

Drought Matrices

TJ

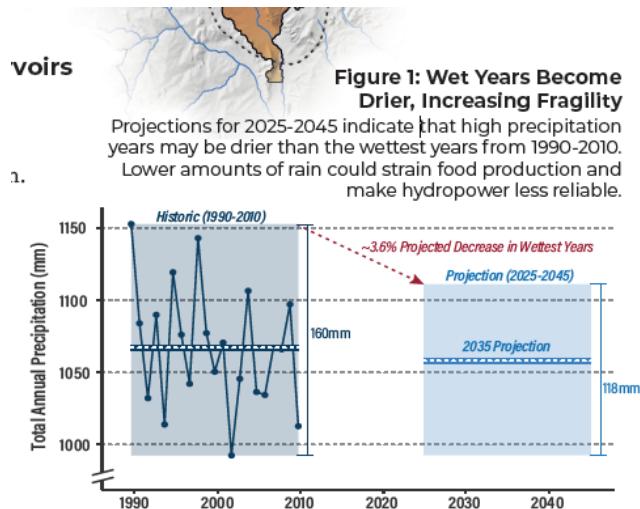
Why GWSC need drought matrices

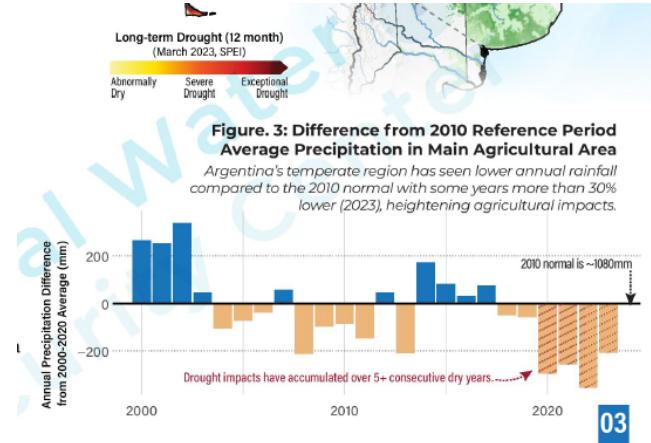
GWSC typically encounters questions such as:

1. How is precipitation during traditionally wet years expected to change in the future, and what are the potential impacts?
2. What do recent rainfall declines imply for agricultural vulnerability and resilience?

To answer them, we typically use scientifically grounded and technical metrics such as:

- Consecutive Dry Days (CDD)
- Consecutive Dry Months





We are looking for drought metrics to answer

- Frequency of drought events: How often droughts occur relative to a baseline period
- Duration of drought: How long a drought persists
- Intensity of drought: The magnitude or severity of drought conditions

Why existing drought metrics are not aligning with GWSC's needs?

Four interconnected types

- Meteorological - prolonged deficits in precipitation
↓
- Hydrological - reduced surface and groundwater availability
↓
- Agricultural - soil moisture deficits affecting crop growth
↓
- Socioeconomic - broader disruptions in water-dependent social and economic systems

Existing meteorological drought indices

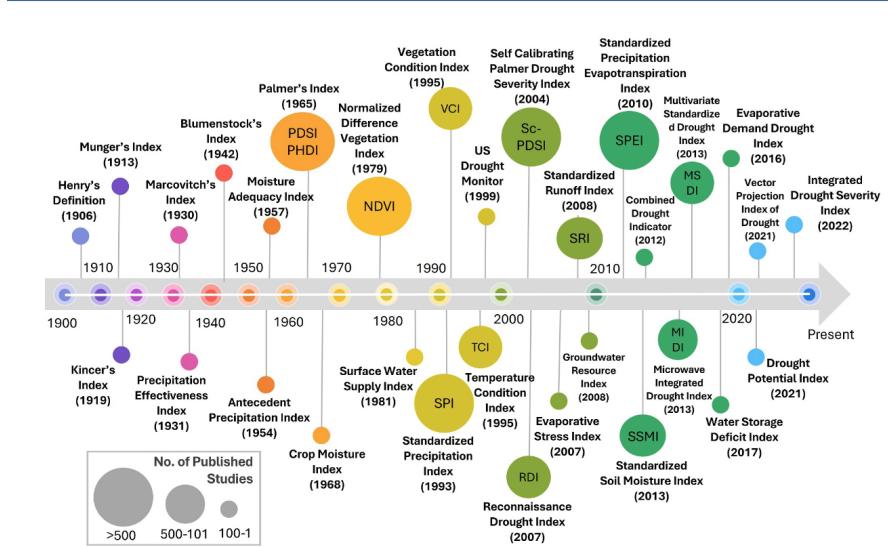


Figure 4. Advancements in Drought Monitoring Using Drought Indices: Timeline illustrating the development of major drought indices from 1900 to the present. The position of each circle along the timeline represents the temporal introduction of the index, while the size of the circle indicates the number of publications referencing the index up to 2023, reflecting its scientific usage and impact.

