

FIRST WORKSHOP PRESENTATION · JUNE 9, 2026

# AB 825 Jobs Assessment

PREPARED FOR

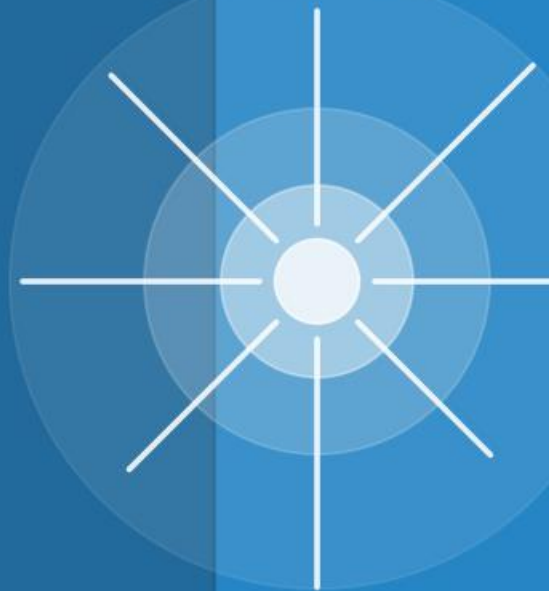
The California Independent System Operator

PREPARED BY

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# Today's Agenda

1

## INTRODUCTION

### Project Background & Scope

Motivation, CGE model design, and project schedule

2

## BASELINE JOBS

### Employment by Generation Technology

Fleet inventory, O&M workforce, wages, and construction intensity

3

## ECONOMIC MODEL

### Gravity Model Methodology

Bilateral pairwise specification, general performance statistics

4

## 2040 FORECAST

### Two-Stage Forecasting Pipeline

ARDL capacity + gravity-model share allocation

5

## UNCERTAINTY

### Monte Carlo Risk Framework

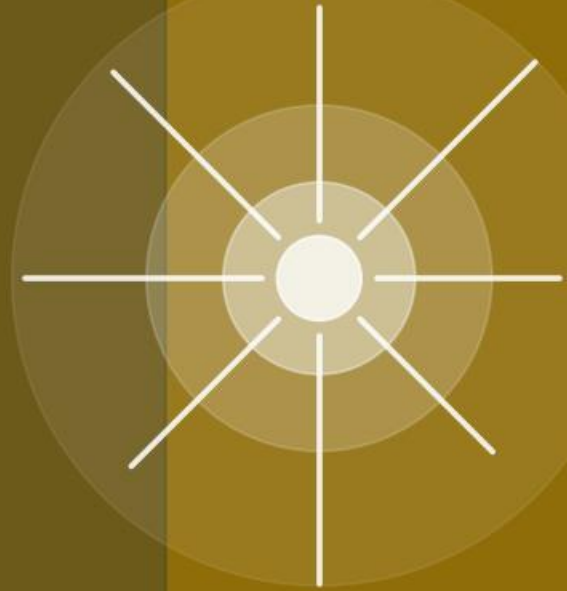
6 stochastic layers, 5,000 runs, Wilson 95% confidence intervals

# 1. Introduction to the Jobs Assessment

Motivation, modeling approach, and project schedule through  
December 2026



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- AB 825 authorizes CAISO to transfer regional market governance to an independent Regional Organization for Western Energy (ROWE).
- Section 4.(f) of the statute mandates an independent assessment of employment impacts — specifically jobs constructing and maintaining powerplants in California.
- The study must be completed by December 31, 2026, delivered to the Legislature, and incorporated into the ISO's findings and resolution.

***"up to \$10 billion in ratepayer savings over the next decade"***

— Estimate cited by California policymakers for ROWE regional market benefits

**Apr 2026**

BEAR LLC commissioned for the study.

**Dec 2026**

Statutory deadline for final report

**23 yrs**

BEAR's economic assessment experience in California

195

Activity sectors in the 2023 California SAM

23

Occupation categories — tracks powerplant jobs explicitly

10

Household income deciles (distributional detail)

