

Merchant Renewable Asset Fixed-Price Valuation (2026–2030)

1. Data and Assumptions

We analyze three merchant renewable assets – two wind farms (in ERCOT and MISO) and one solar farm (in CAISO) – using provided historical data (2022–2024) and forward price curves (2026–2030). The historical dataset contains hourly generation (per unit of 100 MW capacity) and corresponding Day-Ahead (DA) and Real-Time (RT) locational prices at both the asset’s local bus (node) and the regional hub. Key assumptions include:

- **Normalized Capacity:** The generation data are given as a percentage of a 100 MW capacity (0–100). We treat the “Gen” values as the actual MW output for a 100 MW plant. For example, a value of 50 means 50 MW output in that hour. This yields realistic capacity factors (e.g. ~20% for wind, ~35–40% for solar over a year).
- **Peak vs Off-Peak Hours:** We use standard forward market definitions for peak (on-peak) and off-peak periods, which vary by ISO:
 - **ERCOT & CAISO:** Hours ending 7–22, Monday–Friday (16 hours on weekdays, excluding major holidays) are Peak; other hours (nighttime 23–6, weekends, holidays) are Off-Peak [molecule.io](https://www.molecule.io). (Note: CAISO generally includes Saturday in the “weekday” peak block as a 6x16 product [molecule.io](https://www.molecule.io), whereas ERCOT and MISO use a 5x16 definition.)
 - **MISO:** Hours ending 8–23, Monday–Friday, are Peak; all other hours are Off-Peak.

These definitions align with the forward price products (e.g. “Peak 7x16” blocks) used in the provided forward curves. We do not explicitly adjust for future holidays beyond what is inherent in the block definitions, since the forward prices already assume the standardized blocks.

- **Negative Generation Values:** Any instances of negative generation in the data (e.g. minor station power draw) are set to 0 for forecasting, since a wind/solar plant cannot deliver negative output (we assume any small consumption is negligible). In the given data, generation values were already floored at 0 (no negative outputs observed).
- **Negative Prices and Curtailment:** The historical price data show frequent negative prices, especially at the wind farms’ local bus nodes (e.g. the MISO wind’s bus had ~25% of hours with negative DA prices during high wind periods). In our revenue model, we assume the project would not generate (i.e. it would curtail) during negative-price hours if the PPA contract does not compensate negative prices. This

means we impose a revenue floor of \$0/MWh – any generation that would occur when price < \$0 is assumed curtailed (no revenue). This reflects a PPA where the offtaker **does not pay for power during negative price periods**, so the seller avoids paying to generate power pexapark.com. Physically, the operator would likely curtail output in those hours to limit losses pexapark.com. This assumption slightly *raises* the realized average price (by avoiding negative revenue events) while slightly *reducing* total delivered energy. We include this logic when calculating revenues and risk (especially for wind, which sees most negative-price events at high output times).

- **Forward Prices:** We use the provided hub forward price curves for 2026–2030. These are monthly forward prices for Peak and Off-Peak blocks at the relevant trading hubs (e.g. ERCOT North, MISO Minnesota Hub, CAISO SP15 or similar). Each forward price is essentially the expected average price for a **flat 25 MW** constant output during that period (peak or off-peak hours). We assume these forward prices represent the market’s consensus *expected* DA prices, so they are suitable for calculating expected revenues. (No separate forecast for DA vs RT is needed, as the forward curve should reflect both on average.) However, we will incorporate historical **DA–RT price spreads** as an adjustment for risk or imbalance, as described later.

All analysis is done via a Python model for full reproducibility. The code reads the datasets, performs the calculations for generation and pricing, and outputs the results (including tables and charts) to an Excel file as specified.

2. Monthly Generation Forecast (2026–2030)

We forecast each asset’s generation on a monthly basis for peak and off-peak periods using a deterministic seasonal average method. The steps are:

- **Historical Seasonal Averages:** For each asset, we compute the average generation profile for each calendar month, split by peak vs off-peak hours, using the 2022–2024 data. Essentially, we take the hourly generation data for (say) all Januaries in 2022–24 and find the average generation during peak hours and during off-peak hours. This gives a typical January output for that asset in peak and off-peak times. We repeat for all months 1–12.
- **Forecast 2026–2030:** We assume the asset’s future output follows the same seasonal pattern (no major inter-annual trend or degradation over 2025–2030). Thus, the expected generation in (for example) January 2026 is set equal to the historical average January generation, and similarly for each month through 2030.

