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Karma Mroueh
Shantanu Samant

Drought in California





April 27, 2021



July 22, 2021



Agenda



Context



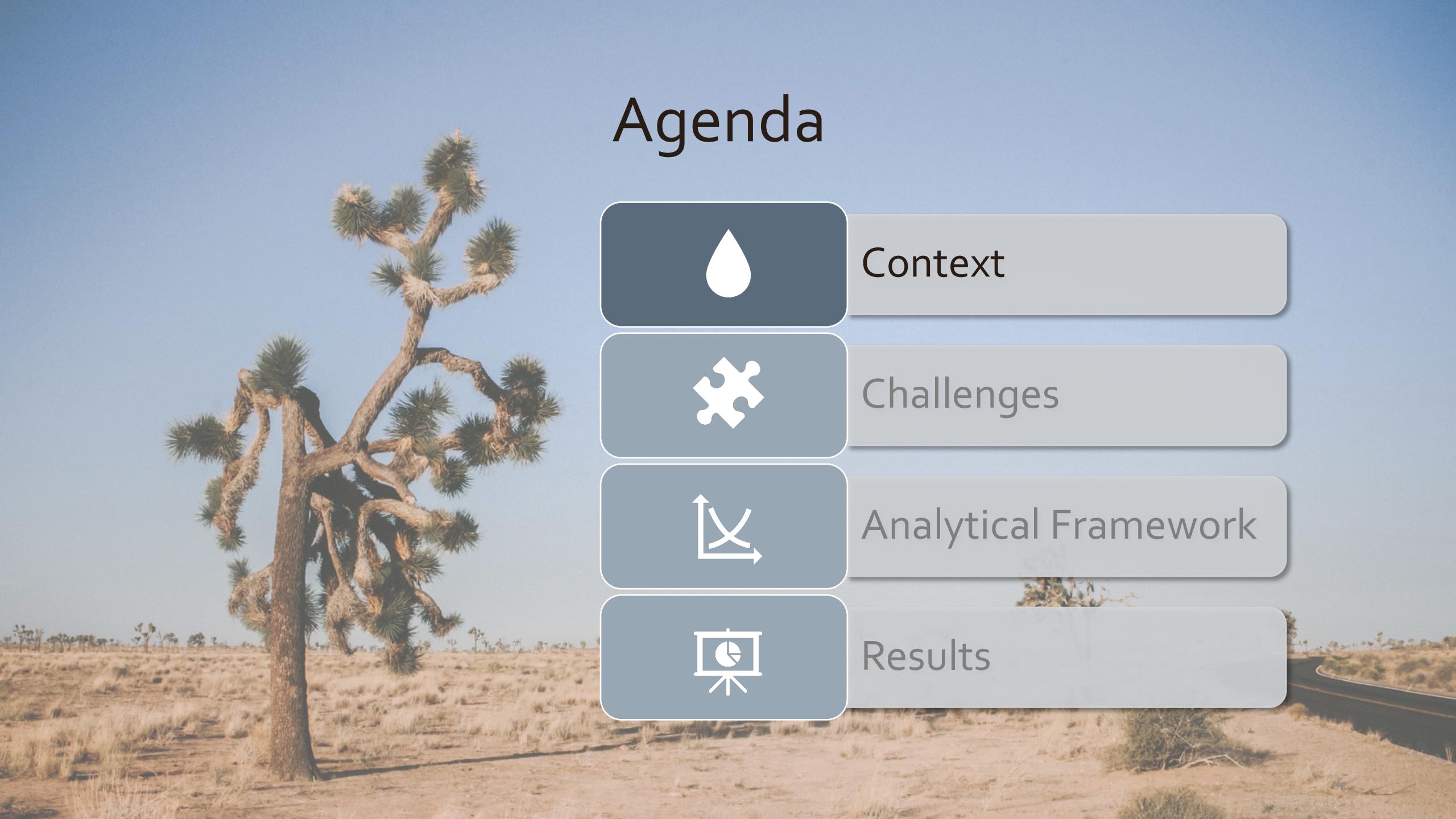
Challenges



Analytical Framework



Results



Agenda



Context



Challenges



Analytical Framework



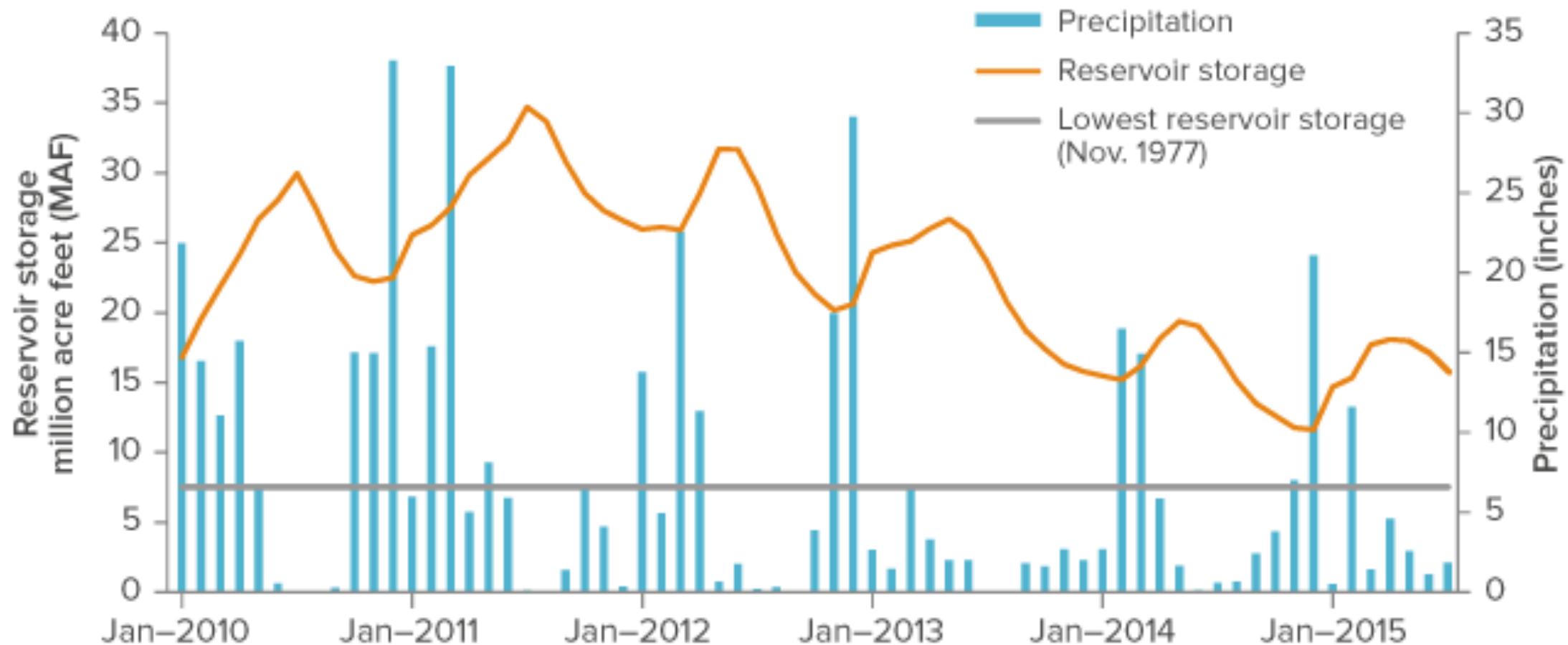
Results

Inequality and the Drought in California

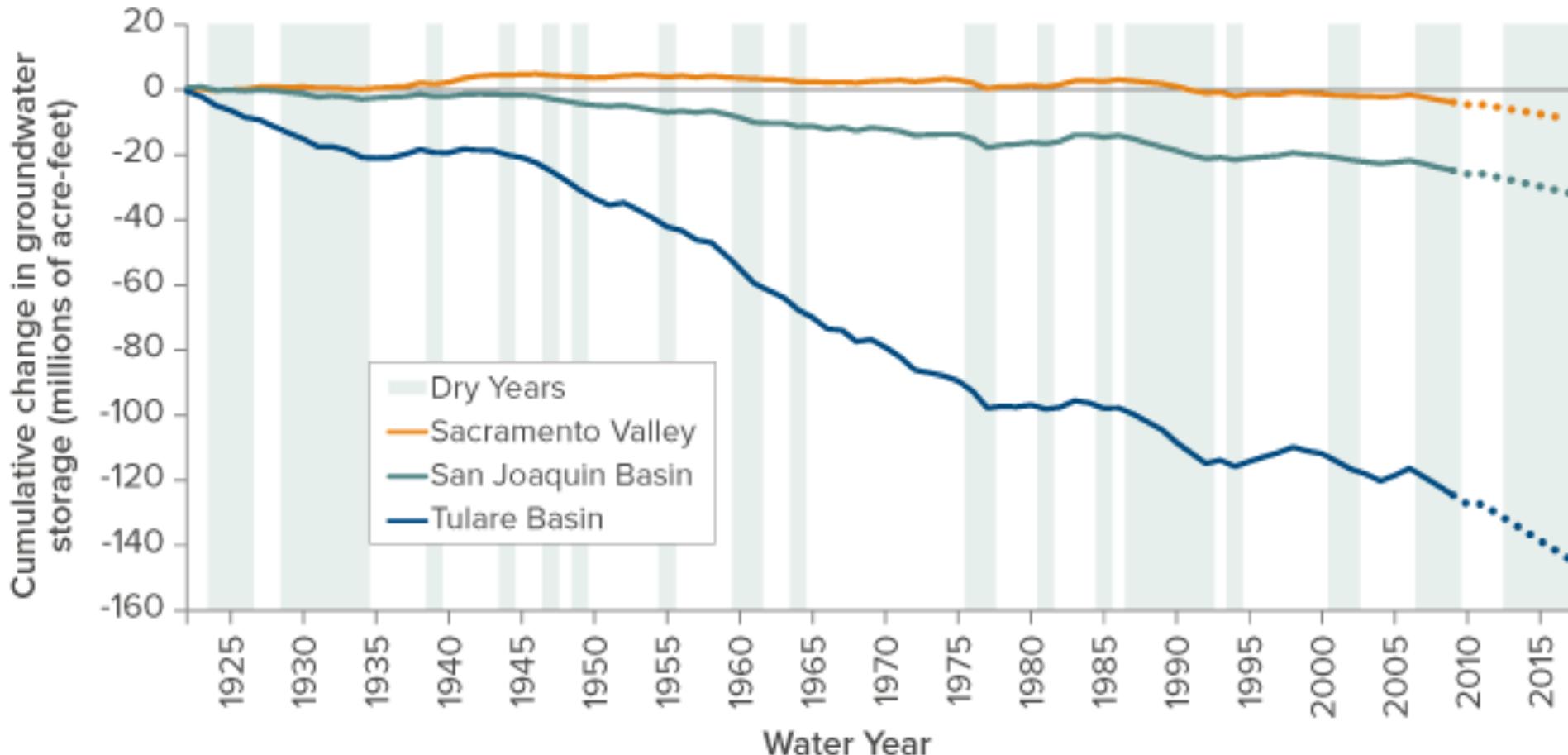
- Total direct and indirect costs of \$1.7 billion
- Loss of over 14,000 jobs
- Water prices at an all-time high of \$1,144.14 per acre-foot

As the drought persists, how can we mitigate its economic ramifications and equitably allocate scarce water supplies?