25

Factors of production, with full government fiscal accounts

## 2023 California SAM

195 sectors · 23 occupations · 10 household deciles

## Model Baseline Calibration

Projected forward via DoF forecasts

## AB 825 Scenario Codification

Economic changes coded relative to Baseline

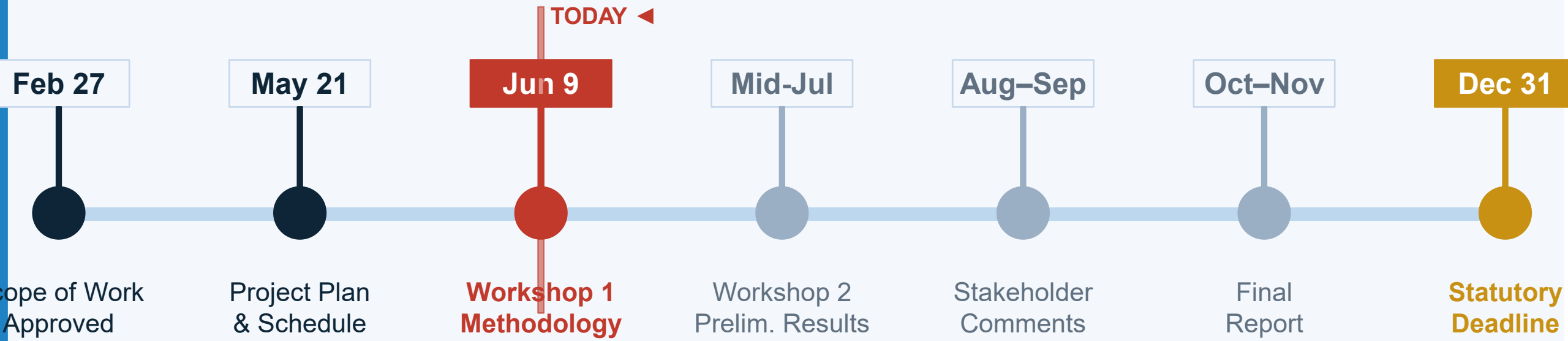
## Employment by Occupation

23-occupation decomposition tracks power-sector jobs

## Report to ISO and Legislature

Final report by December 31, 2026

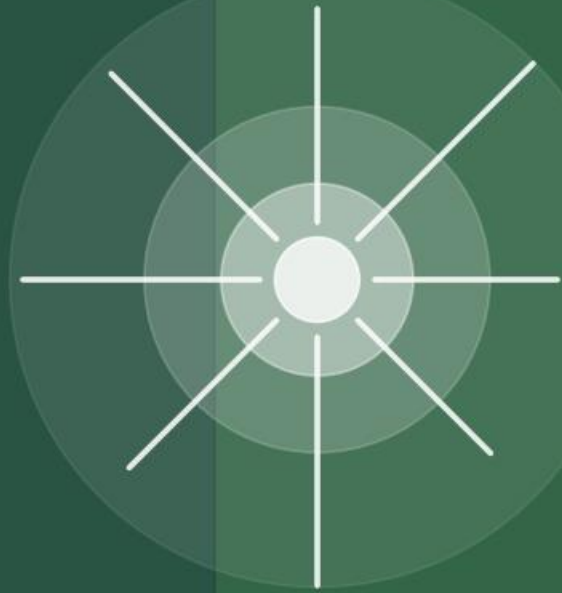
# Project Schedule – Key Milestones to December 2026



Statutory deadline: December 31, 2026 · Study delivered to the California Legislature

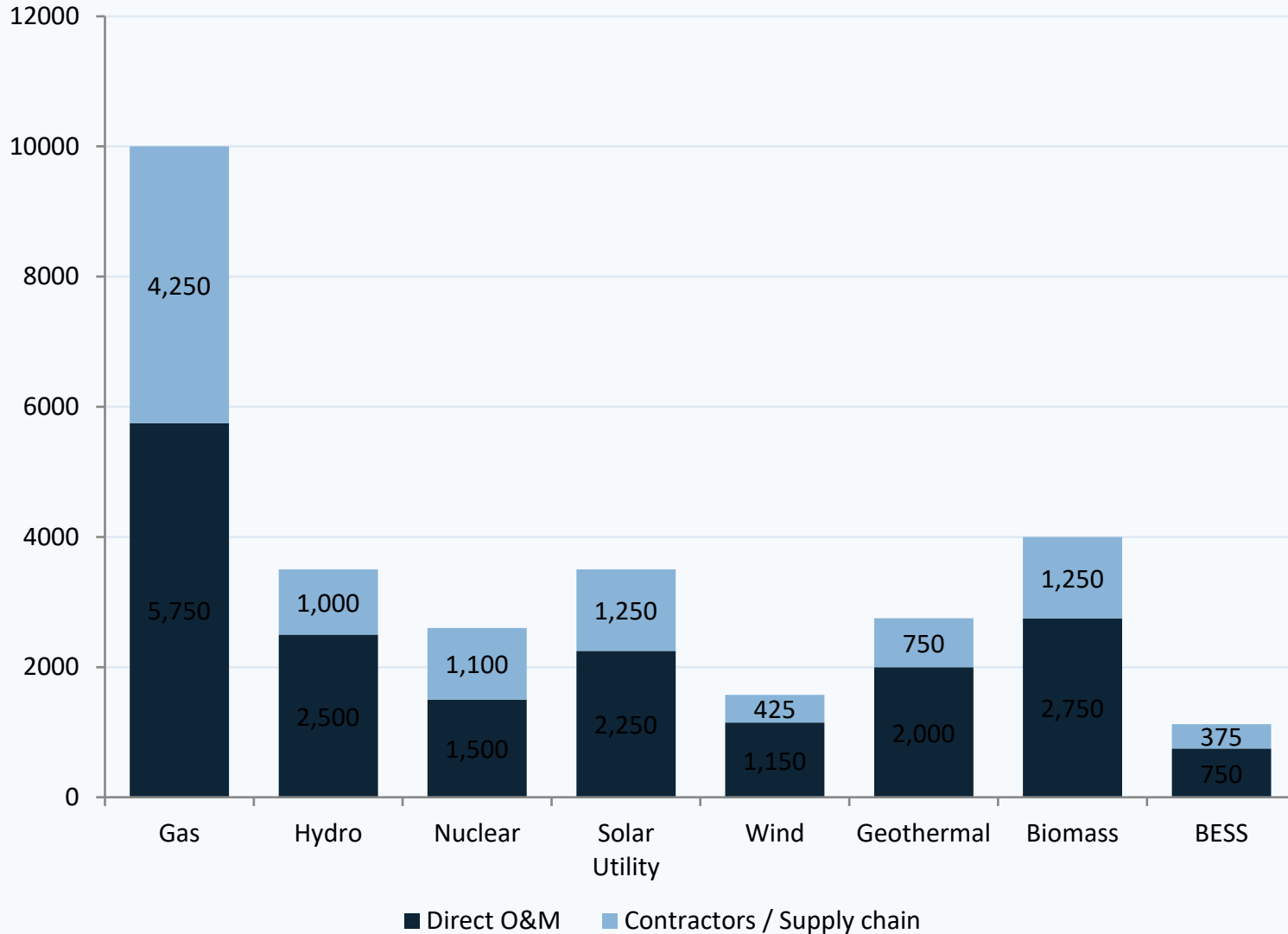
# 2. Jobs in California Electric Power Generation