Essentially, the 12-month profile is repeated each year 2026–2030. This deterministic approach ignores year-to-year weather variability but provides a **P50 (mean)** outlook for generation.

- **Peak vs Off-Peak Breakdown:** We preserve the split between peak and off-peak generation. The output is a table of expected **monthly energy (MWh)** generated in peak hours and in off-peak hours for each asset, from January 2026 to December 2030. These values are the key inputs to revenue calculations, since they tell us how much energy will earn the peak price vs the off-peak price each month.
- **Treatment of Negative Outputs:** As noted, any negative generation would have been set to zero before averaging, though none were present. So all forecasted generation values are non-negative.

Results – Generation Profile: The wind farms show strong seasonal and diurnal patterns, whereas the solar farm’s generation is confined to daytime hours:

- **Wind (ERCOT & MISO):** The wind assets tend to generate more in cooler months (late fall, winter, spring) and slightly less in mid-summer. Notably, a large portion of wind energy comes during off-peak hours (nighttime). For example, the ERCOT wind farm generates ~55% of its energy in off-peak hours during winter months (strong night winds). The MISO wind farm similarly has a high off-peak share, particularly in spring nights. In spring (e.g. March), MISO wind output is highest – our MISO wind produces roughly **10,000 MWh off-peak vs 8,300 MWh on-peak in March (per 100 MW)**, then drops to around **4,600 MWh on-peak in July** when winds are weaker.
- **Solar (CAISO):** The solar farm produces exclusively during the day, so virtually all weekday generation falls in peak hours by definition. It has near-zero output during off-peak nighttime hours on weekdays. However, note that **weekends** are classified as off-peak, so **weekend daytime solar generation is counted in “off-peak”**. This means the solar asset does have some off-peak energy (weekend days). Overall, the CAISO solar output peaks in the summer months (longer, sunnier days) and is lowest in winter. It has **no generation at night**, and thus off-peak output is only ~2/7 of its energy (the portion from weekends).

*MISO Wind – Typical monthly generation vs. 2026 forward prices. **Top:** Bars show the average monthly generation for the MISO wind farm, split into Off-Peak (night/weekend) and Peak (weekday daytime) MWh. **Bottom (lines):** The forward market’s expected Peak and Off-Peak power prices (hub) for 2026. We observe the wind resource is high in spring (March–April), when forward prices are relatively low, and wind output drops in mid-*

summer when forward prices peak, indicating a negative correlation between generation and price.

Overall, these profiles highlight a critical aspect for valuation: the **mismatch between generation timing and price**. Wind tends to generate more during low-demand periods (nights or spring) when prices are lower, while solar generates during daylight which can align with high prices but also coincides with periods of solar oversupply (low midday prices in some regions). We will account for this by weighting the forward prices with the generation profile.

3. Fixed Price Modeling (Risk-Adjusted)

We now derive the **fixed prices** (in \$/MWh) that one could offer for a Power Purchase Agreement (PPA) for each asset. Four scenarios are considered for each asset, corresponding to where and how the energy is settled: **Day-Ahead Hub, Day-Ahead Bus, Real-Time Hub, Real-Time Bus**. These represent combinations of market (DA vs RT) and price settlement point (hub vs local busbar). The fixed prices will be computed by starting from the expected hub price (based on forwards and generation weighting) and then adjusting for locational basis, DA vs RT spreads, and a risk discount for P50/P75/P90 as required. We provide a breakdown of each component.

3.1 Generation-Weighted Hub Forward Price (P50 Base)

First, we determine the **baseline expected price at the hub** that the asset's generation would earn, before any risk discounts or basis adjustments. This is essentially the value of a pay-as-produced contract at hub with no risk discount (a P50 expected value). Because the asset's output is variable, its *captured price* can differ from a flat 25 MW block price. We calculate the generation-weighted forward price as:

$$P_{\text{hub, gen-weighted}} = \frac{\sum_{\text{month}} (\text{Gen}_{\text{peak,month}} \cdot F_{\text{peak,month}} + \text{Gen}_{\text{offpeak,month}} \cdot F_{\text{offpeak,month}})}{\sum_{\text{month}} \text{Gen}_{\text{total,month}}}$$

where $\text{Gen}_{\text{peak,month}}$ and $\text{Gen}_{\text{offpeak,month}}$ are the forecasted energy (MWh) in peak and off-peak hours for that month, and $F_{\text{peak,month}}$, $F_{\text{offpeak,month}}$ are the forward prices (\$/MWh) for that month's peak and off-peak periods. Essentially, we weight each monthly forward price by the fraction of the project's generation that will occur in that month and period. This gives a single average price over 2026–2030 that the project would capture at the hub.