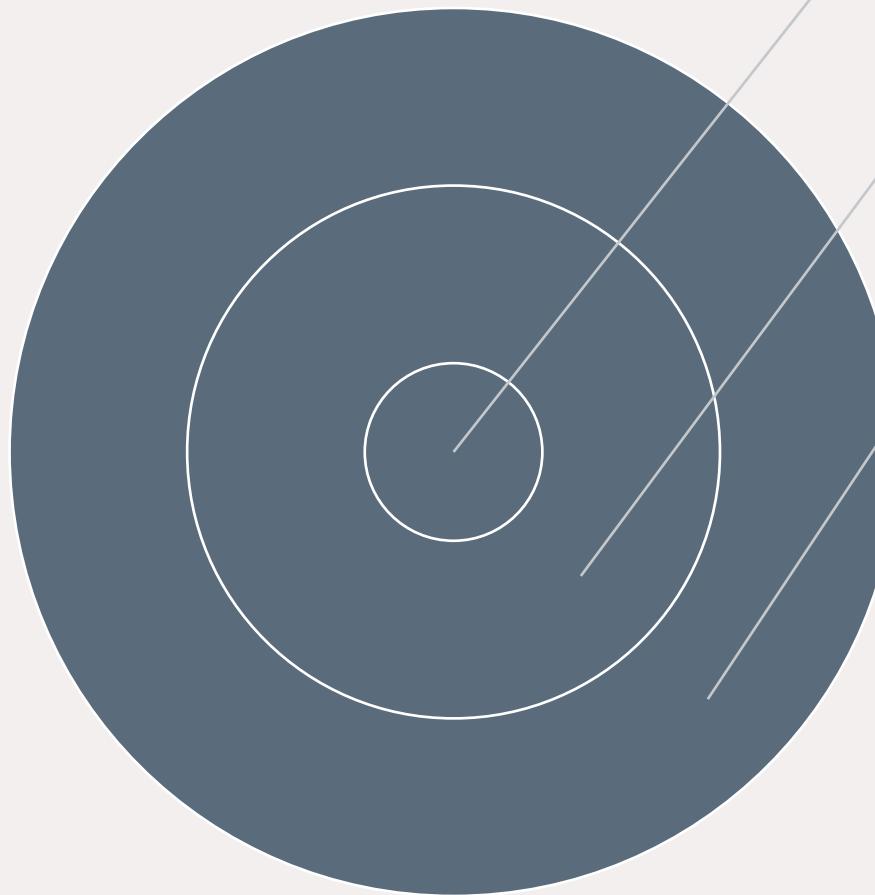
Decrease in reservoir storage



Decrease in groundwater storage



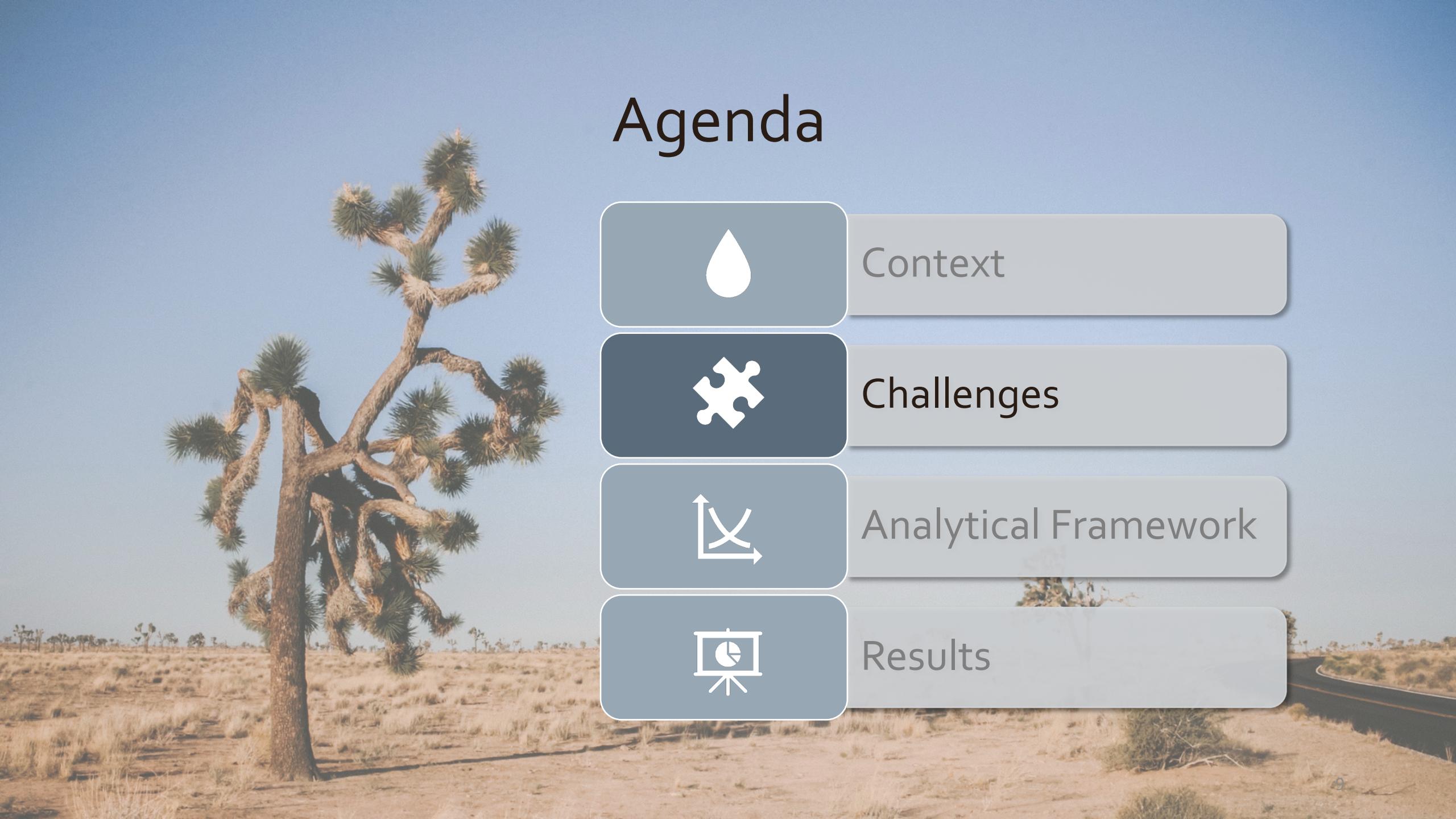
Our goal



Predict future
water production

Simulate consumer
demand

Determine best water
price policy



Agenda



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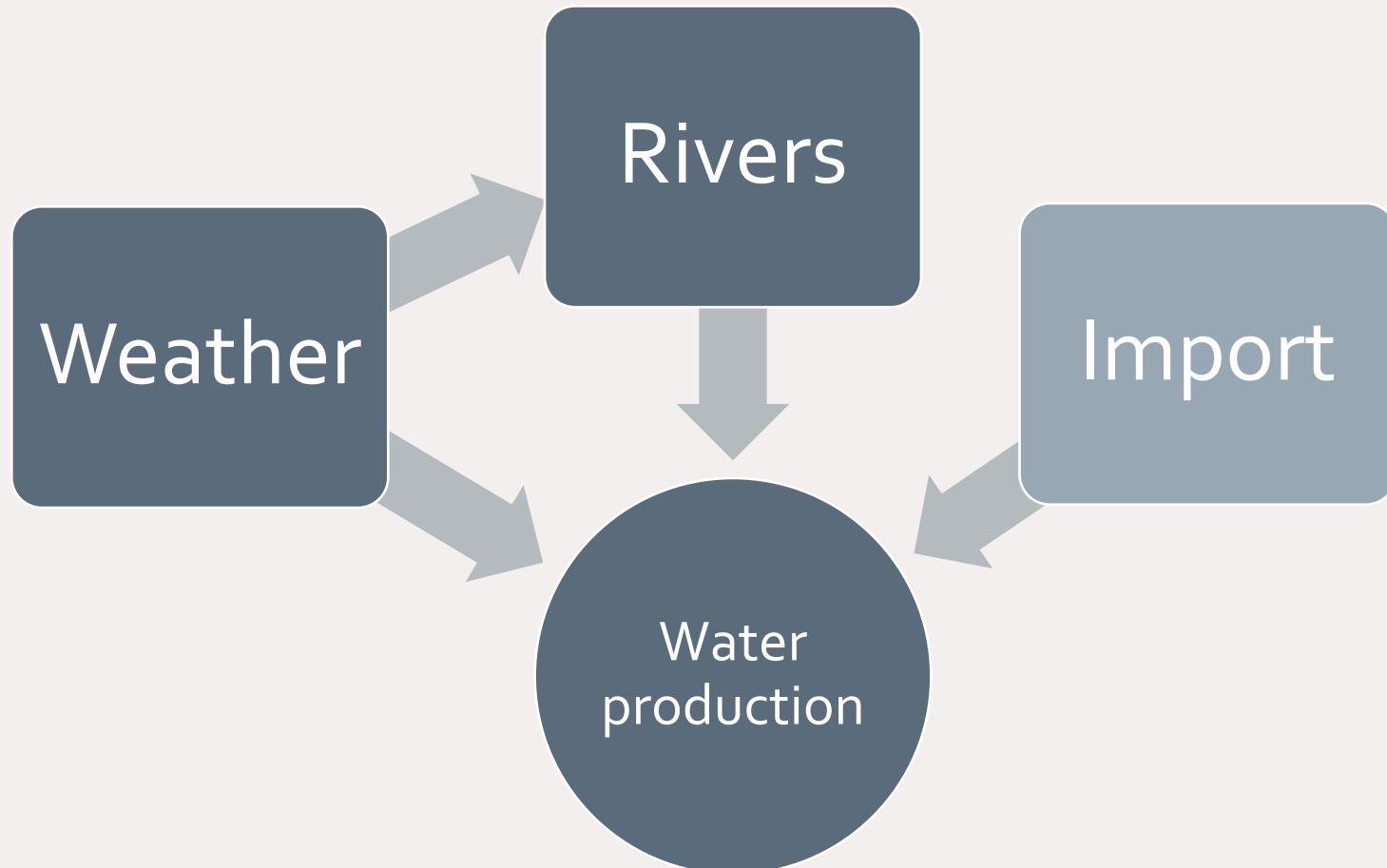
Results

Analytical challenges

1. Understanding the causal relationships behind water production
2. Combining prediction with simulation



Analytical challenges



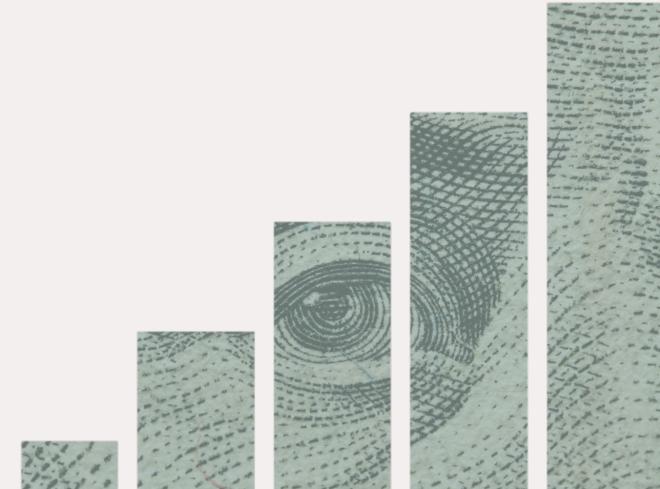
Data challenges

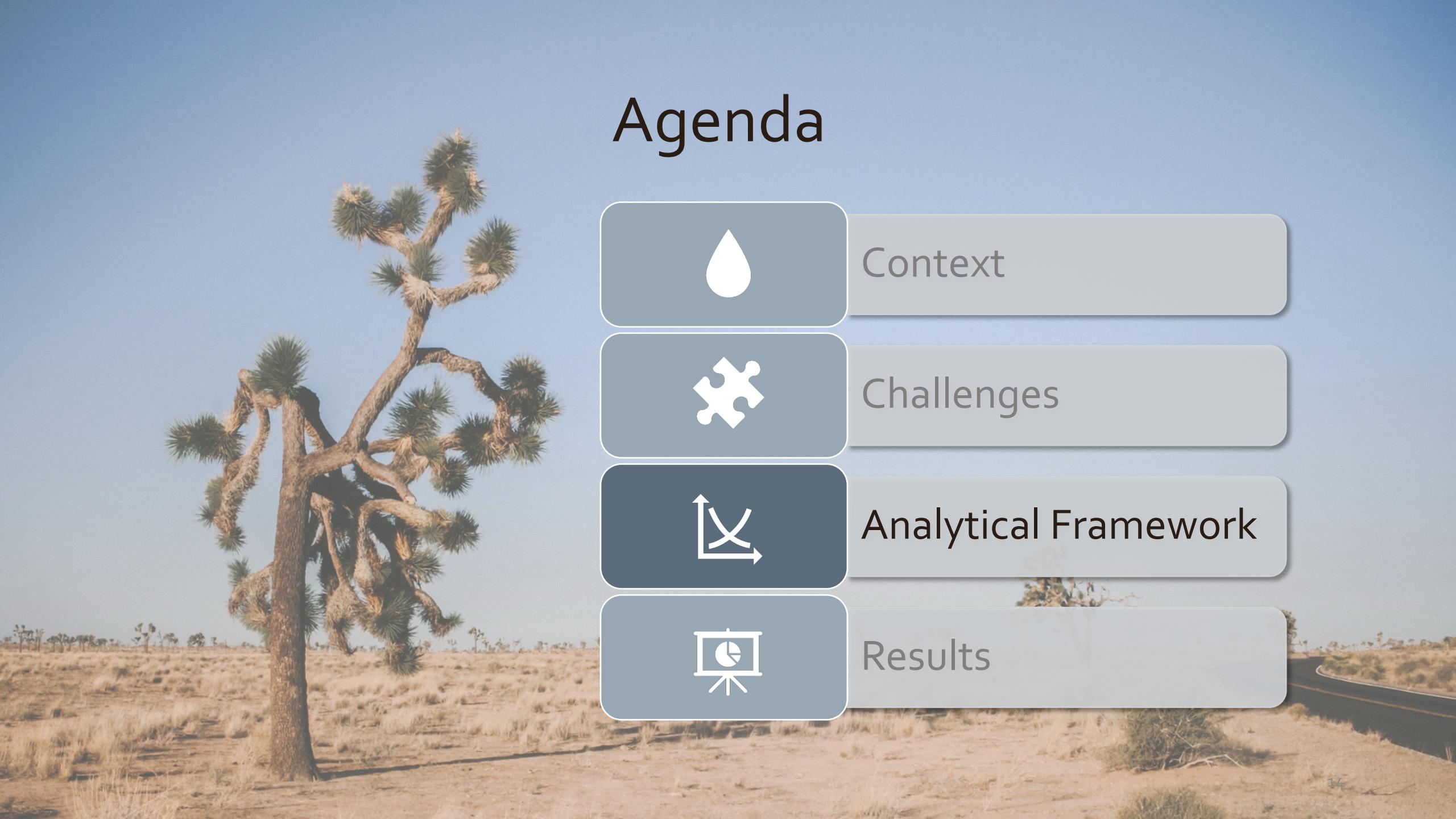
1. Finding variables that predict water production
2. Obtaining and extracting data from weather services
3. Dealing with time-series data