Employment by generation technology — fleet size, O&M workforce, wages, and construction intensity



# Direct O&M Employment by Generation Technology, 2024

BASELINE JOBS



## Natural Gas

CAISO RMR contracts sustain 1,000–3,000 jobs at plants that would otherwise retire.

## Nuclear

Diablo Canyon pays the highest wages of any technology — \$110K–\$130K mean direct.

## Solar

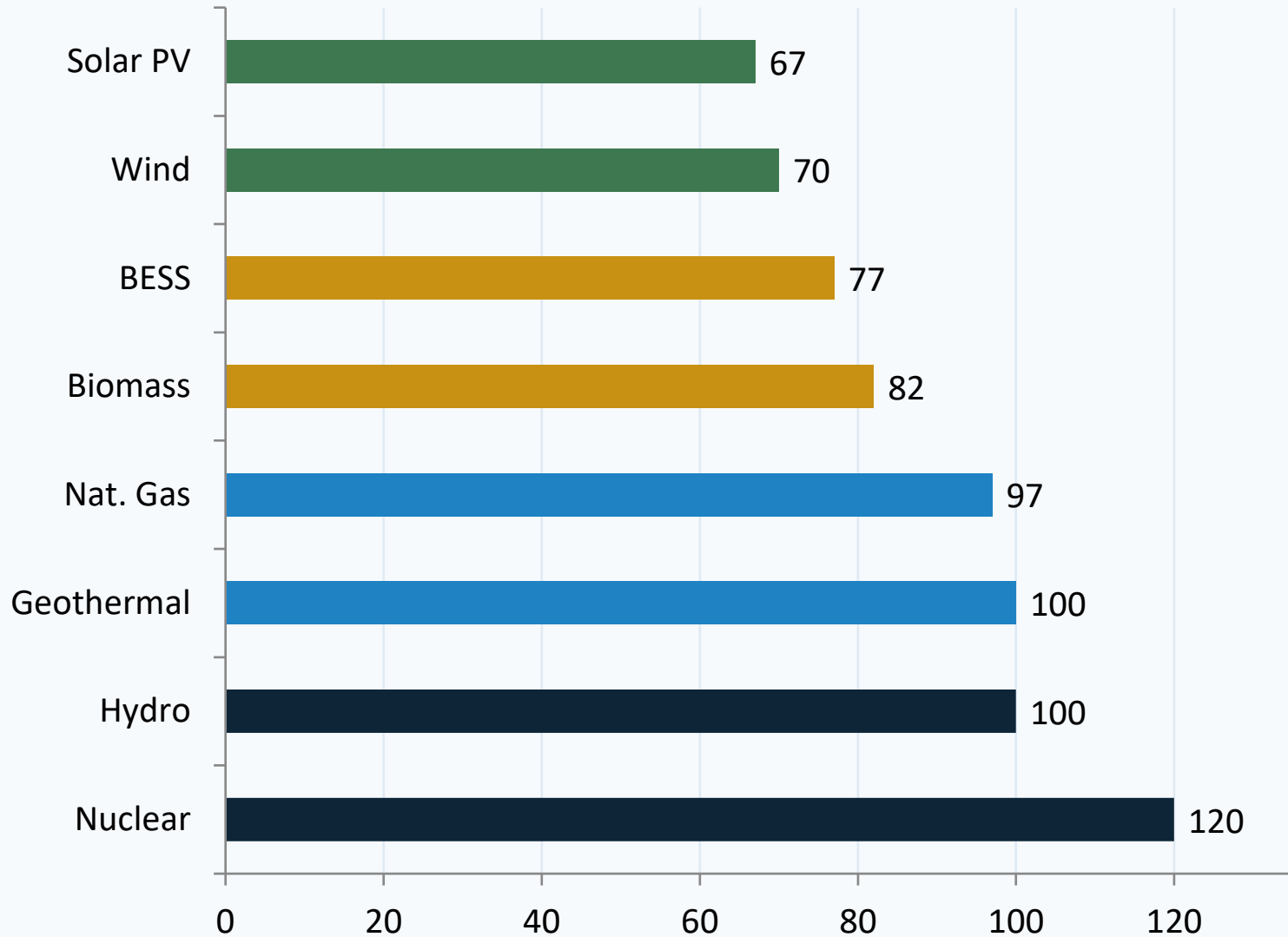
Distributed solar adds 30K–45K workers; construction runs 5–7 job-years/MW.