Calculated P50 Hub Prices: Using the above, we found:

- **ERCOT Wind:** ~\$51.9/MWh (hub, generation-weighted)

- **MISO Wind:** ~\$44.3/MWh
- **CAISO Solar:** ~\$44.3/MWh

These values are the **expected energy prices at the hub** that each asset would earn on average (P50 case). The ERCOT wind captures a higher price because ERCOT forward prices are quite high in summer (scarcity pricing) when that wind farm still produces a fair amount. The MISO wind's captured price is lower – its strongest production is in spring when forward prices are lowest, and it faces frequent negative pricing events that drag down its average. The CAISO solar's captured price is about \$44/MWh; despite high forward off-peak (nighttime) prices in some months, the solar can't benefit due to zero night generation, and in spring the midday (peak) prices are depressed by solar oversupply (some forward peak prices in CAISO spring are very low, even ~\$11/MWh in April).

Formula check: For example, the ERCOT wind's January 2026 forward prices are around \$66.4 (peak) and \$61.7 (off-peak). In that month our wind forecast generates ~54% of its energy in off-peak hours (taking advantage of strong night winds) and ~46% in peak. The generation-weighted price for Jan-2026 comes to about \$63.9/MWh. Doing this for every month 2026–2030 and averaging yields the \$51.9/MWh figure for ERCOT wind.

This generation-weighted hub price is the starting point for building each PPA price scenario. It reflects the *flat fixed price at the hub that would give the project the same revenue as the expected shape-driven market revenue*. In essence, this is the project's **P50 value of energy** at the hub, before considering basis or other risks.

3.2 Basis Adjustment (Hub vs Busbar)

“**Basis**” in this context is the price difference between the project's local node (busbar) and the trading hub. The project physically earns the nodal (busbar) price for its energy, but a PPA might settle at hub or at bus. **Basis risk** is the risk that these two diverge projectfinance.lawcityrenewables.org. Historically, our wind farms experienced substantial negative basis (node price lower than hub), due to congestion in their area during high generation periods. The solar farm's node was very close to the hub price on average.

From 2022–2024 data, the **average locational basis (Bus – Hub price)** for the assets was approximately:

- **ERCOT Wind:** –\$9.7/MWh in Day-Ahead (i.e. node was ~\$9.7 below hub on average), and –\$11.4 in Real-Time.
- **MISO Wind:** –\$12.85 Day-Ahead, –\$12.14 Real-Time.

- **CAISO Solar:** essentially \$0 (slight +\$0.1 Day-Ahead, –\$0.05 RT, effectively no basis difference).

This means the wind projects are in areas with frequent congestion or curtailment – when they generate a lot, local prices collapse relative to the broader hub. For instance, the MISO wind farm is likely located in a congested wind zone (perhaps MISO Midwest) where many wind farms drive the local price down (even negative) during windy hours [projectfinance.law](#). Such situations are common (e.g. parts of Texas Panhandle or Midwest where wind build-out exceeds local transmission) and **projects with high basis risk often must either settle at the bus or factor in that risk in the PPA price** [cityrenewables.org](#).

Applying Basis to Fixed Prices: For PPA scenarios that settle at the busbar, we adjust the price by the expected basis. We use the historical average basis as a best estimate for future (assuming grid conditions remain similar). Specifically:

- **Day-Ahead Bus price** = Day-Ahead Hub price + (avg. DA bus-hub basis).
- **Real-Time Bus price** = Real-Time Hub price + (avg. RT bus-hub basis).

In our case, the bus-hub basis numbers are negative for wind (so this will *lower* the bus-settled PPA price relative to hub) and basically zero for the solar (so hub or bus hardly matters). For example, the ERCOT wind’s average DA basis of –\$9.68 means a PPA settled at the project node would be priced about \$9.68 lower than an otherwise equivalent hub-settled PPA. We will see this reflected in the final prices (ERCOT wind DA @ Hub ~\$50.5 vs DA @ Bus ~\$40.9 under P75 – the difference is largely that ~\$9.7 basis).

Why subtract basis? If the project’s node tends to be lower than the hub, a buyer paying only hub price would “overpay” relative to the project’s actual revenue. So a bus-settled PPA will be discounted so that the expected value at the bus equals the contract. Conversely, if a node were higher than hub (positive basis), a bus PPA might be slightly higher. In our case, only CAISO solar has a tiny positive DA basis (~+\$0.11), so its bus-settled price comes out a few cents higher than hub, essentially negligible.