Most Common Indices

1. Standardized Precipitation Index (SPI)

Strengths:

- Most widely used globally
- Simple - only requires precipitation data
- Can be calculated for multiple timescales (1, 3, 6, 12, 24 months)
- Standardized, so comparable across regions and climates

Formula: Normalizes precipitation using gamma distribution

2. Standardized Precipitation Evapotranspiration Index (SPEI) Strengths:

- Includes temperature effects (evapotranspiration) - interesting but at GWSC's scale of assessment how we can ensure we calculate accurate ET (adding another uncertainty)
- More sensitive to climate change than SPI
- Better for agricultural and hydrological applications
- Same interpretation as SPI

Formula: Uses climatic water balance ($P - PET$)

3. Palmer Drought Severity Index (PDSI) Strengths:

Comprehensive - uses precipitation, temperature, soil moisture - again it would be challenging or adding more uncertainty at the scale GWSC Self-calibrating version (sc-PDSI) works across different climates

Weaknesses:

Complex calculation Fixed timescale (not flexible like SPI) - this is another reason that it won't align with GWSC's needs

Decisions process

We are likely looking for a meteorological index. It appears that something similar to the Standardized Precipitation Index (SPI) may be appropriate. To understand the limitations of SPI and align it with GWSC's needs, we first examine how standard SPI values are calculated and then consider what modifications or extensions may be required.

SPI Calculation Process

Step 1: Collect Precipitation Data

We need at least 30 years of monthly precipitation data for the location.

Step 2: Choose Time Scale

Decide the accumulation period (1, 3, 6, 12, or 24 months).

For example:

SPI-3: Sum precipitation over 3-month periods SPI-6: Sum precipitation over 6-month periods

Step 3: Aggregate Precipitation

For each month, sum the precipitation over your chosen time scale. Example for SPI-3 at March 2026:

Sum = Jan 2026 + Feb 2026 + Mar 2026

Step 4: Fit Probability Distribution

Fit the aggregated precipitation data to a Gamma distribution (most common) or other distribution. The Gamma distribution is defined by two parameters:

Step 5: Calculate Cumulative Probability

Transform the gamma distribution to cumulative probability.

This gives you the probability of getting precipitation x.

Step 6: Transform to Standard Normal Distribution

Convert the cumulative probability to standard normal distribution (mean=0, std=1):

```

# Load library
library(SPEI)

# Package SPEI (1.8.1) loaded [try SPEINews()].
# Create sample data (30 years monthly)
set.seed(123)
n_years <- 30
n_months <- n_years * 12

precip <- rgamma(n_months, shape = 2.5, scale = 20)

# Create time series
precip_ts <- ts(precip, start = c(1990, 1), frequency = 12)

# Calculate SPI-3
spi_result <- spi(precip_ts, scale = 3, na.rm = TRUE)

[1] "Calculating the Standardized Precipitation Evapotranspiration Index (SPEI) at a time s

# View structure
print(class(spi_result))

[1] "spei"

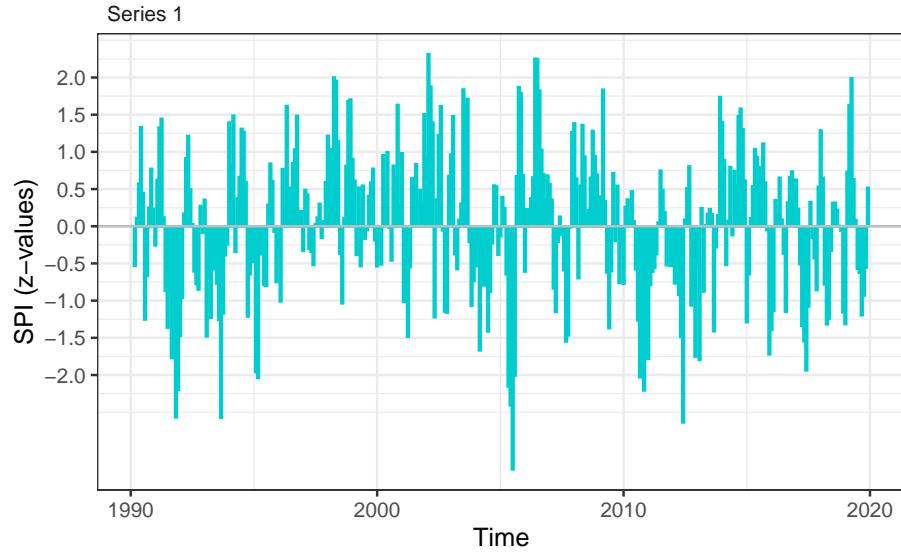
print(names(spi_result))

[1] "call"           "info"          "fitted"         "coefficients" "scale"
[6] "kernel"         "distribution"  "fit"            "na.action"

# Try plotting
plot(spi_result, main = "SPI-3")

Warning: Removed 358 rows containing missing values or values outside the scale range
('geom_point()').

```



Notes to later use

Limitation with relying on Precipitation - flash drought - elevated temperatures and heightened evaporative demand

drought is not solely a result of natural variability but also a product of human decisions and actions - i.e. The increasing demand for water due to population growth, urbanization, and agricultural expansion has heightened the risk of water scarcity, even in regions not traditionally prone to drought

More earliest definition of drought:

- 21-day period with total rainfall less than 30% of the regional average, relying solely on precipitation thresholds to characterize prolonged dryness