## BESS

Only 2–6 FTE per 100 MW — lowest O&M intensity in the fleet.

# Mean Annual Wages by Generation Technology (\$000s)

BASELINE JOBS



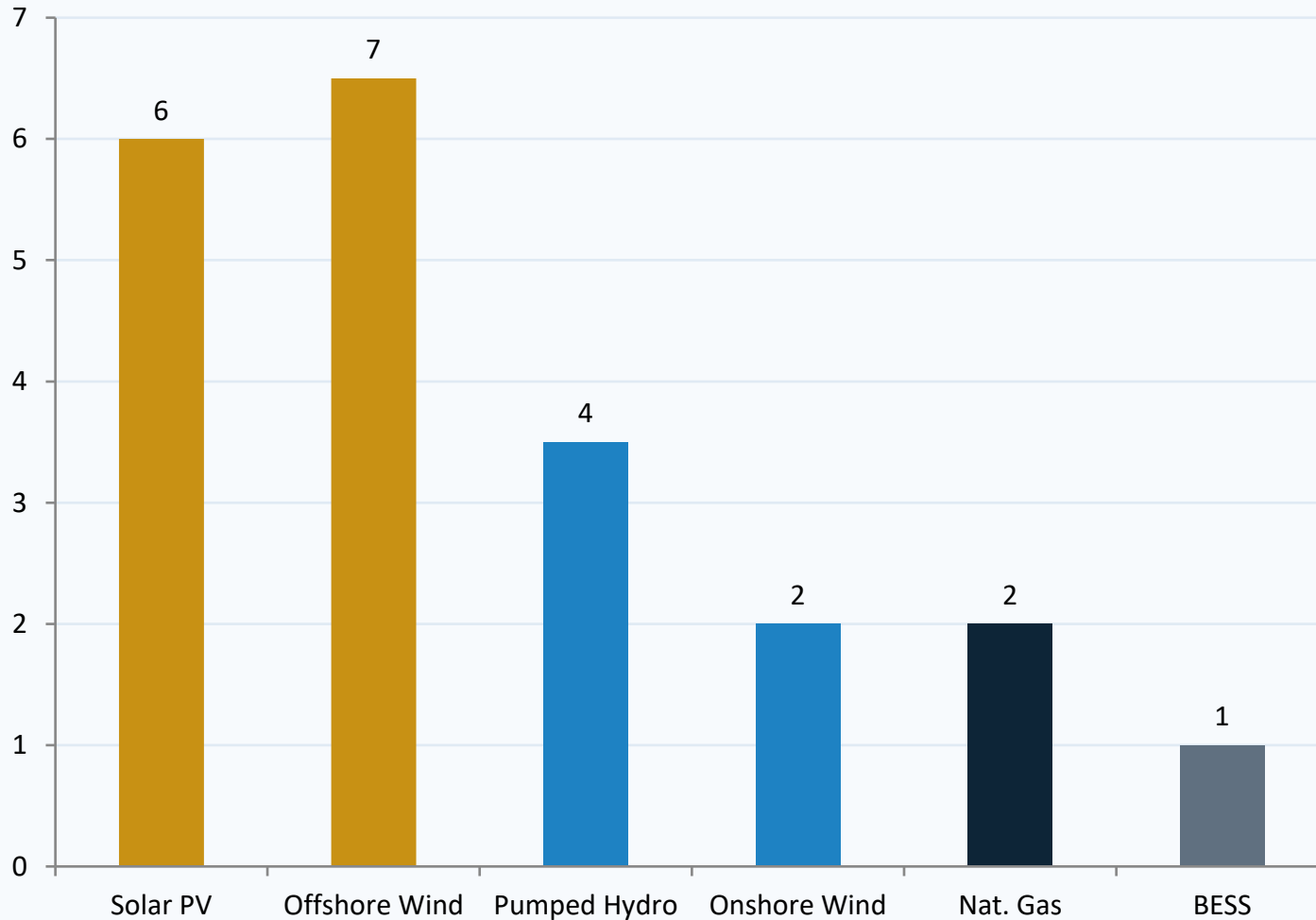
## Key Findings

- Nuclear commands the highest wages — licensed reactor operators earn \$120K–\$155K
- Hydro and geothermal significantly outpay wind and solar O&M
- Solar and BESS — California's fastest-growing technologies — carry lower wage profiles
- Total compensation including benefits is typically 25–40% above mean direct wages