Policy challenges

1. Defining set of possible policy responses
 - Focusing on price changes in a two-tier system
2. Mathematically formulating possible policies





Agenda



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Challenges

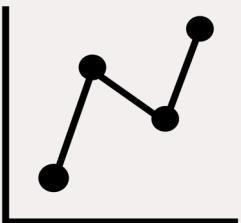


Analytical Framework

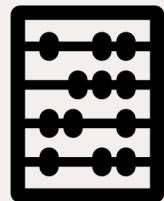


Results

Pipeline



Predicting water production



Modelling consumptions and price levels



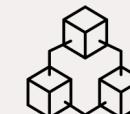
Simulating policy interventions



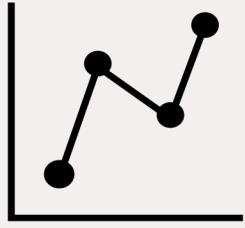
Machine Learning



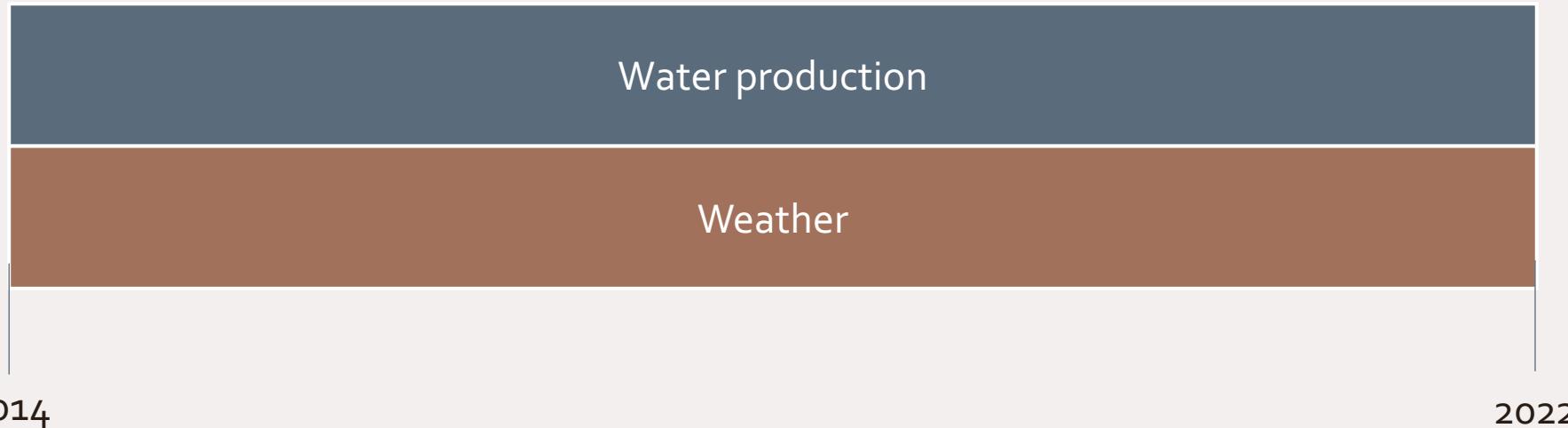
Historical data and behavior modelling

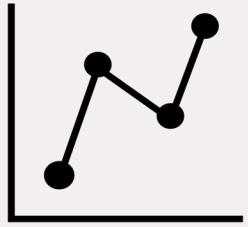


Simulation



Predicting water production



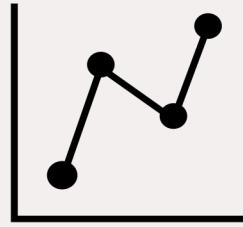


Predicting water production

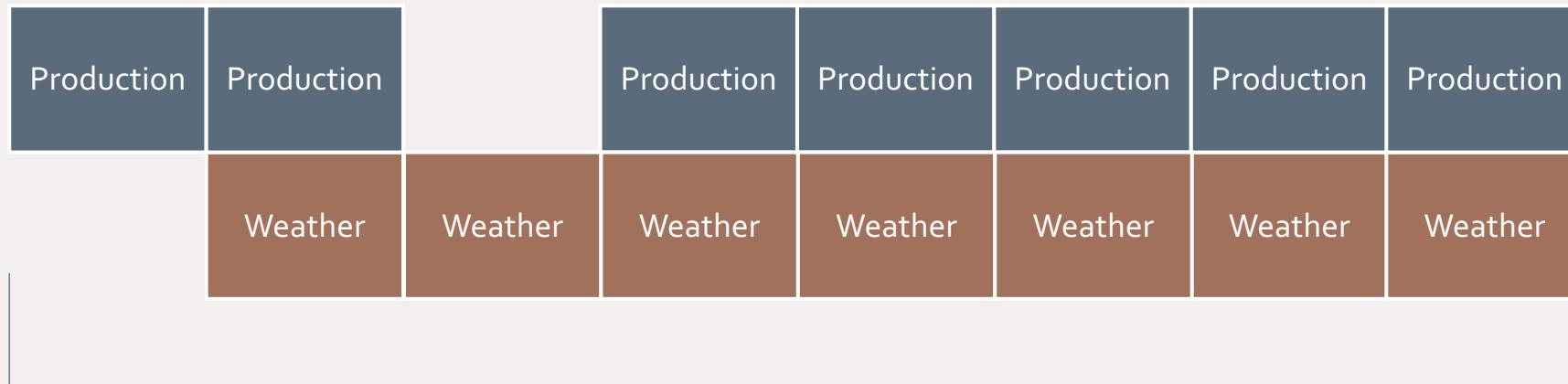
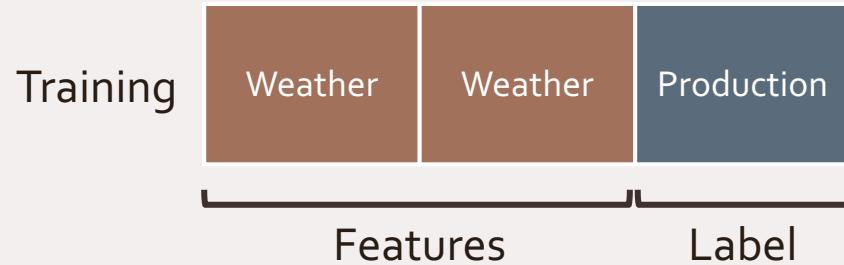
Production							
Weather							