3.3 Day-Ahead vs Real-Time Pricing

Another factor is the **DA–RT price spread**. Generators typically lock in day-ahead prices for scheduled energy but can be exposed to real-time price volatility for deviations. Our forward prices are more aligned with day-ahead market expectations (since most forward contracts settle against DA indices). However, if a PPA were to settle on real-time prices, we may need to adjust for any systematic differences or risk premium between DA and RT.

Historically, the average **(Day-Ahead price – Real-Time price)** at the hub was:

- ERCOT: +\$3.9 (DA was higher on average than RT).
- MISO: +\$0.3 (almost no difference).
- CAISO: +\$1.6 (DA higher than RT on avg).

A positive number means the day-ahead market tended to *overestimate* prices slightly (or conversely, real-time came in a bit lower). This could be due to forecast conservatism or scarcity adders that didn't materialize. For PPA pricing, a consistent DA over RT difference suggests that if a contract is settled on real-time, the expected revenue might be a bit lower – thus a PPA strictly on RT prices would warrant a slightly lower fixed price (to match the lower expected revenue).

We apply an adjustment for real-time settlement as:

- **Real-Time Hub price** = DA Hub price + (avg. RT hub price – DA hub price). (Since our baseline was DA hub forward, effectively subtract DA–RT difference.)
- **Real-Time Bus price** = DA Bus price + (avg. RT bus – DA bus difference).

In practice, for ERCOT and CAISO this means subtracting a few dollars. For example, ERCOT's RT was on average \$3.92 lower than DA at the hub, so the RT @ Hub PPA price is about \$3.9 less than the DA @ Hub price. MISO's difference is negligible (~\$0.3), so RT vs DA pricing there is almost the same. (In real operations, RT prices are more **volatile** even if the average is similar molecule.io, but here we only adjust for the mean bias. Additional volatility risk could be handled via a volatility discount or shaping factor, but we assume the forward already captures expected value and we handle risk via P-level discounts.)

3.4 P-Level Risk Discounts (P50, P75, P90)

Finally, we incorporate **generation risk** into the fixed price. PPA pricing often uses **P50 vs P90** outputs to account for uncertainty in annual energy: P50 is the expected (50% exceedance) production, P90 is a downside case with 90% exceedance probability (i.e. only 10% chance of falling below that output) greensolver.net. A P90 energy yield is lower than P50, reflecting inter-annual resource variability and model uncertainties greensolver.net. Offtakers or financiers may prefer prices based on P75 or P90 to ensure a cushion – effectively paying a lower price per MWh, but with more confidence that the project will meet or exceed that MWh volume.

Approach: We apply a **risk discount** to the P50 price proportional to the reduction in expected energy from P50 to P75/P90. In other words, if a project's P75 energy is 97% of its P50 energy, we might offer ~97% of the price (a 3% discount) so that the total revenue expected at P75 equals the revenue at P50. This ensures the seller has a ~75% chance to

meet or exceed the revenue target. *This is akin to the “probability of exceedance” method – using a lower production volume to set a safer price*greensolver.net. It addresses **volume risk** (resource uncertainty year-to-year) and provides a buffer for other risks.

From the 3-year sample, we estimated the variability of annual generation:

- The wind farms showed about 5% standard deviation in annual output (coefficient of variation ~5%). This implies the **wind P75** (roughly 1 standard deviation above P50 in exceedance terms for a normal approximation of annual variations) might be around 97–98% of P50. Indeed, using a simple analysis, we found ~2–3% reduction for wind P75 vs P50. Wind P90 could be on the order of ~90–95% of P50 (though with only 3 years data this is uncertain; typically P90 for wind is ~10-15% below P50 for a single yeargreensolver.net).
- The solar farm had very low inter-annual variability in those 3 years (as expected, solar is more predictableember-energy.org). P75 for solar was ~99% of P50 (only ~1% lower). So the risk discount for solar is minimal at P75. Even P90 might be ~95–98% of P50 for solar given resource stability.

For our **P75 scenario** (often a balance between too aggressive P50 and very conservative P90), we applied approximately a **2–3% price discount for the wind PPAs and ~1% for the solar PPA**. In dollar terms, for ERCOT wind: base \$51.9 * (1 – 0.973) ≈ –\$1.39, so about \$1.4/MWh off. MISO wind: \$44.3 * (~2% risk) ≈ –\$0.94. CAISO solar: \$44.3 * (~1%) ≈ –\$0.40. These discounts are subtracted from the price after other adjustments. If we were to do a P90, we would use a larger discount (perhaps ~8–10% for wind, ~3–5% for solar), resulting in a lower fixed price that provides a 90% confidence level of achieving that revenue.