# Fleet Overview & Construction Employment Intensity

BASELINE JOBS

Construction Job-Years per MW (Mid-Range Estimate)



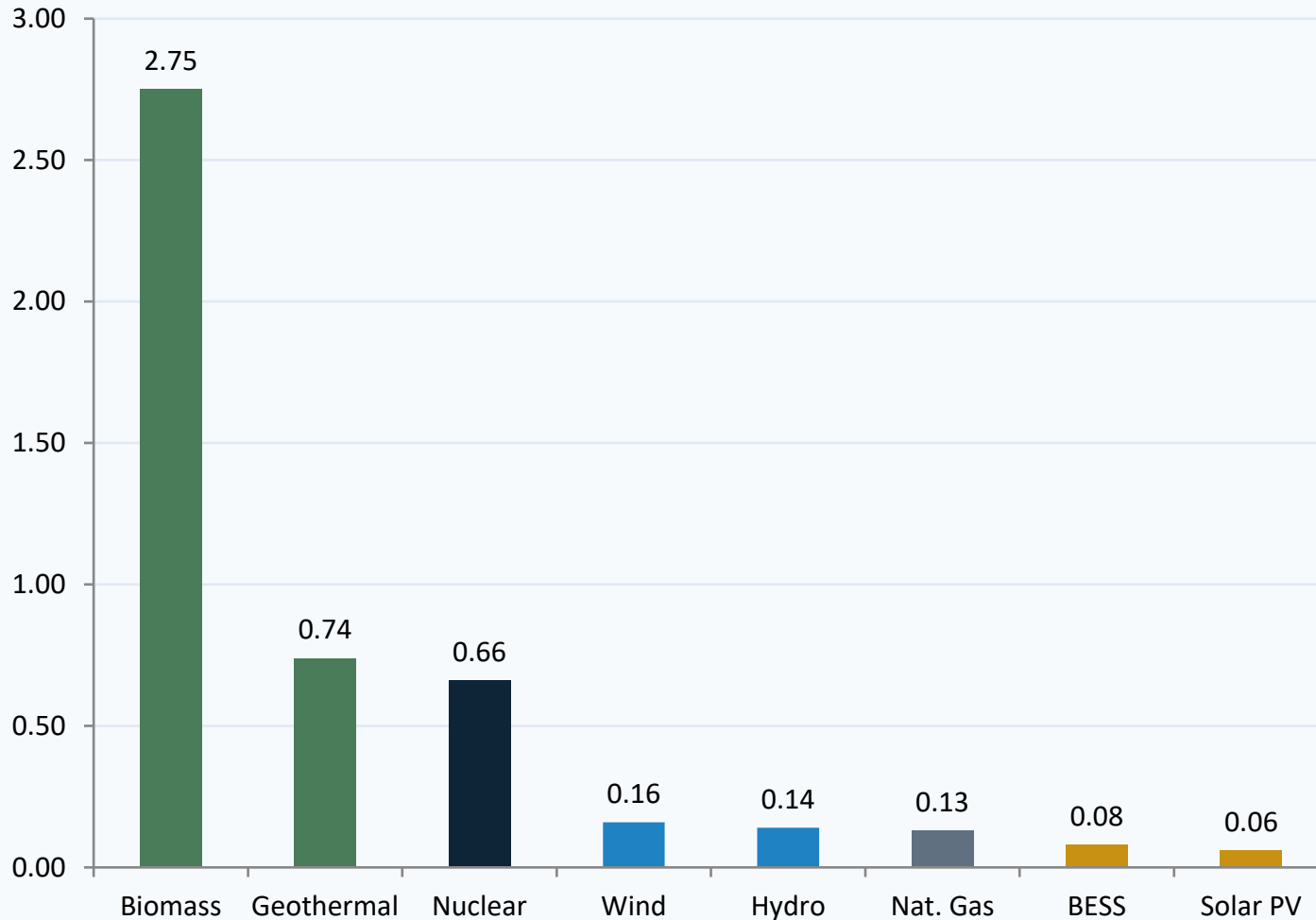
Technology	Fleet (MW)	Gen. Share
Natural Gas	~45,000	35–50%
Solar (utility+DG)	~40,000	25–30%
Hydro + Pumped	~17,500	10–15%
Battery Storage	8,000–10,000	—
Wind (onshore)	~7,000	8–12%
Nuclear	2,256	8–9%
Geothermal	~2,700	~2%
Biomass	~1,000	~1%
<b>Total (2025Q1)</b>	<b>~136,000 MW</b>	<b>—</b>

*Solar is California's most construction-employment-intensive pathway at 5–7 direct job-years per MW.*