2014

2022

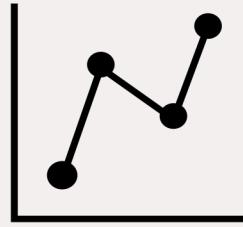


Predicting water production

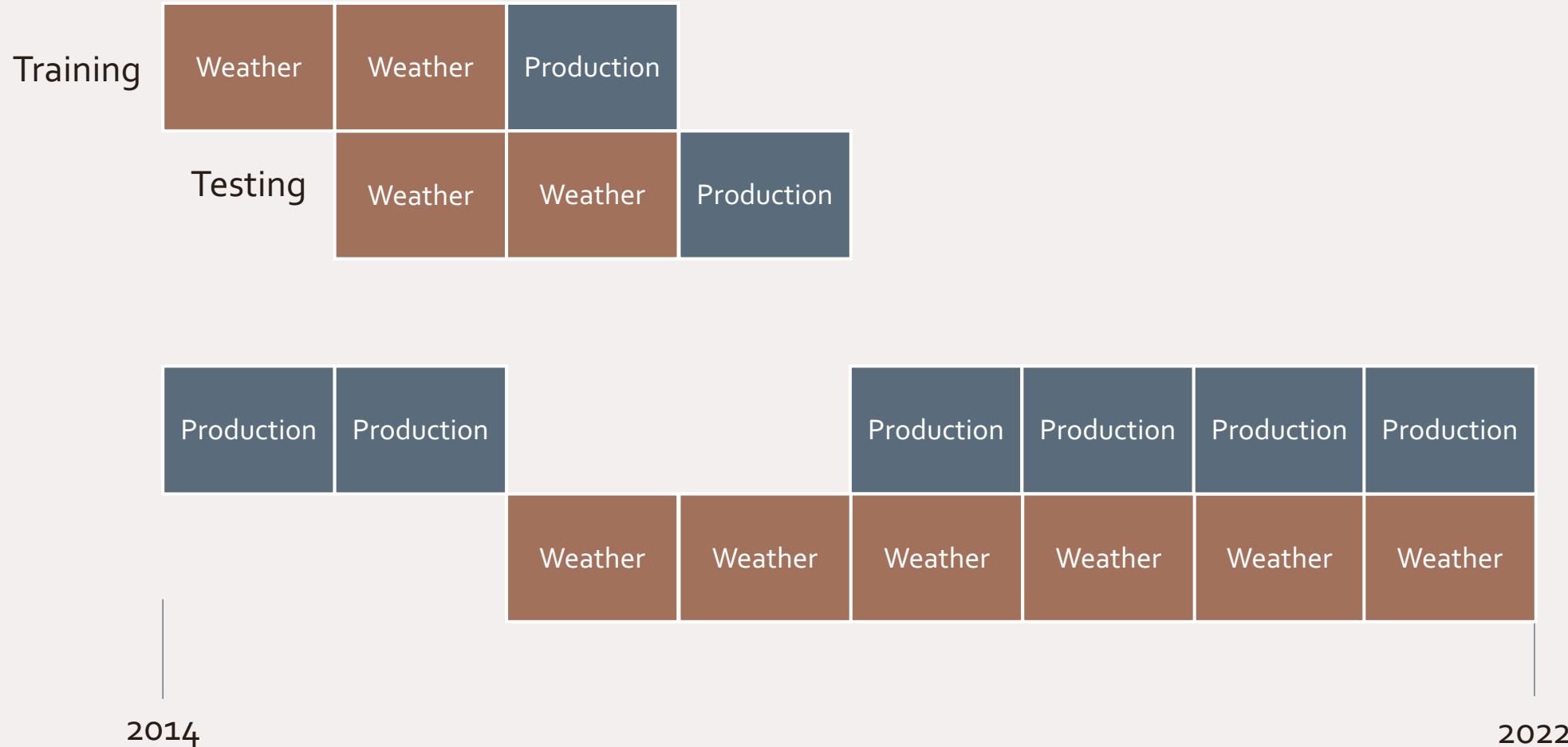


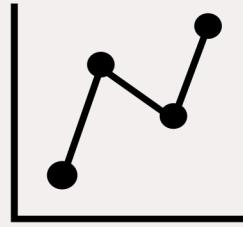
2014

2022

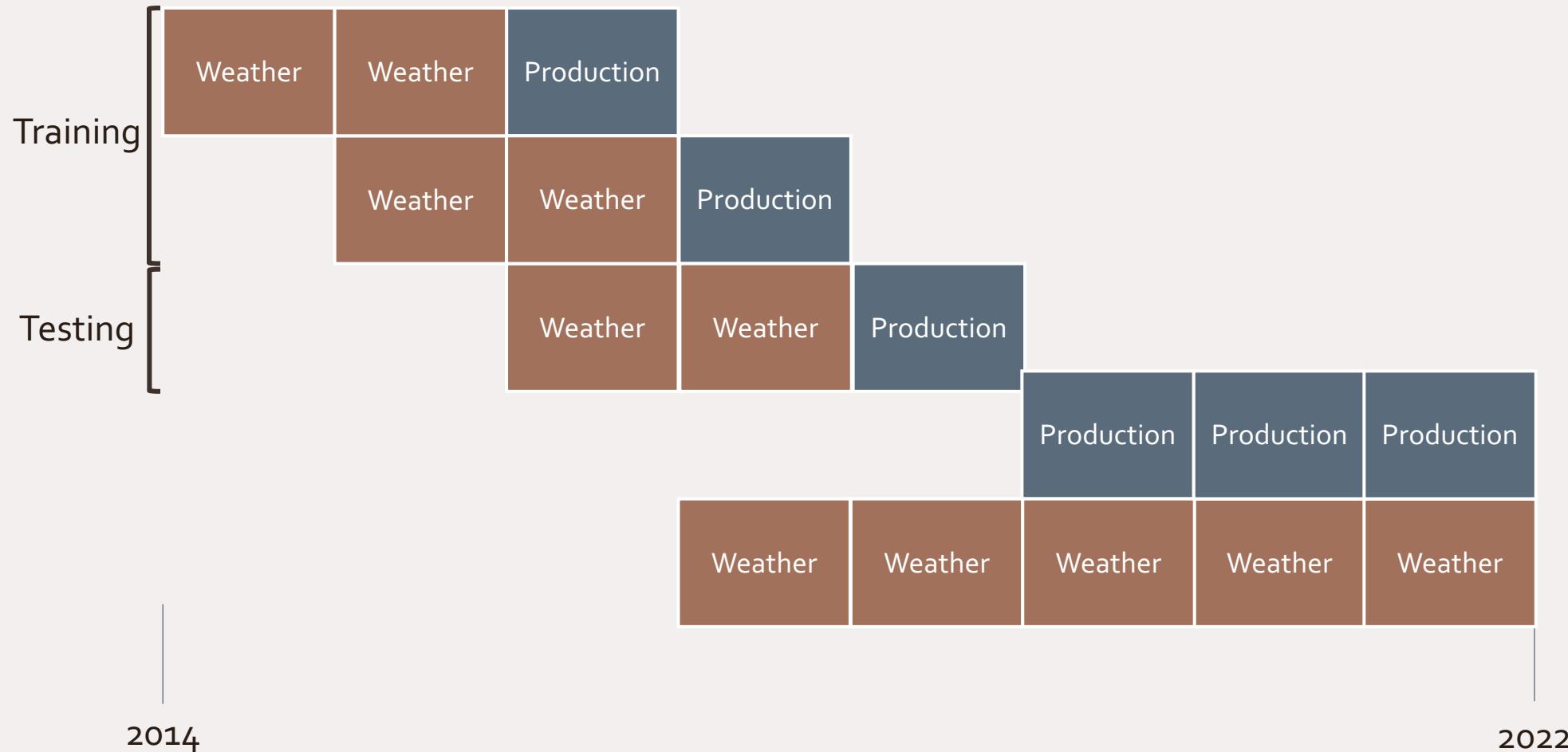


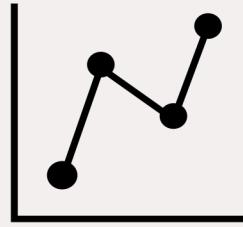
Predicting water production



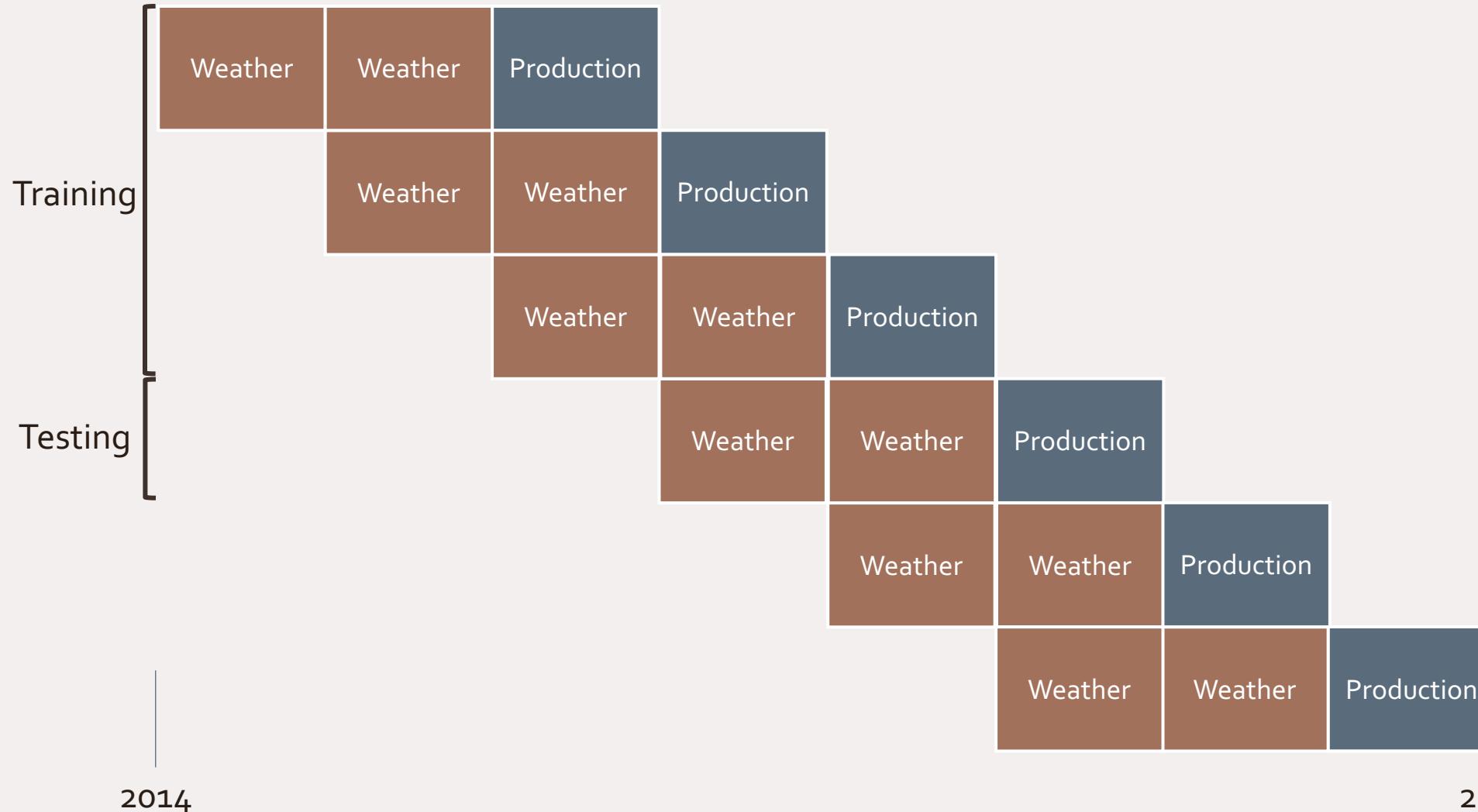


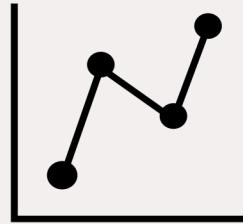
Predicting water production



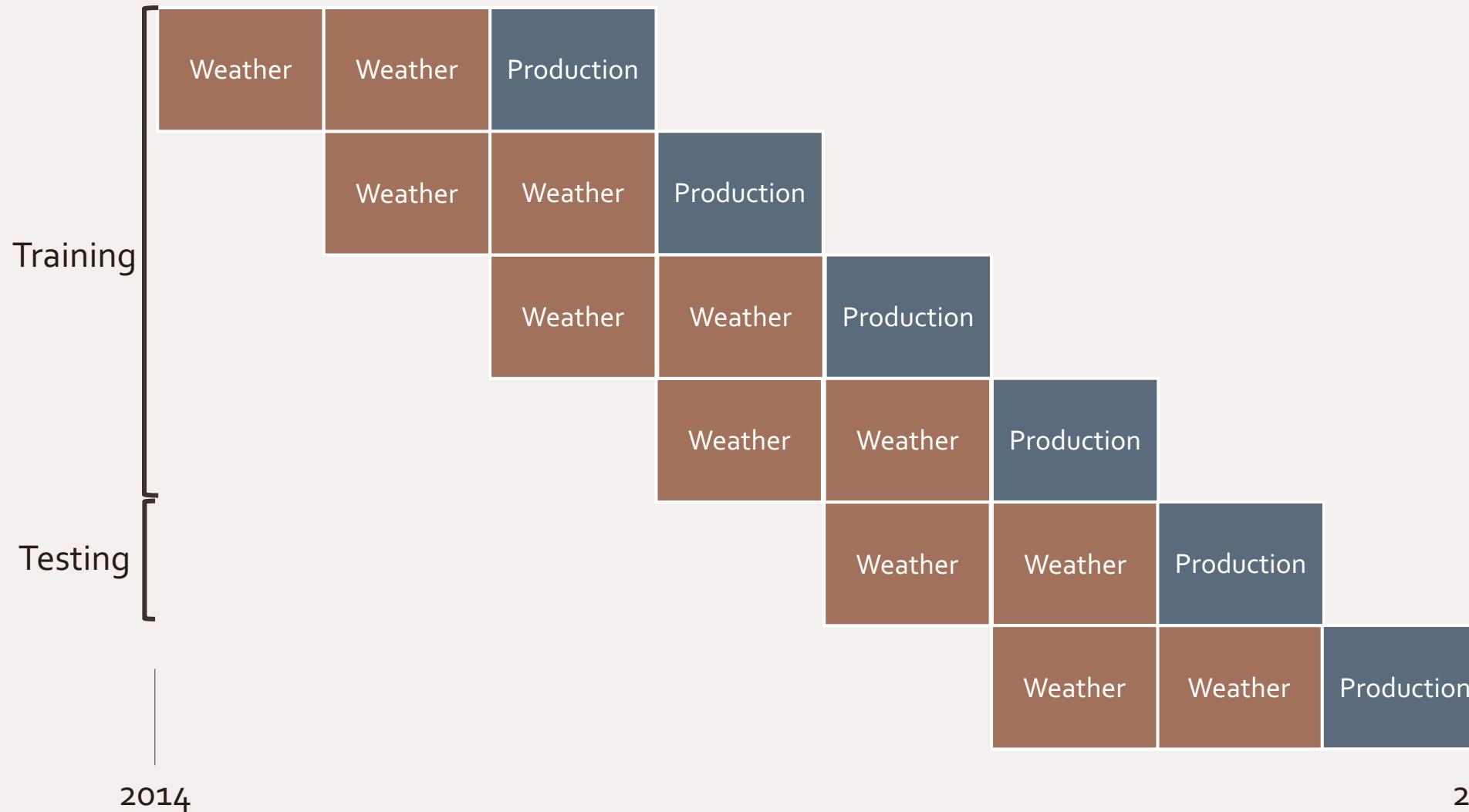


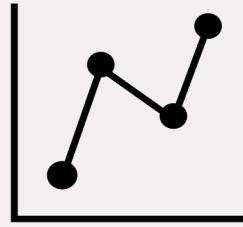
Predicting water production



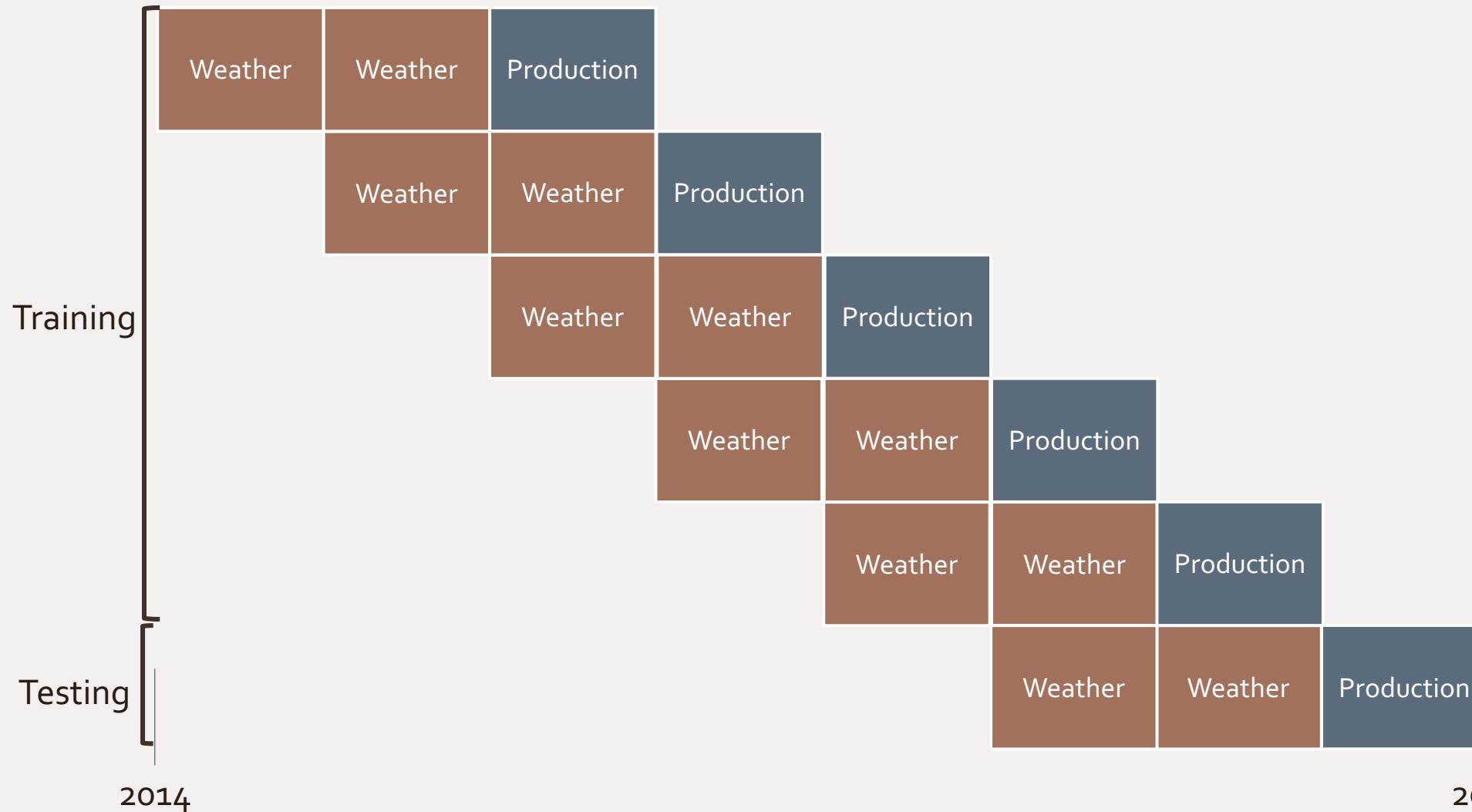