The **P50 case** would have no discount (it’s just the raw generation-weighted price). We can thus produce a range of fixed prices: higher for P50 (risk-neutral), lower for P75, even lower for P90 (more risk-averse). Typically, **P50 is used by equity investors (average expectation), P90 by lenders (worst-case for debt service), and P75 is sometimes used in negotiations as a middle ground**greensolver.net.

4. Fair PPA Price and NPV Calculation

With all components above, we can calculate the fixed PPA price for each scenario (market & location) and each risk level. The **fair PPA price** is essentially the price that makes the expected value of the PPA equal to the project’s expected revenue (or cost) over the term. In our case, since we’re basing it on expected market value, the P50 hub price (~\$52 for ERCOT wind, etc.) is the fair price that yields NPV=0 against the merchant value (before risk adjustments). Once we include risk discounts (P75/P90) to protect the buyer, the PPA price is lower – that difference is effectively the risk premium that the seller “pays”.

We compute the Net Present Value of the energy revenues for each case to ensure comparability. Using a 7% annual discount rate (typical for a renewable energy project's hurdle rate), we discount the monthly cash flows from 2026–2030:

- Monthly cash flow = (Fixed PPA price) × (projected generation that month).
- Discount factor = $\frac{1}{(1+0.07)^t}$ for mid-month of that month t in years (we assume monthly compounding negligible and just apply annual discount pro-rated).

Summing the discounted cash flows gives the NPV of revenue. For the P50 fair price at hub, this NPV equals the NPV of selling into the market (by construction). For the P75/P90 prices, the NPV is a bit lower (reflecting the risk taken by the seller). We also look at the total 5-year revenue (undiscounted) as a check.

Fair Price Range (P50–P90): Considering P50, P75, P90:

- **ERCOT Wind:** P50 @ Hub ~\$51.9/MWh (as above). P75 @ Hub ~\$50.5 (about 2.7% lower). P90 @ Hub could be in mid-\$40s (perhaps ~\$46 if ~10% discount). At the bus, these become roughly ~\$42 (P75) and high-\$30s (P90) given the –\$9 basis.
- **MISO Wind:** P50 @ Hub ~\$44.3. P75 @ Hub ~\$43.3. P90 maybe around ~\$40. Bus prices are dramatically lower (~\$30 at P75 as seen below) due to basis.
- **CAISO Solar:** P50 @ Hub ~\$44.3. P75 ~\$43.9 (almost the same, ~1% off). P90 maybe ~\$42. Bus is essentially same as hub (~\$44).

Below is a summary of the **P75 risk-adjusted fixed prices** we derived for each asset and scenario, with component breakdown:

Asset (Type)	Contract	Hub Fwd (P50)	Basis Adj.	DA–RT Adj.	P75 Risk Disc.	P75 Fixed Price
ERCOT Wind	DA @ Hub	\$51.92	+\$0.00	+\$0.00	–\$1.39	\$50.53
ERCOT Wind	DA @ Bus	\$51.92	–\$9.68	+\$0.00	–\$1.39	\$40.86
ERCOT Wind	RT @ Hub	\$51.92	+\$0.00	–\$3.92	–\$1.39	\$46.61

Asset (Type)	Contract	Hub Fwd (P50)	Basis Adj.	DA–RT Adj.	P75 Risk Disc.	P75 Fixed Price
ERCOT Wind	RT @ Bus	\$51.92	–\$11.36	–\$3.92	–\$1.39	\$35.25

MISO Wind	DA @ Hub	\$44.25	+\$0.00	+\$0.00	–\$0.94	**\$43.31**
MISO Wind	DA @ Bus	\$44.25	**–\$12.85**	+\$0.00	–\$0.94	**\$30.46**
MISO Wind	RT @ Hub	\$44.25	+\$0.00	**–\$0.26**	–\$0.94	**\$43.06**
MISO Wind	RT @ Bus	\$44.25	–\$12.14	–\$0.26	–\$0.94	**\$30.91**

CAISO Solar	DA @ Hub	\$44.32	+\$0.00	+\$0.00	–\$0.40	**\$43.92**
CAISO Solar	DA @ Bus	\$44.32	**+\$0.11**	+\$0.00	–\$0.40	**\$44.03**
CAISO Solar	RT @ Hub	\$44.32	+\$0.00	**–\$1.65**	–\$0.40	**\$42.26**
CAISO Solar	RT @ Bus	\$44.32	–\$0.05	–\$1.65	–\$0.40	**\$42.22**

Table: Fixed PPA prices for each asset under P75 (75% probability) case, with components. “Hub Fwd” is the generation-weighted hub price (P50). “Basis” is bus minus hub (so negative here means subtract). “DA–RT” is the adjustment for real-time settlement (negative if RT < DA). “Risk Disc.” is the P75 discount (negative). The final P75 price is sum of these.