# O&M Employment Intensity per MW

BASELINE JOBS

O&M Jobs per MW (Direct FTE, 2024)

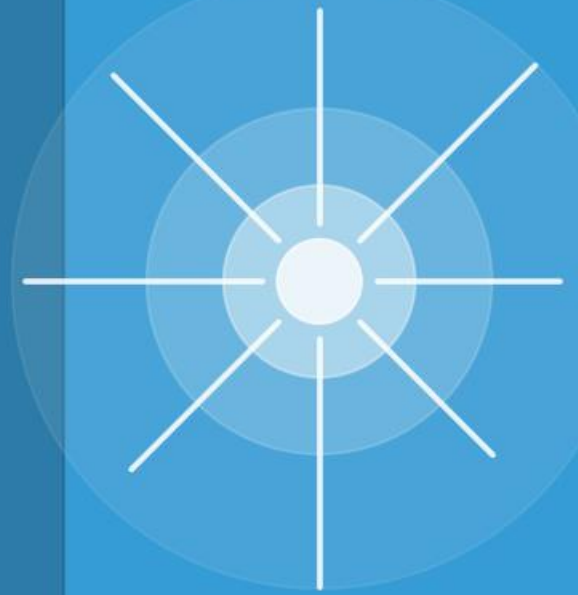


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<b>Total (2025Q1)</b>	<b>~136,000 MW</b>	<b>—</b>

*Biomass leads in O&M intensity (2.75 FTE/MW); geothermal and nuclear follow. Solar PV and BESS — California's fastest-growing technologies — are lowest at 0.06–0.08 FTE/MW.*

# 3. Gravity Model of Power Sector Competition

Bilateral pairwise specification ·  $R^2 = 0.92$  · 3,948 quarterly  
observations



# Bilateral Gravity Model – Specification & Intuition

$$\ln\left(\frac{G_i}{G_j}\right)_t = \alpha + \beta_1 \ln(\text{Cap}_i)_t + \beta_2 \ln(\text{Cap}_j)_t + \beta_3 \ln\left(\frac{\text{Cost}_i}{\text{Cost}_j}\right)_t + \gamma_1 \text{Q1}_t + \gamma_2 \text{Q2}_t + \gamma_3 \text{Q3}_t + \varepsilon_{ijt}$$

$\ln(G_i/G_j)$

Log-ratio of quarterly generation — technology i vs rival j

$\beta_1 \cdot \ln(\text{Cap}_i)$

Own-capacity: more installed MW → greater generation share

$\beta_2 \cdot \ln(\text{Cap}_j)$

Rival displacement — merit-order effect (consistent with solar curtailment)

$\beta_3 \cdot \ln(\text{Cost ratio})$

Relative cost: lower-cost technology gains dispatch share

$\gamma_1, \gamma_2, \gamma_3$

Seasonal effects — Q2 summer premium largest ( $\gamma_2 = 0.099$ )

**3,948**

Observations

**28**

Technology pairs

**141**

Quarters (1990Q1–2025Q1)

**8**

Technologies

**4**

Model variants

# Model Accuracy – Coefficient Estimates & Stability

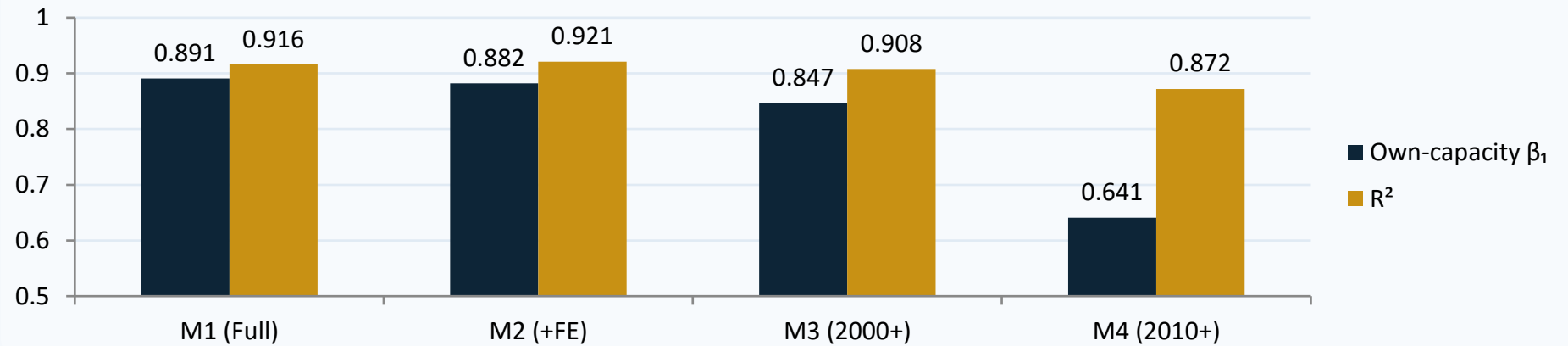
$R^2$

0.92

Full sample · M1  
HC3 robust SEs

Parameter	Estimate	Interpretation
$\alpha$ (intercept)	2.7034***	Baseline generation advantage
$\beta_1$ own-capacity	0.8911 ***	Sub-unitary: diminishing returns in constrained dispatch
$\beta_2$ rival-capacity	-1.2582 ***	Displacement > own-capacity — consistent with solar curtailment
$\beta_3$ cost-ratio	-0.3252***	Lower-cost technologies gain share; capacity dominates short-run
$\gamma_2$ Q2 summer	0.0986 ***	Summer generation premium
$\gamma_3$ Q3 autumn	0.0764	Autumn premium