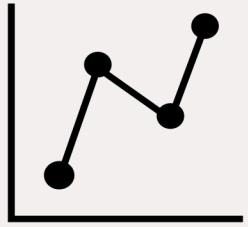
Predicting water production



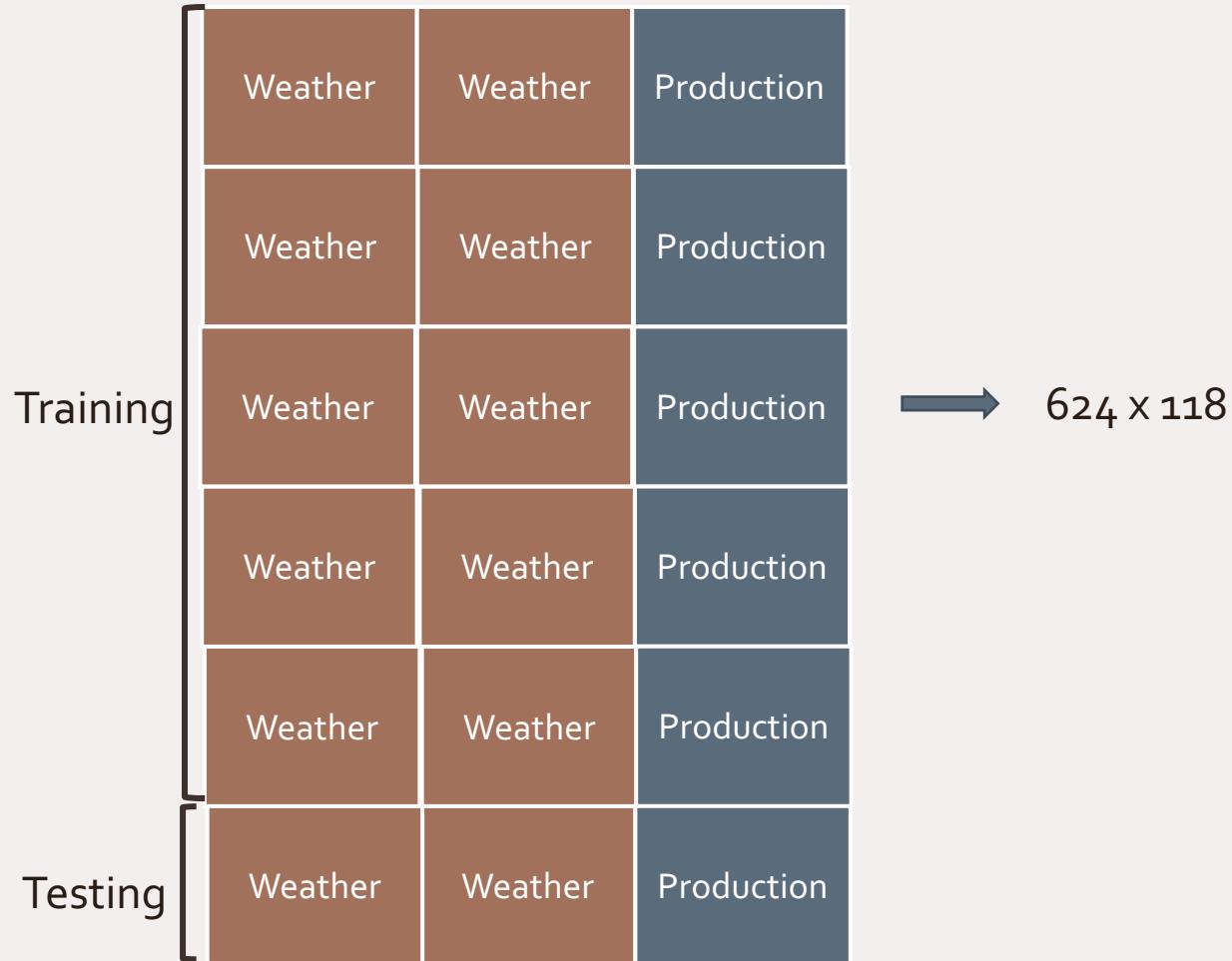


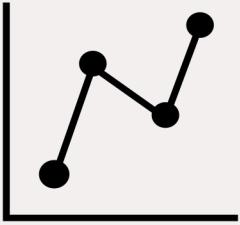
Predicting water production



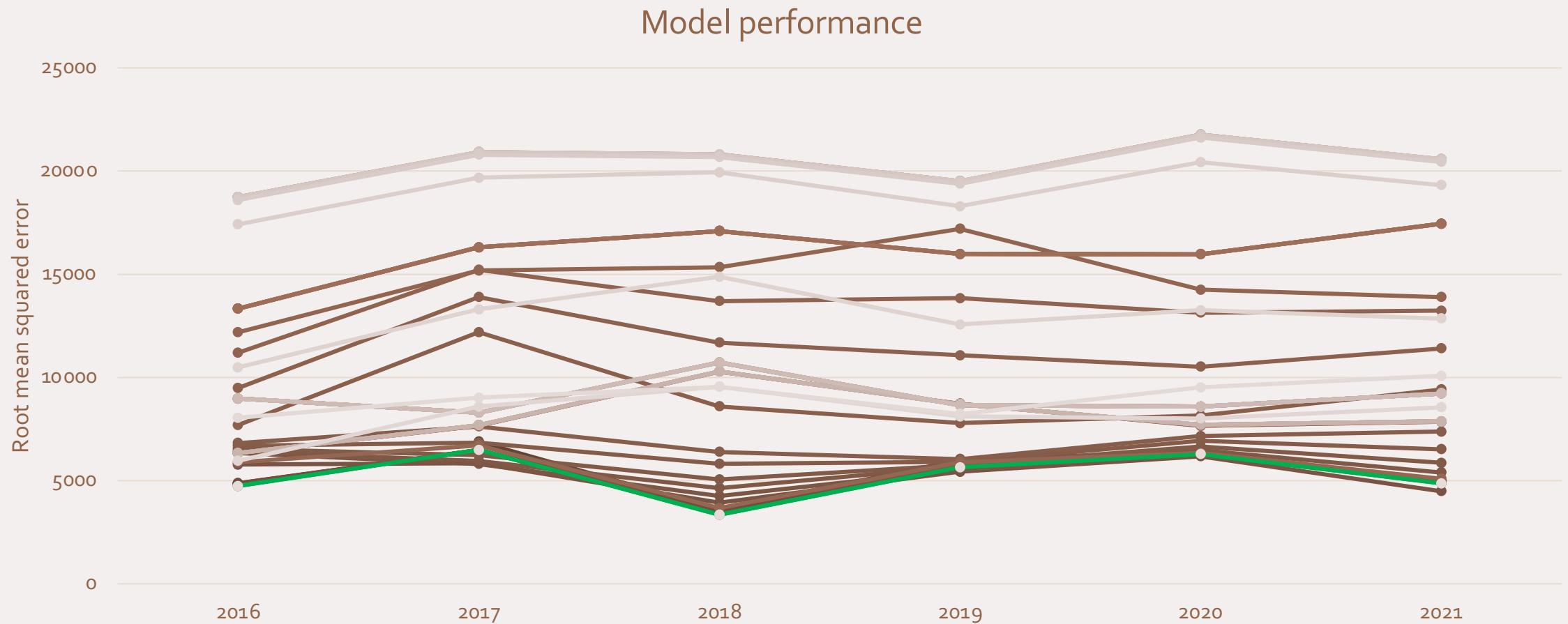


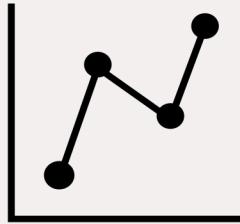
Predicting water production





Predicting water production



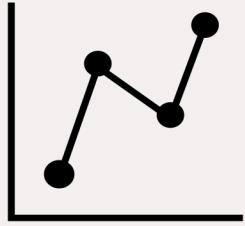


Predicting water production

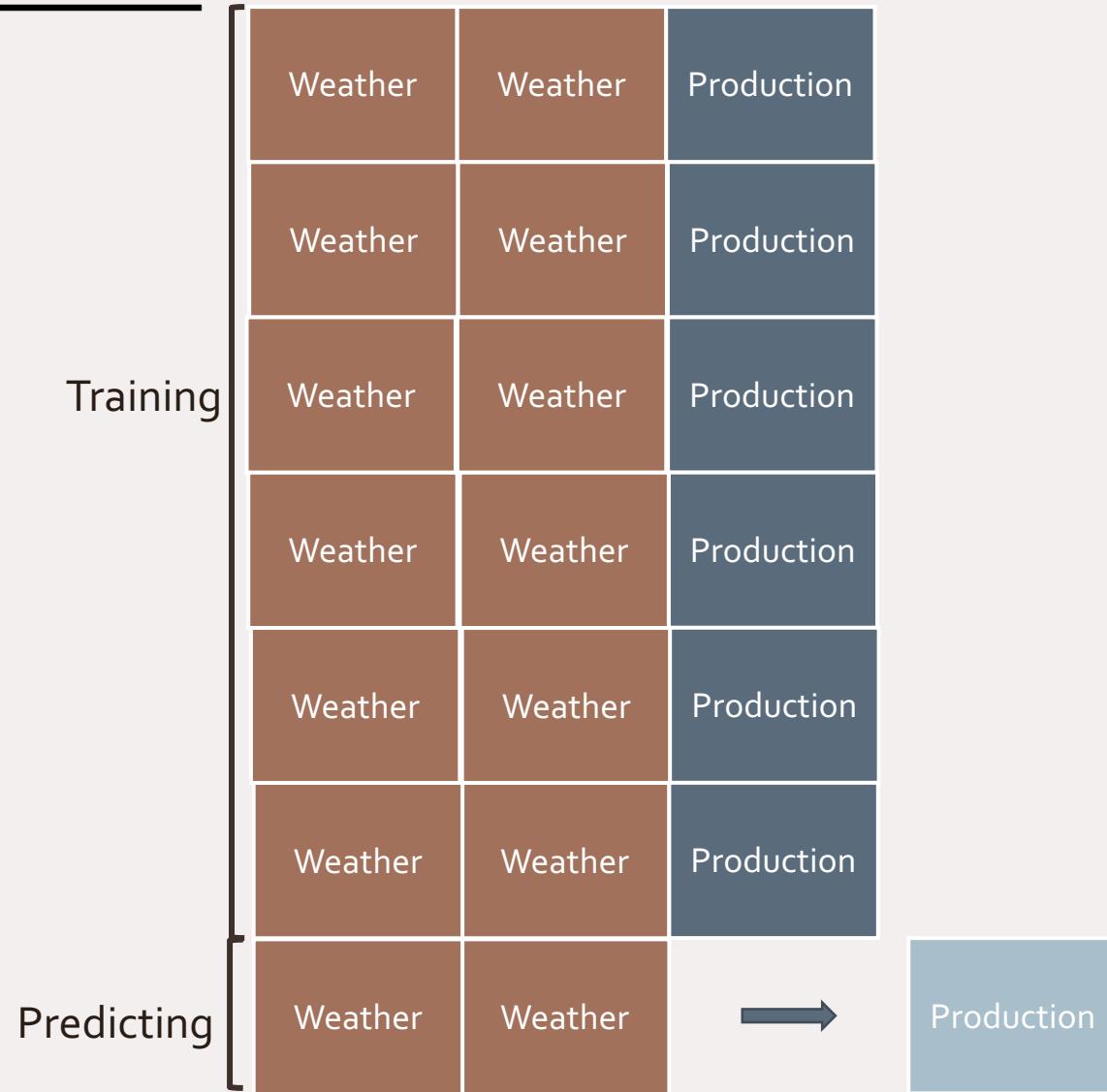
	Weather	Weather	Production
Training	Weather	Weather	Production
	Weather	Weather	Production
Testing	Weather	Weather	Production