For example, **ERCOT Wind Day-Ahead Bus**: Starting from \$51.92, add basis –\$9.68 (so subtract 9.68) and subtract \$1.39 risk → **\$40.86/MWh**. The **most conservative price** above is ERCOT Wind RT @ Bus ~\$35.3, which reflects stacking a large negative basis and the RT penalty. In contrast, the **CAISO solar** project’s bus price is roughly the same as hub ~\$44, because its basis is near zero and its output is very predictable (small discount). The MISO wind’s bus prices (~\$30) are extremely low due to its heavy negative basis – a corporate offtaker buying a “busbar, as-produced” PPA from that project would justifiably insist on a deep discount to compensate for frequent negative local prices.

These fixed prices can be interpreted as **fair PPA offer prices** for a 2026–2030 contract under the given assumptions. For instance, offering the ERCOT wind farm a ~\$50.5/MWh fixed price (DA @ Hub, P75) would give ~75% confidence that the project’s actual generation *at the hub* will earn at least that (the project is likely to curtail in the worst negative-price hours to protect that margin). If the offtaker instead settles at the bus, they’d insist on closer to \$40.9 to account for the node’s lower prices. These prices are purely for energy; they do not include Renewable Energy Credits or capacity value, which could be additional value streams.

Finally, using a 7% discount rate, we find the NPV of revenues if the project goes merchant vs enters each PPA. In the P50 hub case, the NPV(merchant energy) \approx NPV(PPA) by design (around zero difference). In the P75 cases, the NPV to the project is lower (the buyer captures the difference as risk value). For example, for ERCOT wind, the 5-year energy volume ~ 1.1 – 1.2 million MWh; at \$50.5 that's \sim \$56 million nominal revenue, which discounted at 7% gives on the order of \sim \$47 million NPV. At P50 price \$51.9, NPV would be slightly higher (\sim \$49m), so the \sim \$2m difference is essentially the value of the risk mitigation to the buyer.

5. Discussion: Risk Management and Strategic Considerations

Volume and Price Risk: Our model addresses **volume risk** through P-level adjustments and **price risk** via fixed pricing and basis/market adjustments. By using P75/P90 pricing, we acknowledge the uncertainty in yearly generation – the offtaker pays less per MWh to ensure they aren't overpaying in a low-generation year greensolver.net. Price risk (market volatility) is mitigated by the PPA itself (the fixed price shields both parties from wholesale price swings). However, some price risks remain: if the PPA is hub-settled, the seller still bears basis risk (difference between their node and hub); if it's bus-settled, the offtaker bears the full brunt of local price risk. We included basis adjustments so whichever side holds that risk is compensated. Additionally, real-time vs day-ahead volatility is a risk – real-time prices can spike or crash unpredictably molecule.io. If a project cannot perfectly stick to day-ahead schedules, it may face imbalance costs; our modest RT discount approximates this risk. In practice, more sophisticated stochastic modeling (e.g. Monte Carlo on prices and generation) would be used to quantify these risks and possibly add a further risk premium on the fixed price.

Impact of Negative Prices: Negative price events significantly affect revenue if not managed. For a merchant plant, generating during negative prices means paying to sell power – eroding profits. In our analysis, we assumed the project would curtail at negative prices (no revenue rather than negative revenue) pexapark.com. This improves the captured price and is a logical strategy if the plant has the right (or obligation) to shut off when prices are below zero. Under many modern PPAs, **if the buyer does not take energy during negative price periods, the seller is incentivized to curtail** to avoid losses pexapark.com. We accounted for this by effectively treating those hours as zero generation in our revenue model. The result is slightly lower total energy delivered (e.g. the MISO wind's annual output might drop a few percent because we assume it shuts off during the 8% of hours that were negative DA prices), but the average selling price goes up (since we excluded negative prices from the average). **If we did not curtail** (i.e. forced the plant to generate through negative prices), the PPA price required would actually be lower because the

project's average revenue would be dragged down by negative prices – or alternatively, the project would lose money in those hours unless the PPA had a price floor. Thus, curtailment at negative prices is a key assumption to protect value in markets like MISO and CAISO where negative prices are frequent.

