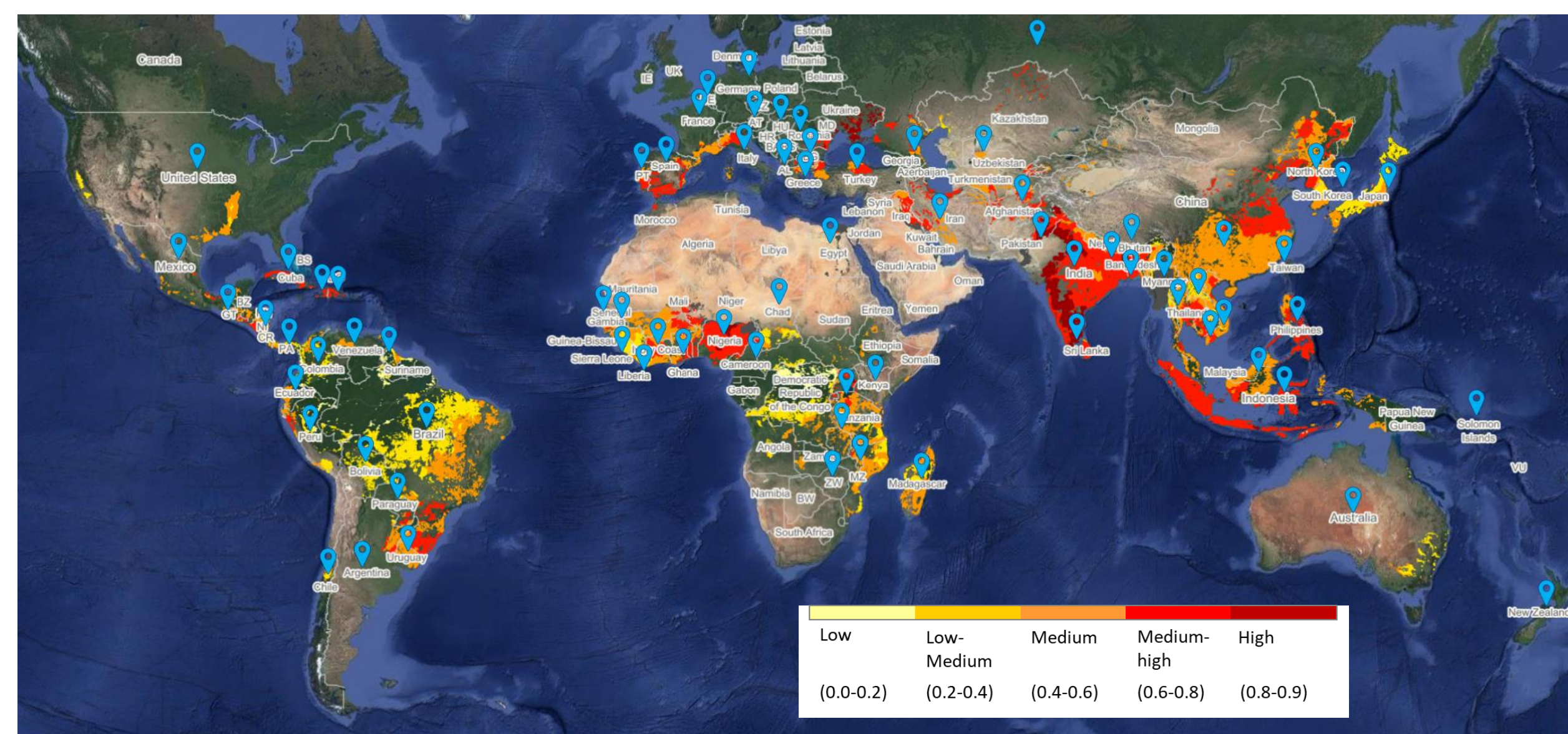


## Background

- Asian rice (*Oryza sativa*) is the primary calorie source for > 50% of the global population.
- Water scarcity is a major challenge for current and future rice production (Fig. 1).
- We will use key trait genetic architectures and existing genetic and varietal resources of rice to predict high-yielding genotypes under drought stress.