Best model:

Lasso (regularized regression), alpha = 1

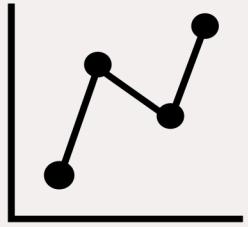


Predicting water production



Best model:

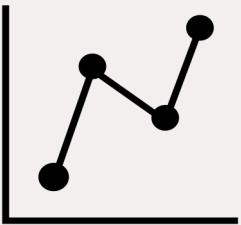
Lasso (regularized regression), alpha = 1



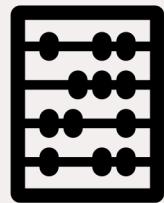
Predicting water production

Production = [45816.23
37276.80
52957.86
68582.92
83653.62
37606.94
58947.98
77416.53
33330.96 ...gallons per person for 2023
95174.87
59950.84
47003.18
87201.67
26546.84
43439.99
...
72873.59]

Pipeline



Predicting water production



Modelling consumptions and price levels



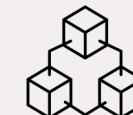
Simulating policy interventions



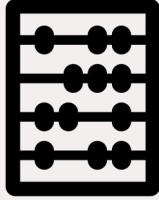
Machine Learning



Historical data and behavior modelling

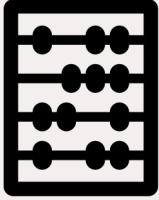


Simulation

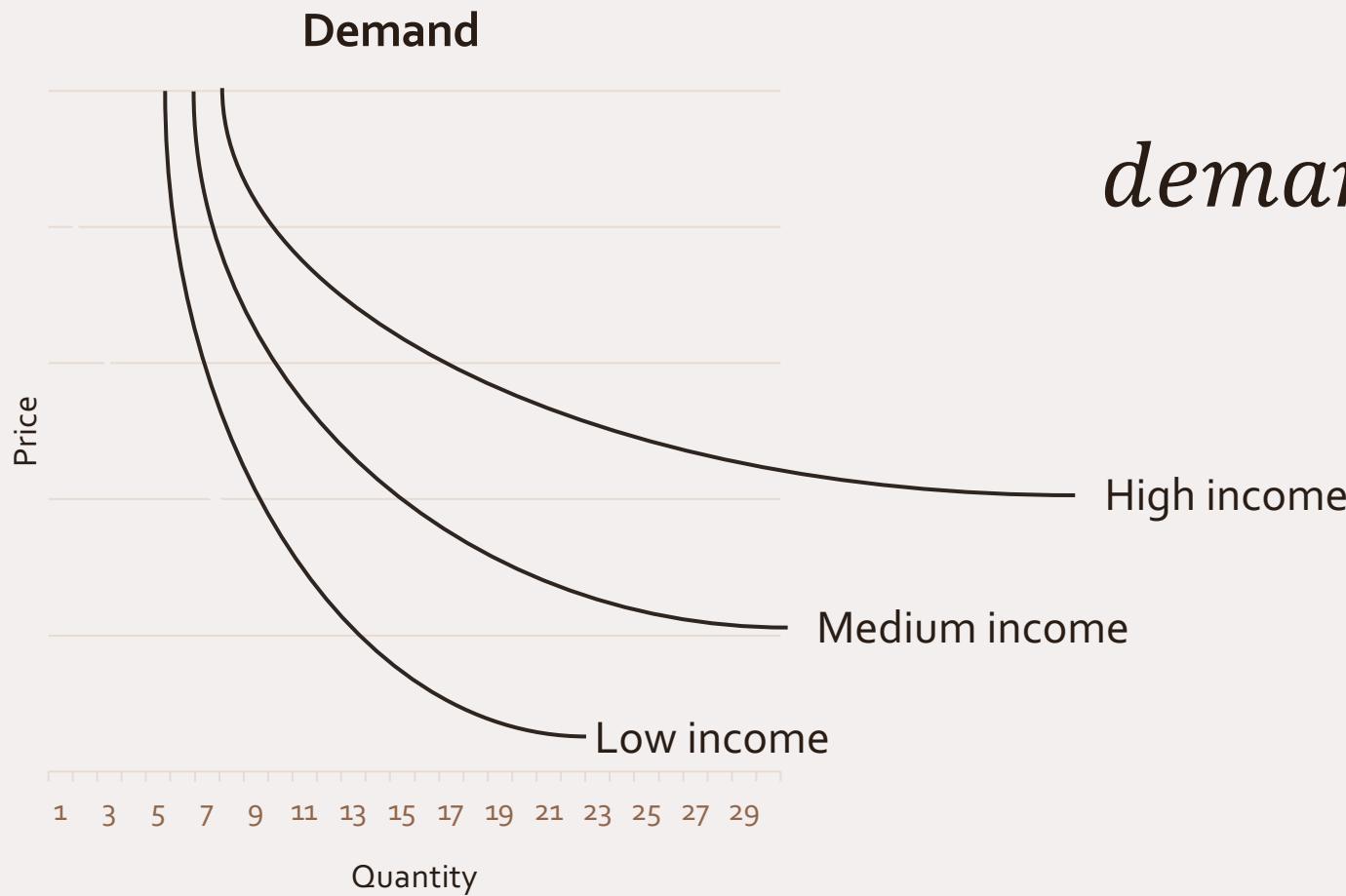


Modeling demand

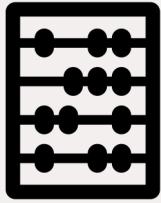
- Average demand elasticity = -0.435¹
- Average use = 100 gallons per person per day²
- Average current expenses = \$77 per month per household³
- Median income = \$77,800⁴ (on log-normal distribution)