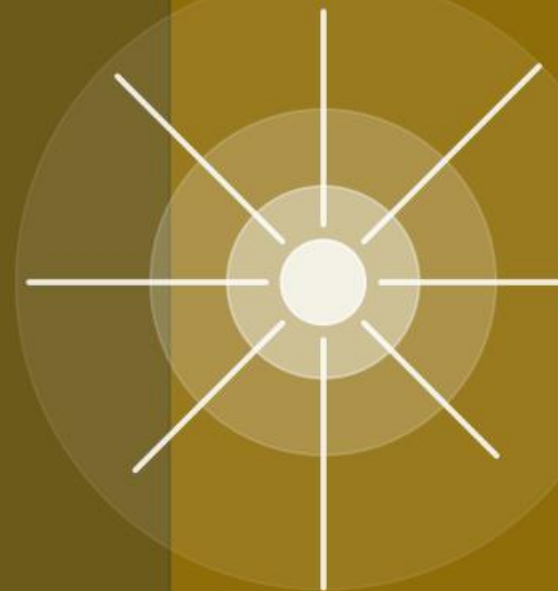
Coefficient Stability Across Model Variants



• \*\*\*  $|t| > 82$  · \*\*  $t = 2.61$  · HC3 heteroskedasticity-robust standard errors

# 5. Forecasting Jobs to 2040

Two-stage pipeline: ARDL capacity projection → gravity-  
model generation share allocation



## STAGE 1 – ARDL CAPACITY PROJECTION

- Auto-Regressive Distributed Lag model projects MW by technology from 2025Q1 base (136,066 MW total)
- Three growth scenarios (geometric structure): Low 1.0% · Reference 2.0% · High 4.0% p.a.
- Per-technology capacity bounds: floors at 80–85% of 2025Q1; ceilings cap runaway growth
- Nuclear floor in Low Scenario: Diablo Canyon held online through 2035
- AR sum rescaled for stationarity



## STAGE 2 – GRAVITY-MODEL SHARE ALLOCATION

- Gravity model converts ARDL capacity + cost-ratio inputs into bilateral log-generation scores for all 28 pairs
- Scores assembled into an antisymmetric 8×8 matrix
- Thurstone row-mean scoring + softmax normalization → portfolio generation shares
- Scale-invariance property: technology mix converges similarly across scenarios
- Employment intensities (jobs/MW, jobs/TWh) applied → quarterly trajectories to 2039Q4

# Three Capacity Growth Scenarios — Geometric Structure

LOW

**1.0%**  
**p.a.**

Constrained build environment

*Diablo Canyon nuclear floor held online through 2035*

REFERENCE

**2.0%**  
**p.a.**

Central case — current policy

*Reference =  $\sqrt{(\text{Low} \times \text{High})}$   
Geometric mean by design*

HIGH

**4.0%**  
**p.a.**

Accelerated clean energy deployment

*Offshore wind, deep electrification,  
longer nuclear operation*

Geometric structure: Reference =  $\sqrt{(\text{Low} \times \text{High})} \rightarrow \sqrt{(1.0\% \times 4.0\%)} = 2.0\%$  · Reference is the geometric mean — no arithmetic upward bias

2025Q1 Base  
(136,066 MW)

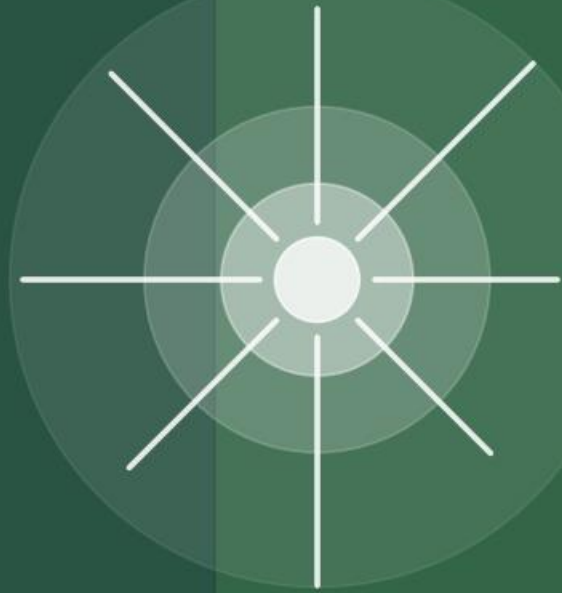
ARDL  
Capacity Forecast

Gravity Model  
Share Allocation

Employment  
Trajectories → 2040

# 5. Dealing with Uncertainty

Monte Carlo simulation framework — six stochastic layers propagated across 5,000 runs