PPA Customer Not Taking Negative-Price Power: This scenario is exactly what we modeled – the PPA has a clause that if spot prices are negative, the offtaker's obligation to pay is suspended (a **\$0/MWh floor**). Many corporate PPAs have such clauses now, often with a cap on the number of negative-price hours they won't pay for pexapark.com. We assumed a full no-pay for all negative hours, which is conservative for the seller. The result is the seller curtails output during those hours (since they won't get paid) pexapark.com. Financially, this shifts the risk of negative prices entirely to the project. The offtaker is protected (they never have to “pay to take power”), but the project's PPA price will be lower to compensate for this risk. In our results, the PPA prices especially for MISO wind are indeed much lower partly because of this – the project has to account for losing revenue in those negative-price intervals. If instead the offtaker agreed to pay even during negative prices (some PPAs have full compensation or a partial price floor pexapark.com), the fixed price could be higher since the project is guaranteed revenue for all production. In short, **we handled the negative price clause by assuming no revenue during negative hours**, which lowers overall revenues but avoids actual negative cash flow.

Hedgeability by Market: Some markets and projects are more “hedgeable” than others. Key factors are the volatility of prices, the prevalence of basis risk, and the reliability of the resource. From our analysis:

- **CAISO Solar** is relatively easy to hedge (from an offtaker perspective) because its node price is very close to the hub (negligible basis risk) and the solar output is highly predictable year-to-year (low variability). The biggest challenge in CAISO solar is the **cannibalization risk** – solar-heavy hours having low or negative prices – but a buyer can mitigate that by negotiating negative-price clauses (as we did) or adding storage.
- **MISO Wind** is the hardest to hedge here: it has a large basis difference (meaning the project's local price can diverge a lot from hub) and frequent negative prices. An offtaker settling at the hub would leave the project exposed to huge basis losses; settling at the bus transfers that risk to the offtaker but they then demand a much lower price (we saw ~30% lower). So either the project or offtaker will bear that risk. Hedging this completely would require financial transmission rights or other basis hedges, which aren't always perfect. Thus, a project in a congested area is less hedgeable – the PPA prices will reflect a lot of risk premium.

- **ERCOT Wind** is intermediate. ERCOT has price spikes and some basis issues in certain areas. Our ERCOT wind had a notable basis (~\$10) in the Panhandle or West Texas perhaps. ERCOT's market is also energy-only with high volatility (e.g. occasional \$9,000/MWh spikes). While our forward curve and P75 discount implicitly handle typical volatility, extreme events are hard to hedge. ERCOT's hub price can diverge from node if there's local congestion (as seen in Aug 2019 where a project could lose money if hub vs node differed during a spike[projectfinance.law](#)). In general, a **diversified hub (like ERCOT North)** is easier to hedge than a single-node PPA because hub prices are more stable[projectfinance.law](#). Projects in ERCOT often prefer hub-settled PPAs to avoid basis risk, even though that means project takes basis risk; some are now pairing with batteries or other means to manage it[projectfinance.law](#).

So, markets with **less congestion and more stable pricing (or where your project is at a strong node)** are more hedgeable. CAISO's major hubs and ERCOT's North hub are relatively liquid and robust. MISO, depending on location, can have serious constraints. **Liquidity of forward markets** also matters – ERCOT and MISO have reasonably liquid forward markets for power; CAISO does too. But nodal liquidity is low – hence why hub settlement is common and basis risk is a concern[projectfinance.law](#).

Merchant vs PPA: Would staying merchant be better for any of these projects? It depends on risk appetite:

- In a pure expected value sense, our P50 analysis suggests the *average* revenue is the hub forward price. If the project is comfortable with volatility and has no financing constraints, it might earn more by staying merchant if prices turn out favorably (and indeed avoid giving up the P75/P90 discount). For example, the ERCOT wind has upside in scarcity events – a merchant plant might capture a few hours of extreme prices that a fixed PPA would miss. If one believes prices will exceed the current forwards or that basis congestion will improve, merchant could yield higher revenues than the conservative PPA price.
- However, **merchant exposure carries substantial risk**. Lenders usually won't finance a wind/solar project without a PPA or hedge because the downside (prices collapse, or generation coincides with low prices) could undermine revenue. The PPA fixes the revenue to a more certain level. In our results, for instance, the MISO wind's P75 bus PPA is only ~\$30/MWh – quite low. If the project stays merchant, could it beat \$30? Possibly, especially if transmission upgrades or retirements improve MISO congestion. But it could also do worse if negative prices persist. So merchant might be “better” in expected terms if one is optimistic, but it's far riskier.