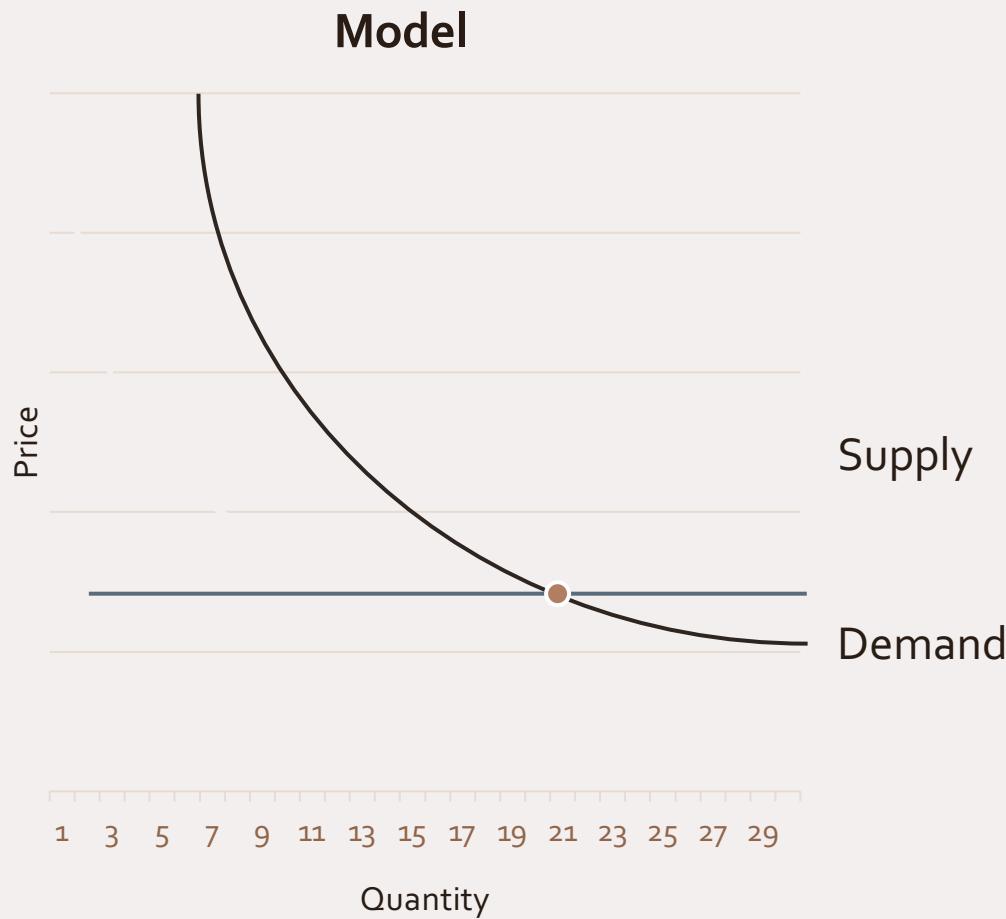
Modeling demand

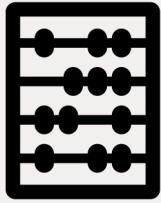


$$demand = q_1 * \left(\frac{price}{p_1} \right)^\varepsilon$$

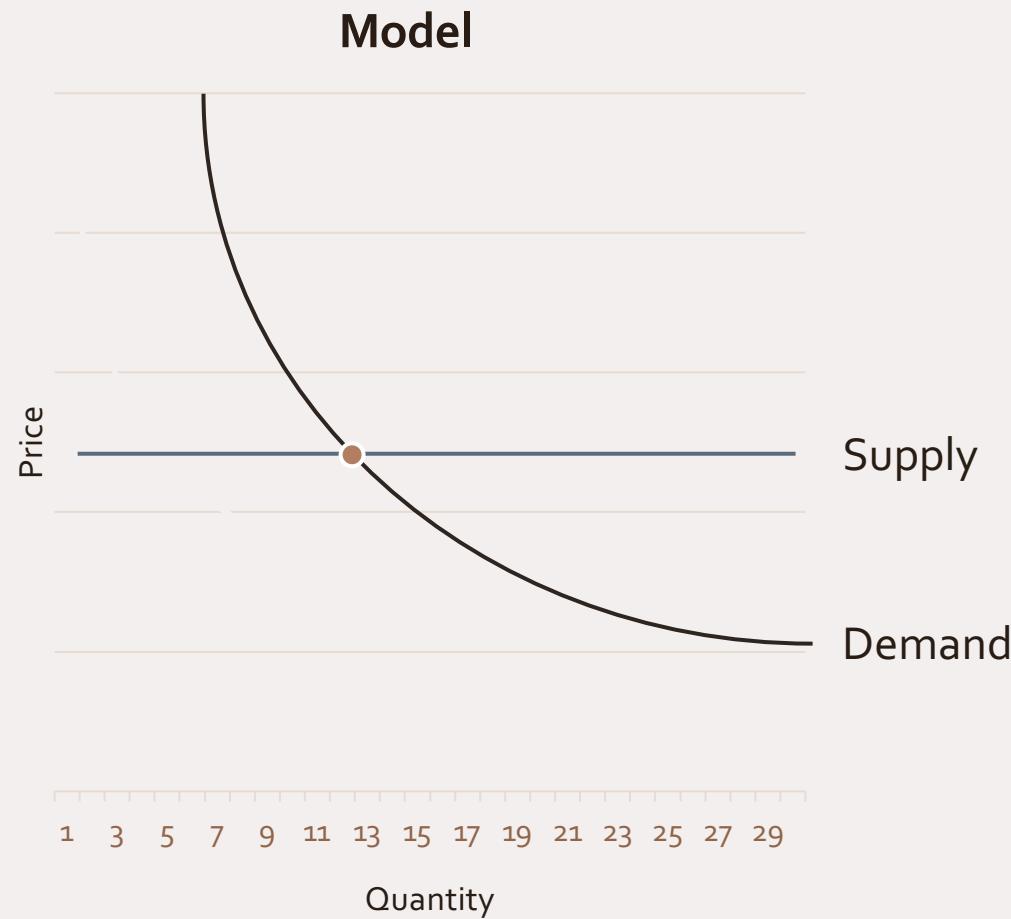


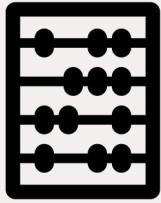
Modeling demand



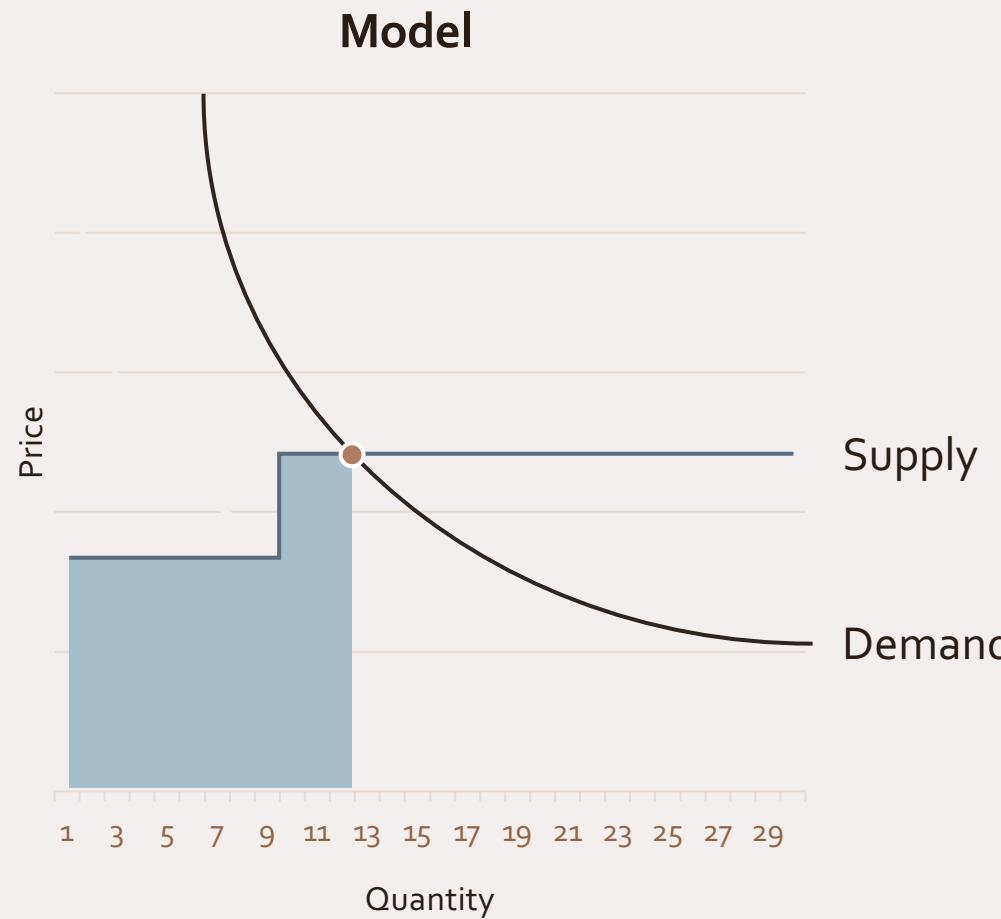


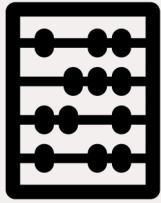
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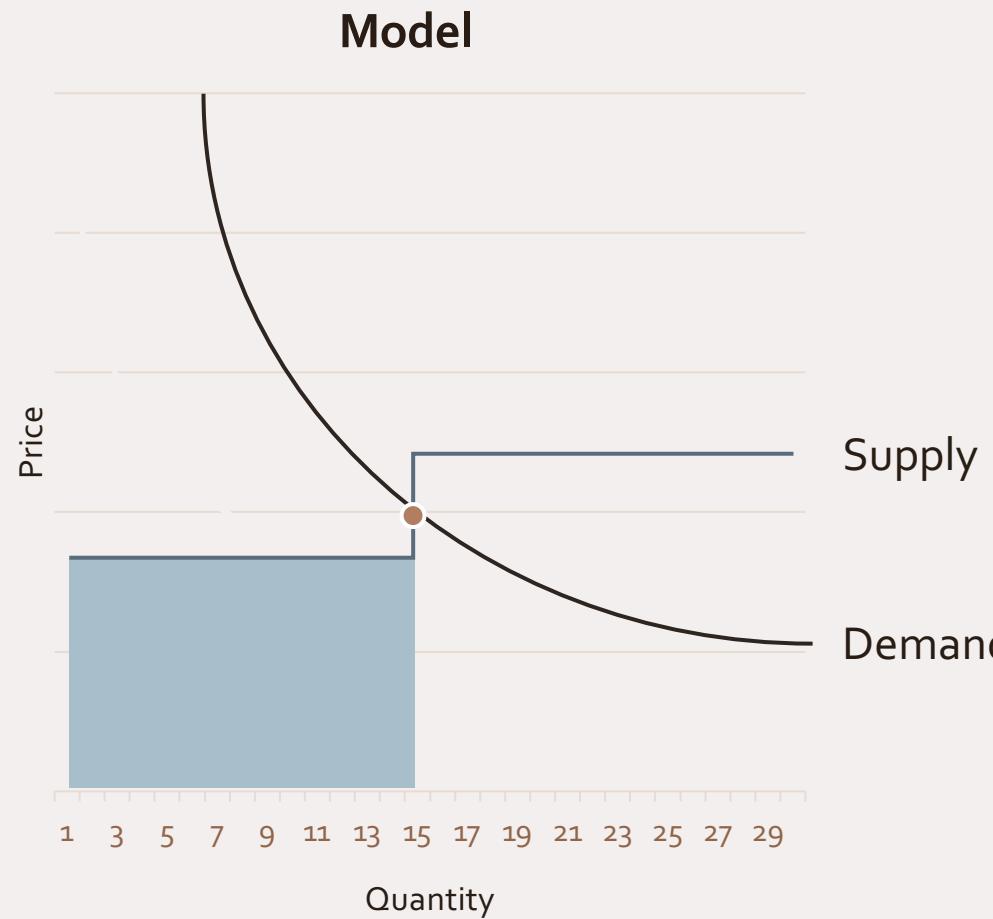


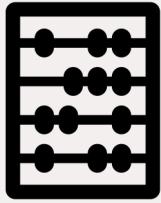
Modeling demand



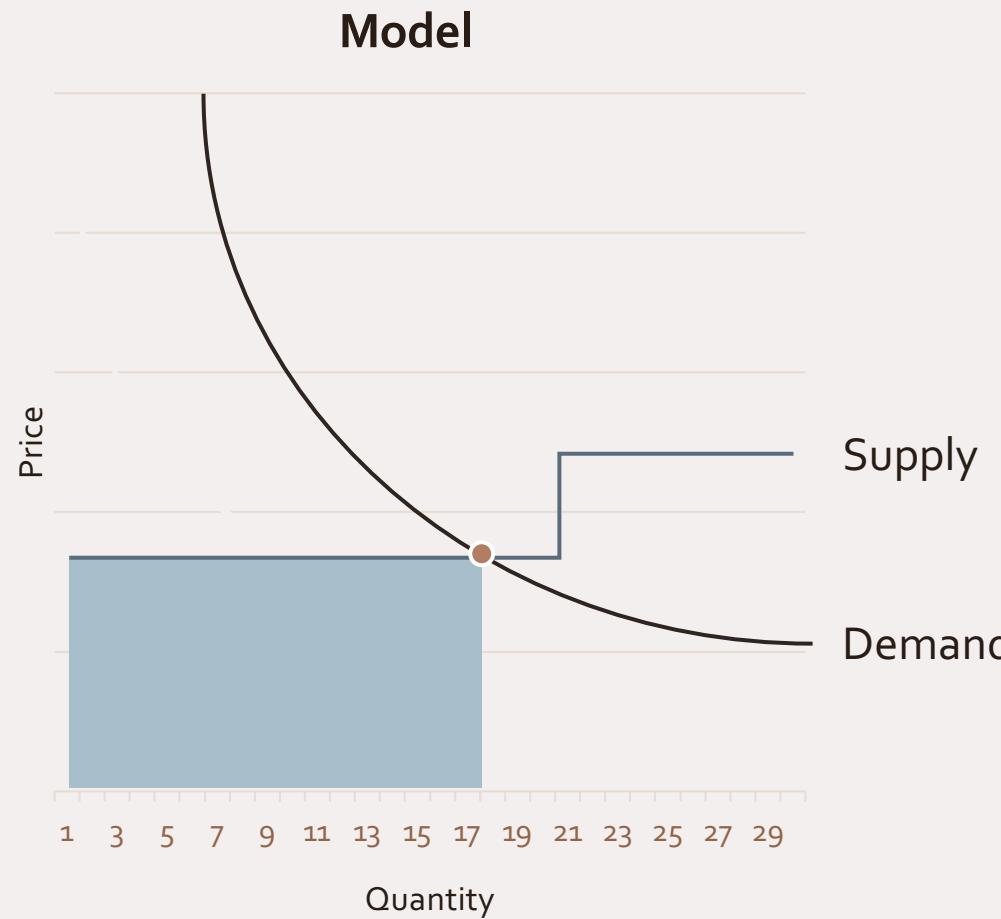


Modeling demand

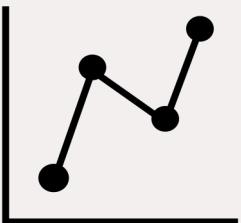




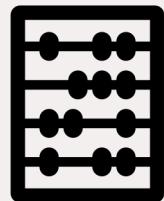
Modeling demand



Pipeline



Predicting water production



Modelling consumptions and price levels



Simulating policy interventions



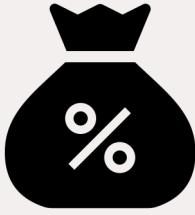
Machine Learning



Historical data and behavior modelling



Simulation



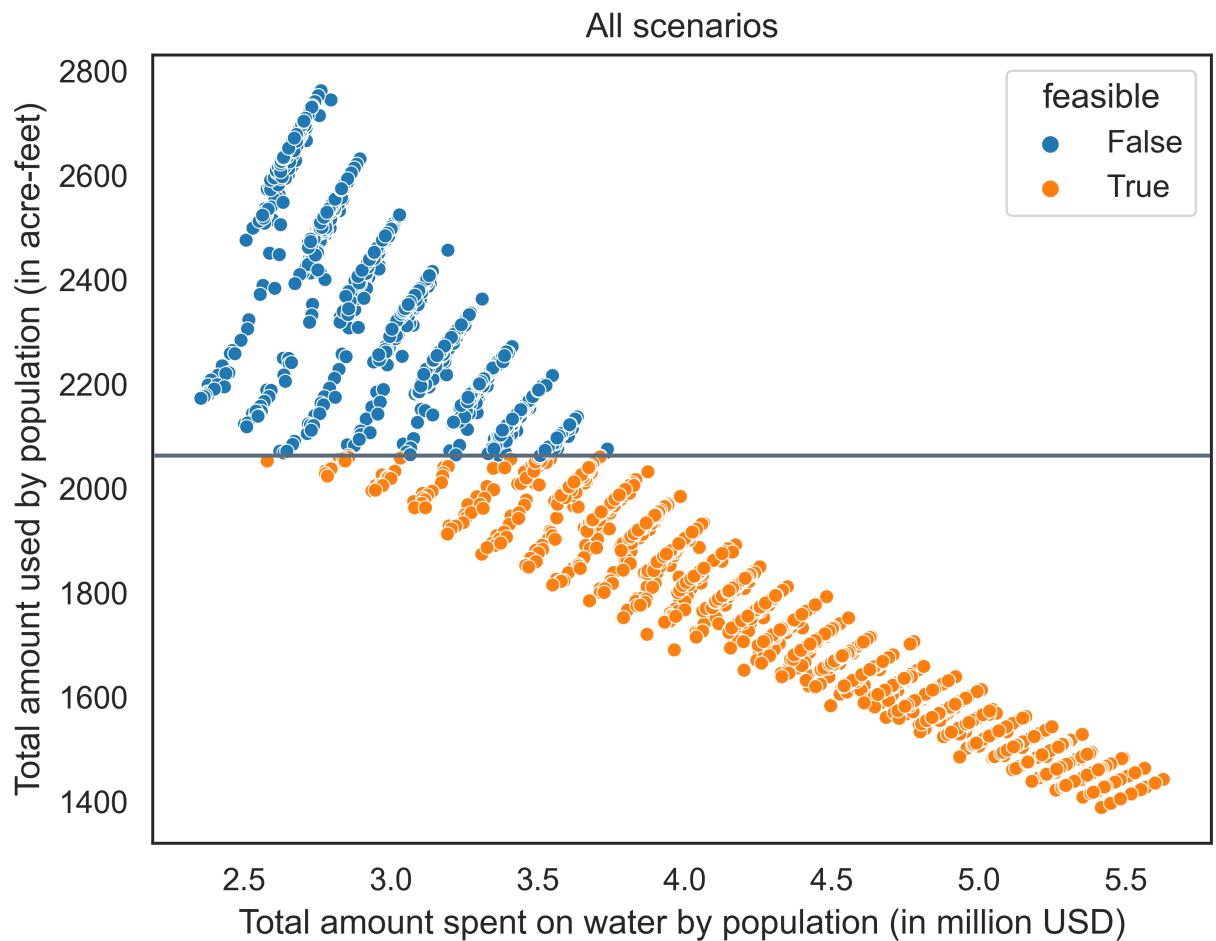
Simulation

- Simulating a fictitious town of 20,000 people
- All households have 4 persons
- Income distribution is equal to California's income distribution

Our demand curve revisited

Our model generates ~18,000 different scenarios by varying...