# Monte Carlo Risk Framework – Six Stochastic Layers

UNCERTAINTY

**5,000**

Simulation runs

**≈8 sec**

Python runtime

**60**

Forecast quarters

**1/3**

Equal scenario probability

**L1**

## Scenario Draw

Low / Reference / High drawn with equal probability (1/3) at the start of each run

**L2**

## Capacity Growth

ARDL recursion with per-technology log-capacity bounds; floors at 80–85% of 2025Q1

**L3**

## Cost Trajectories

Technology costs via Exponential Smoothing (ES); feeds cost-ratio inputs to gravity model

**L4**

## Gravity Coefficients

$\beta$  vector drawn from MVN([2.703, 0.891, -1.258, -0.325, ...],  $\Sigma_{HC3}$ ) each simulation

**L5**

## Capacity Factor Sigma

Low: 1.1× sigma multiplier; High: 1.3× to reflect faster deployment variability

**L6**

## Employment Intensity

Stochastic variation around technology-specific jobs/MW and jobs/TWh baseline ratios

## How the Layers Connect

1. Draw scenario,  $\beta$  cost paths, cap-factor sigma

2. Run ARDL recursion → projected capacity by technology

3. Compute 28 bilateral log-generation scores

4. Assemble antisymmetric 8×8 matrix

5. Thurstone scoring + softmax → portfolio shares

6. Apply employment intensities → quarterly job trajectory

**Repeat × 5,000 runs · seed = 42 · ~8 seconds ·  
Wilson 95% CIs**

## Simulation Outputs

`mc_summary_stats.csv`

Percentile distributions by quarter and technology

`mc_scenario_means.csv`

Scenario-conditional mean trajectories (Low/Ref/High)

`mc_risk_metrics.csv`

VaR-style employment risk metrics

`mc_sensitivity.csv`

Sensitivity decomposition across stochastic layers

`mc_run_log.txt`

Full run log — seed=42 for reproducibility

*Python 3.10+ · NumPy / SciPy / pandas · Wilson 95% confidence intervals*

## NEXT STEPS

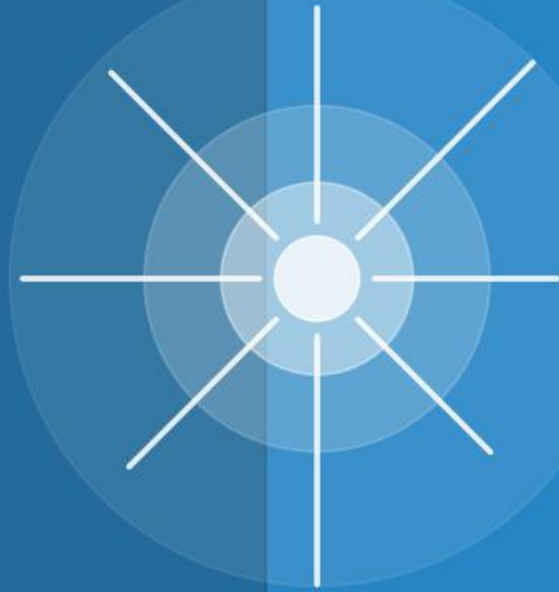
# What Comes After This Workshop

**Mid-July 2026** Workshop 2 — Results by scenario and technology

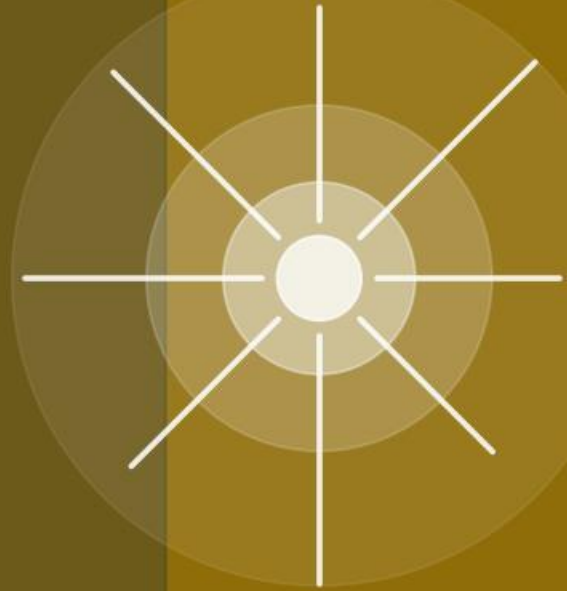
**Aug–Sep 2026** Stakeholder comment period on methodology and results

**Oct–Nov 2026** Final report incorporating stakeholder feedback

**Dec 31, 2026** Statutory deadline — report delivered to the Legislature



# Questions & Discussion



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