | –0.85 × 10⁶ m³/day          | –0.13 |
| Rainwater Harvesting⁸                       | 2017 estimate (339 million cubic feet)                       | 9.6 × 10⁶ m³               | –1.46 |

¹Computed from data at Chennai Metropolitan Water Supply and Sewage Board (CMWSSB); http://123.63.203.150/public/lake.htm
²The Indian Express, 10 January 2019 (ref. 14).
³5850 million cubic feet (mcft) of Krishna water released. The Times of India, 22 February 2020; https://timesofindia.indiatimes.com/city/chennai/with-5-85-tmcft-of-krishna-water-released-so-far-chennais-water-supply-can-last-6-months/articleshow/74251398.cms
⁴The Times of India, 20 April 2019 (ref. 15).
⁵Minjur and Nemmeli salination plants; CMWSSB (ref. 16).
⁶Chennai’s water demand estimates vary by a factor of 2: From –0.5 mcft/day (or 0.14 × 10⁶ m³/day) by The Times of India (20 April 2019; ref. 15) to ~1200 MLD by CMWSSB, and ~1,300 MLD by others (https://www.futuredirections.org.au/publication/india-chennai-confronts-another-water-crisis/).
⁷FutureDirections.org.au (26 June 2019; https://www.futuredirections.org.au/publication/india-chennai-confronts-another-water-crisis/).
⁸Water Management in Chennai; https://en.wikipedia.org/wiki/Water_management_in_Cheennai#cite_note-TH_VanishingWaterbodies-2

in the Chennai Sub-basin. To offset reservoir losses (especially evaporative), more would need to be harvested and stored: while existing reservoir capacity will permit additional storage, can more (or even 42%) runoff be harvested? Chennai’s water woes certainly do not stem from limited water warehousing capacity.

Concluding remarks

Southeastern India experienced acute water shortages during the summer of 2019, headlined by the Chennai Water Crisis. Just four years prior, the same region was in the news for the late November–early December deluge that led to significant loss of life and property. As this section of the Coromandel Coast, which receives ~50% of annual rainfall in winter, reels from the impact of extreme rainfall swings, questions on the origin of these extremes have arisen, in particular, on the anthropogenic contribution from rising global concentration of greenhouse gases as well as regional land-use land-cover change.

Analysis of the 1901–2016 rainfall record in the Chennai Sub-basin – an ~6700 km² region that surrounds and includes the ~173 km² Chennai Metro Area – shows insignificant (but positive) trends in monsoon rainfall, except in June and December when the trends (still positive) are notable. Natural climate variability, exemplified by Atlantic Multidecadal Oscillation, contributes to these trends, especially multidecadal ones, cautioning against cavalier attributions of recent-period trends and the Chennai Water Crisis to climate change.

Runoff – the rainwater leftover after its hydrologic and atmospheric processing – can be harnessed for immediate or later use through storage. Analysis of objectively obtained and observationally constrained (by river basin discharge) estimates of runoff in the Chennai Sub-basin shows low values during summer monsoon because of modest rainfall and soil-moisture recharge (from the preceding season being dry). The runoff-to-rainfall ratio...
rises steadily during winter monsoon, from ~0.1 to ~0.3.

A novel water metric is introduced to increase awareness of the runoff harvesting potential: Water stores/resources in the Chennai Sub-basin and Chennai city’s water recruitment, demand and supply are expressed in units of winter monsoon basin-runoff days, even if only a fraction of the basin runoff is being (can be) harvested for consumptive needs. For example, Chennai’s daily water demand is equivalent to ~0.18 days of winter monsoon runoff in the Chennai Sub-basin, while the city’s four reservoirs can be filled to their December and January average-fill level using ~27 days of the Chennai Sub-basin’s runoff.

It is shown that harnessing and storing ~42% of winter monsoon runoff in the Chennai Sub-basin can fulfil about seven months (January–July) of Chennai city’s water demand, covering the dry period (January–May) and buf-fering the impact of delayed monsoon onsets (June and July). It is furthermore shown that storing this runoff will not require new warehousing facilities as the four reservoirs currently supplying water to Chennai city can store ~42% of the winter runoff at just 80%-full capacity.

Can 42% of the winter runoff in the Chennai Sub-basin be harvested? Answering this question is difficult, but working backwards from the reservoirs’ average-fill state (175 x 10^10 m3) yields ~29% harvesting fraction, assuming reservoirs are fed by winter runoff but not direct rainfall. Even if this is an overestimate, harvesting 42% of the winter runoff no longer seems beyond the realm of possibilities.

In short, Chennai city’s water woes arise not from insufficient rainwater in the regional sub-basin, but sub-optimal harnessing of related runoff. The present analysis supports the notion ‘Metro cities in India are rain-rich but water-starved’, or its more nuanced version, ‘Metro cities are in runoff-rich regions but water-deprived nonetheless’.

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