- Some markets like ERCOT have seen merchant wind and solar struggle when there's too much supply (negative price times). The question is also contract length: a five-year forward might not fully cover project life. Sometimes projects go merchant for a while hoping for better PPA terms later – but that's speculation.
- **Hedging alternatives:** One could pursue financial hedges (like bank hedges or futures) instead of a traditional PPA. Those come with their own risks (e.g. margin requirements). Our fixed price essentially mirrors a contract-for-differences style hedge at a fixed strike.

In summary, unless one expects significant market improvements, a PPA provides revenue certainty that is usually necessary for project viability. The exception might be if PPA prices are so low (due to risk discounts or lack of interested buyers) that a developer might delay or opt to accept merchant risk in hopes of higher future prices or basis improvement.

Potential Data/Model Improvements: We used a relatively simple deterministic approach. Several enhancements could improve accuracy and insight:

- **Longer historical records or reanalysis:** Using 10+ years of weather data would improve estimates of P90/P99 output and capture resource variability more robustly greensolver.net. Three years is short; a poor wind year might drop output ~10% below mean, which we barely captured. More data would refine the P-level discounts.
- **Hourly Simulation:** Instead of monthly average pricing, an hourly simulation of 2026–2030 (stochastically or using historical patterns scaled to forwards) could capture the covariance of generation and price. This would let us explicitly model negative price curtailment impacts, price spike events, etc., and compute distribution of revenues. It would answer questions like “what’s the probability revenue falls below X?” more directly.
- **Correlation of basis with generation:** We assumed a constant average basis. In reality, basis blows out specifically when the project generates a lot (e.g., wind congestion). A more granular model would reduce price during high-gen hours more than low-gen hours. This could be done by modeling a price shape at the node (e.g., hub price minus a congestion adder that is a function of regional wind output). With that, one could better estimate how much revenue is lost to basis during peak output. Similarly, modeling DA vs RT with volatility (instead of a fixed adjustment) would allow evaluating imbalance risk (e.g., a sudden outage or forecast error causing deviation).

- **Alternative risk metrics:** We did a simple P75/P90 adjustment. One could use Monte Carlo draws of yearly generation and price to directly compute the distribution of project cash flows and then find the fixed price that equates to, say, 75% probability of exceeding a target NPV. This would incorporate price uncertainty as well, not just volume.
- **Additional value streams:** Including potential **capacity market payments** (for MISO) or **Renewable Energy Credit (REC) value** could increase the “fair value” and perhaps allow a higher PPA price or at least explain the full project value. We ignored these here.
- **Scenario analysis:** We could test sensitivity to, say, gas prices or demand changes by altering the forward curve, to see how the fixed price might differ in higher-price or lower-price scenarios.
- **Granular peak definition adjustments:** We roughly aligned with standard blocks. If one wanted to refine, incorporating actual 2026–2030 holiday calendars (to adjust which hours are off-peak) or using 8760 hourly shapes for forward prices would make the weighting more precise. (Our use of monthly peak/offpeak forwards is a limitation since within those categories price isn’t uniform in reality. An 8760 forward price curve would allow exact hour-by-hour revenue calc.)

Despite these possible improvements, our integrated model provides a **transparent, reproducible framework** to go from historical data to a risk-adjusted PPA price. All calculations (from reading data, computing monthly profiles, applying forward prices, adjustments, and producing Excel outputs and plots) are contained in the Python script, which is well-documented for clarity. This ensures that stakeholders can trace every assumption – from why MISO’s PPA price is so low (answer: huge negative basis and lots of wind at low-price times) to how a 7% discount rate over 5 years affects NPV. By adjusting inputs (e.g. using P90 instead of P75, or assuming no negative price curtailment), the model can quickly show the impact on the “fair” fixed price, helping decision-makers evaluate the trade-offs between merchant risk and PPA certainty.

In conclusion, the model demonstrates a methodical approach to **forecasting output** and **pricing a PPA** that accounts for time-of-day/seasonal generation patterns, forward market conditions, locational price differences, and risk tolerances. Such analysis is crucial for both project developers and offtakers to strike a deal that reflects the true value and risks of a renewable energy project in different markets. The results highlight that not all electrons are equal – *when* and *where* energy is produced can drastically change its worth, and appropriate risk-adjusted pricing is key to a sustainable PPA for both sides.

Sources:

- Greensolver (2020) – *P50/P90 definitions and importance of considering P90 for risk*greensolver.netgreensolver.net
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