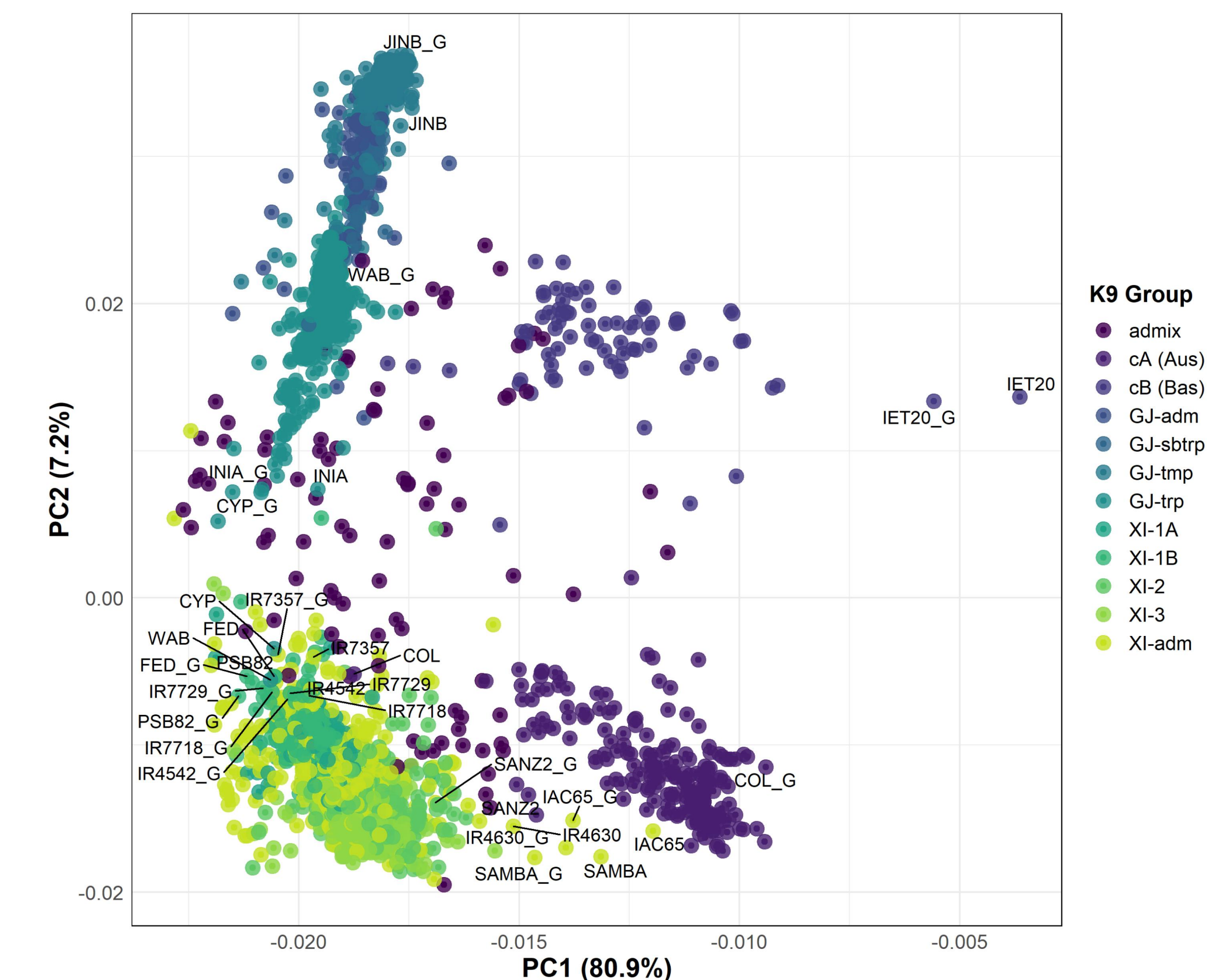
**Figure 1.** Drought risk in irrigated and rainfed rice production areas worldwide. Blue pointers indicate country of origin for the 3,024 rice accessions sequenced by [2]. Created with the Aqueduct Food tool from the Water Resource Institute (USA).

## The Plan

- We evaluated growth, phenology, and yield component traits under drought and well-watered conditions across multiple growing seasons in the 16-parent diverse Global Multi-parent Advanced Generation Inter-Cross (MAGIC) population developed by inter-crossing elite varieties from *indica* and *japonica* subspecies [1].
- We identified significant marker-trait associations for these traits across treatments and seasons, finding that **many loci exhibit dynamic reaction norms of effect size with respect to environment (Fig. 2) and developmental time.**
- We will use these GWAs to predict trait **polygenic scores** [2] in a diverse re-sequenced collection of **3,024 rice accessions** [3] originating from **89 countries** (Fig. 1).
- Shifts in allele frequencies at candidate loci reflect historical and geographic patterns of selection for key traits. **Our aim is to identify both patterns of adaptation in drought response and potential donors of drought tolerance alleles in rice.**

- We aligned genotyping-by-sequencing (GBS) derived variants from the Global MAGIC population against the two best subspecies genomes: (i) *japonica* Nipponbare (7,793 loci); (ii) *indica* 93-11 (7,607 loci).
- We retrieved whole-genome resequencing data from the 3K rice panel [2] and filtered for these variants.
- The population structure of these variants is similar to that from the whole genome data [2], although they primarily differentiate the aus/basmati subpopulations from *indica/japonica*, and only secondarily the two major subspecies (Fig. 3).**
- We observe genotypic discordance between the GBS and 3K data for three of 16 MAGIC parents in both datasets (Fig. 3), likely from low quality 3K data or different DNA source.
- After filtering for high-quality genotypes and imputing missing data, we will estimate trait polygenic scores following the approach of Josephs et al. [4] and examine how these vary with geography.**

**Figure 2.** Effect sizes from genome-wide associations for: (top; left-to-right) plant height, grain yield, harvest index; (bottom; left-to-right) mesocotyl length, straw biomass, and axial root number. Lines connect the same genetic variant across environments. Only genetic variants with a significant association to trait variance in at least one environment are shown. DS14s = dry season 2014 – seedling stage drought stress; DS15s = dry season 2015 – seedling stage drought stress; DS15w = dry season 2015 – well-watered conditions.



**Figure 3.** Genetic variation among the 3K rice accessions and the 16 Global MAGIC parents along the first two Principal Component axes of GBS-derived Global MAGIC variants. Labels indicate the parental lines both from the MAGIC data and the 3K WGS data (labels with the suffix “\_G”). Samples are colored based on population structure as in [2]. cA = Aus type accessions; cB = Basmati; GJ = japonica; XI = indica, with various subpopulations; adm = admixed; sbtrp = subtropical; tmp = temperate; trp = tropical.