- Lower price
- Higher price
- Threshold between prices



How to find the right price?



Economy

A policymaker must consider the overall cost to the economy

higher water prices
→ reduce growth



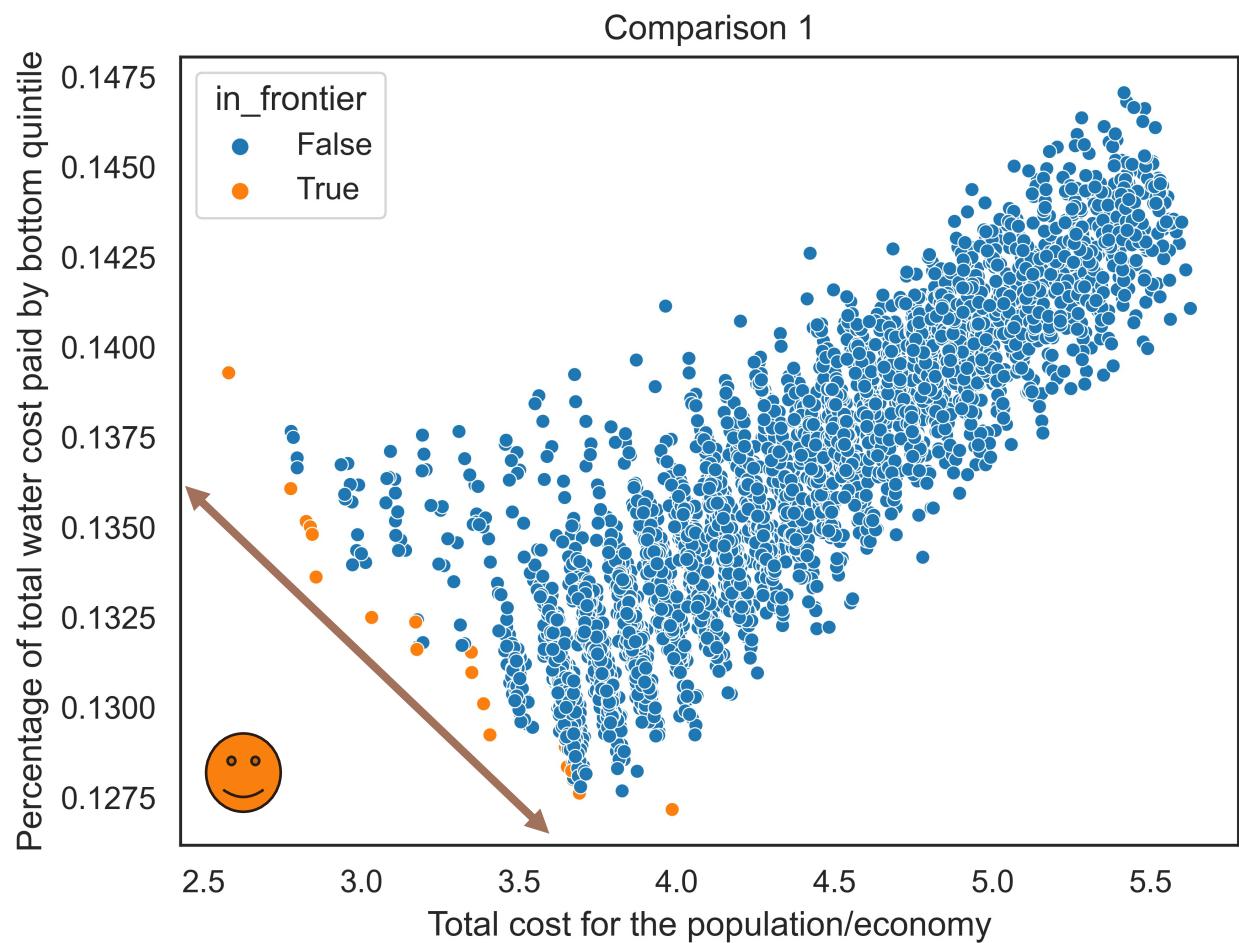
Equity

A policymaker also must make sure that the poorest in society still have access to affordable water

Equity can be defined in many ways

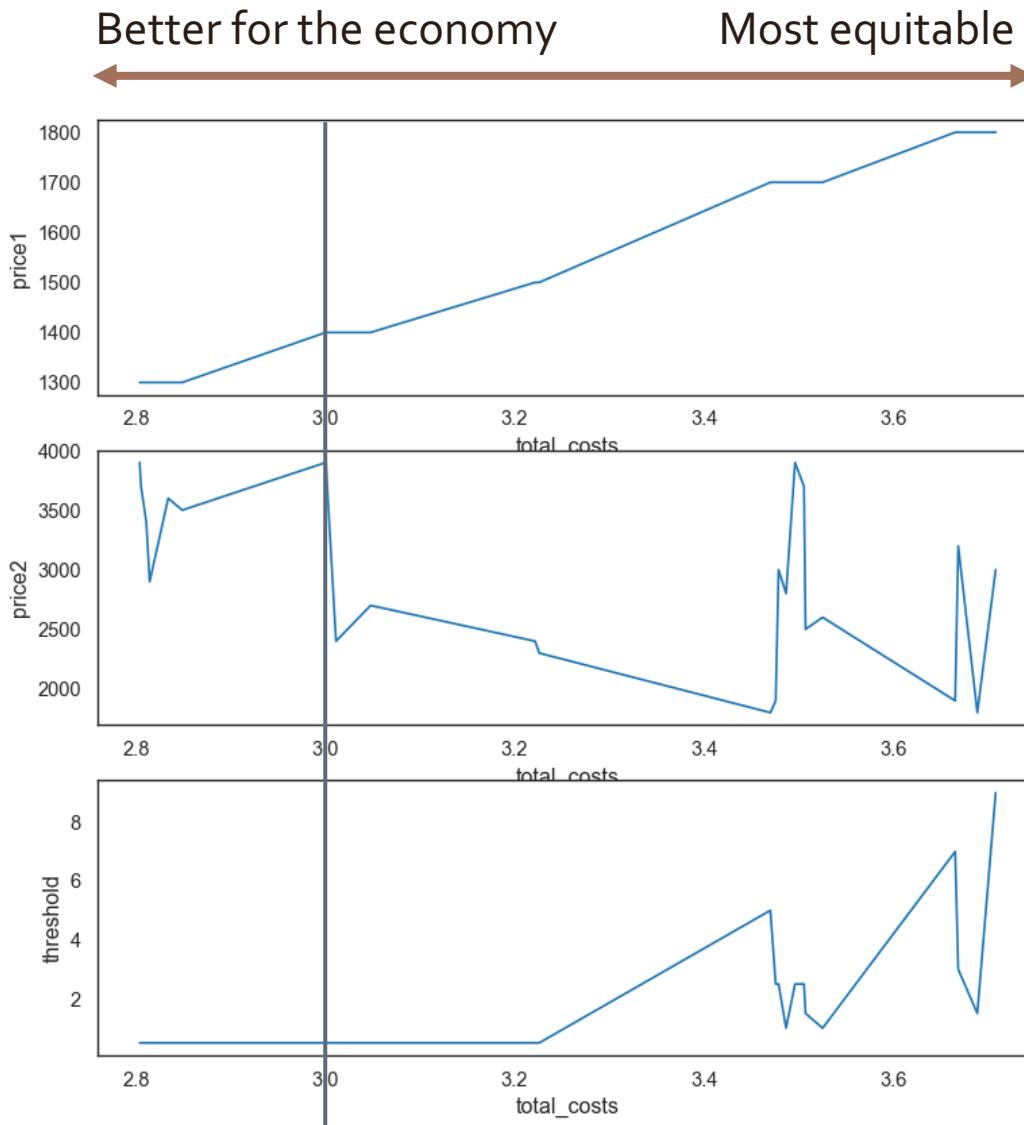
- Percentage of overall water expenses paid by poorest quintile
- Proportion of water used by poorest quintile (ideally 20%)
- Percentage of avg. water price of poorest compared to richest quintile

Economy vs. Equity

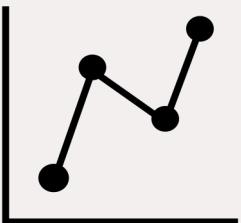




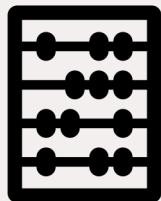
The options on the efficient frontier



Pipeline



Predicting water production



Modelling consumptions and price levels



Simulating policy interventions



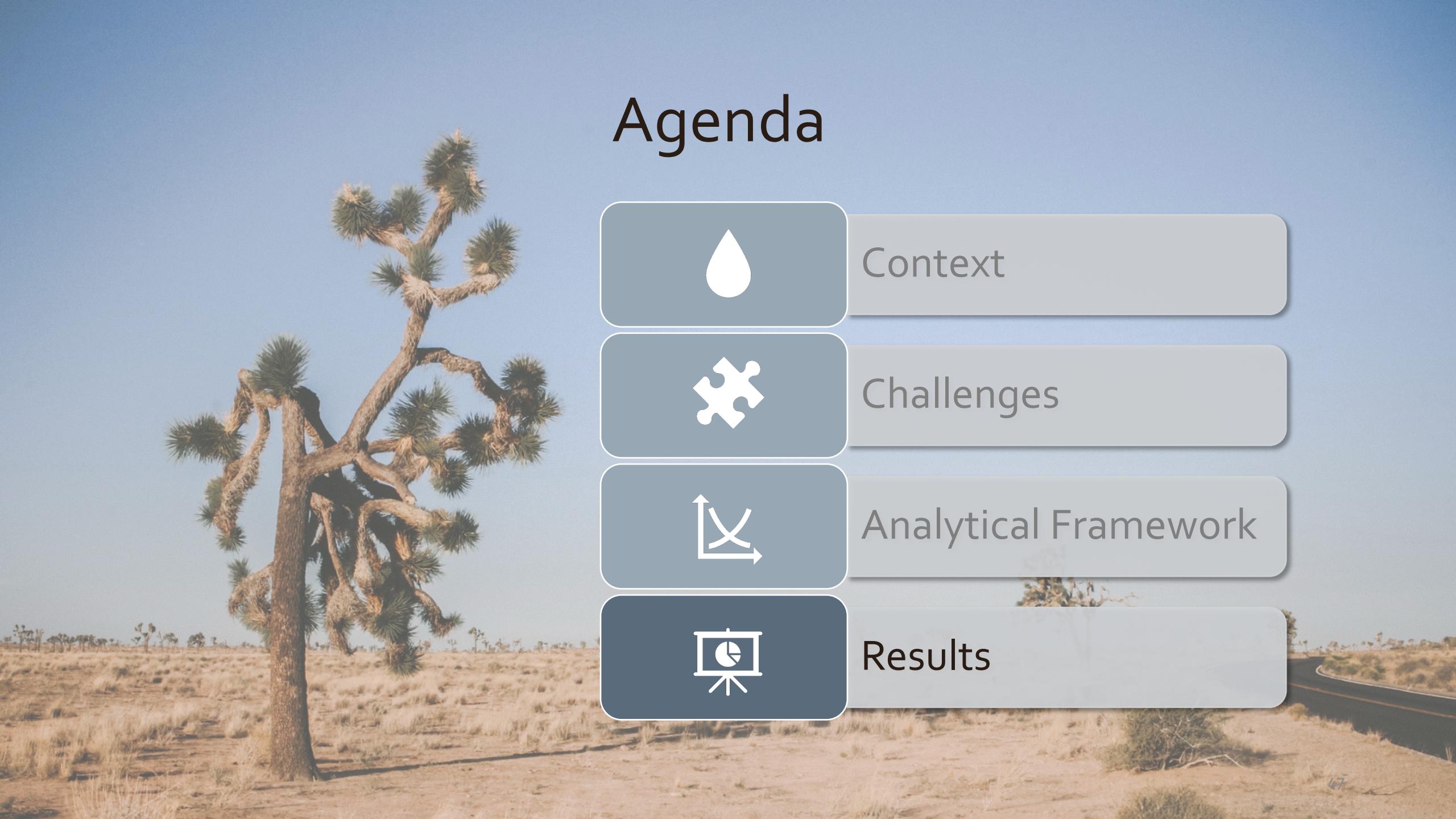
Machine Learning



Historical data and behavior modelling



Simulation



Agenda



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Challenges



Analytical Framework



Results

Menu of options for policymakers



Best for the economy:

\$0.4¢ until 0.4m gallons

\$1.2¢ thereafter

Most equitable:

\$0.5¢ until 2.6m gallons

\$0.9¢ thereafter

Summary



Drought conditions
likely to persist in
California

%

Need for a
comprehensive policy
that matches demand
to supply



Further consideration
by policymakers when
formulating potential
water allocation policy

Q&A



Sebastian Dodt
Sara Khoshhal
Karma Mroueh
Shantanu Samant

Decision Analytics
for Business and Policy

7 Dec 2022