

BENEATH THE SURFACE: Q2-TRAK ILLUMINATES WELL PERFORMANCE WITH DATA-DRIVEN ANALYSIS

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INTRODUCTION: A FRAGMENTED VIEW OF PERFORMANCE

Rod lift is the most widely applied artificial lift method in unconventional resource development. Ensuring long-term performance and reliability requires continuous monitoring and timely optimization decisions. Production data plays a critical role in this optimization effort by guiding intervention timing, supporting equipment selection, and influencing cost management strategies. The challenge is that the value of this data depends on the ability to interpret it alongside run-life results, interventions, and cost impacts. Yet most existing systems still treat these datasets independently, leaving important performance connections unexplored.

This paper examines how integrated visibility can reduce uncertainty and strengthen decision-making in rod pump operations. It introduces a platform designed to combine production outcomes, equipment performance history, and financial records into a single environment, allowing operators to evaluate artificial lift effectiveness with greater clarity. To demonstrate the importance of this capability, the paper reviews the limitations of existing systems, outlines a unified approach to performance evaluation, and presents a field-based case study showing how access to linked data supports optimization across wells.

PROBLEM DEFINITION: DISCONNECTED DATA LIMITS PERFORMANCE EVALUATION

Rod pump operations generate significant volumes of production, equipment, and intervention data, from rate trends and fluid levels to run-life history and failure comments. Yet, this information is commonly distributed across multiple systems and departments: production data with operations, equipment history with engineering, and cost records with accounting. When these critical datasets remain disconnected, the result is delayed, reactive, and assumption-based decision-making. The challenge becomes even more pronounced when operators attempt to evaluate the true impact of specialty components. Without a direct, integrated view linking production behavior to equipment performance and financial outcomes, operators must rely on indirect indicators such as gradual production changes, pump pulls, or maintenance reviews long after the fact. These traditional methods keep essential insight in the dark, leaving uncertainty around whether the right equipment was selected or whether performance improvements are being achieved.

The industry needs a way to illuminate well performance as it evolves, capturing how production, run-life, and spending move together so that optimization decisions can be made with confidence. Operators require access to data that not only shows what is happening to their wells, but why. However, despite advancements in digital monitoring, most systems still leave critical aspects of well performance in the dark, limiting evaluation of artificial lift performance.

STILL IN THE DARK: VISIBILITY GAPS IN EXISTING SYSTEMS

Efficient production data analysis plays a critical role in optimizing artificial lift performance, improving intervention timing, and ensuring that equipment investments deliver measurable value. However, most existing systems focus on diagnosing isolated issues without providing a complete view of well behavior or connecting technical performance to operational and financial outcomes.

Moises et al. (2010) introduces a failure diagnostic system that relies on teardown inspections, expert evaluation, and structured classification methods to determine why rod pump components fail. While this approach improves how failures are recorded and understood, the insights only emerge after the pump has already been pulled, meaning operators still lack visibility into specialty component behavior while the well is producing. This reactive workflow leaves a critical performance gap: operators cannot evaluate whether the equipment they install is improving production or run-life in real time.

Hidayat et al. (2023) demonstrate that real-time IoT monitoring can increase well uptime and reduce field interventions through automated control. Although this system improves awareness of operational status, it does not provide specialty-component insight or link rod-pump equipment selection to production outcomes over time. Key performance interpretation remains separated from the equipment responsible for delivering that performance.

Sindi et al. (2023) apply nodal analysis, ESP modeling, and machine learning to predict equipment degradation before failure occurs. This enhances proactive diagnostics for ESP lift, but evaluation remains focused on pump health rather than its impact on production efficiency or cost performance. Additionally, the solution targets ESP systems, leaving a gap in rod-lift operations where most artificial lift wells reside.

Across these studies, three critical limitations persist:

- **Fragmented data** — Production, run-life, and cost information remain separated across systems, preventing a unified performance view.
- **Limited ability to validate specialty component performance while producing** — Insights often emerge only after failure or pump pulls, leaving uncertainty during active operation.

Integrated Data Engine

Q2-Trak consolidates publicly available production rates with internal records, including pump installations, run-life outcomes, and documented failure mechanisms. This unified context enables operators to directly connect production behavior to the equipment installed in the wellbore and the conditions in which it is operating (Figure 2).

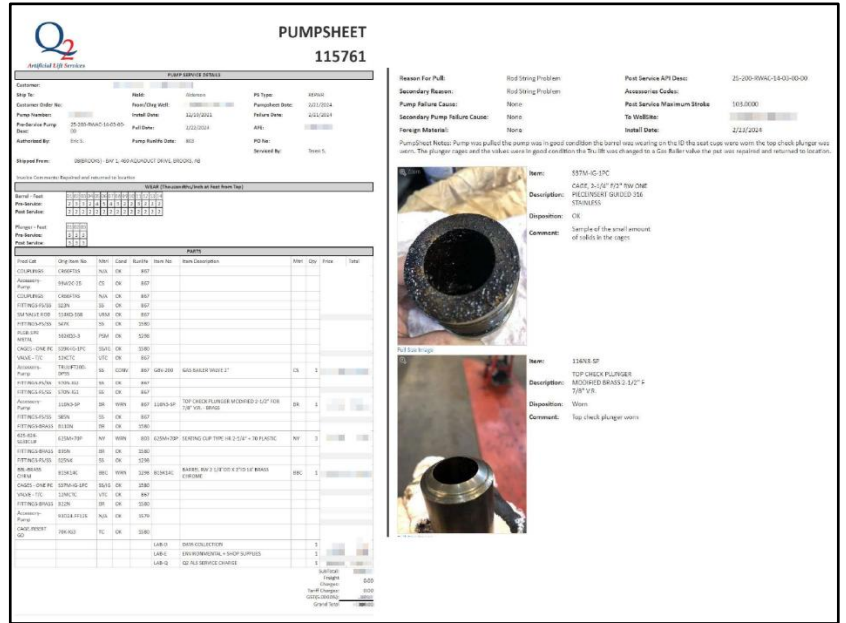


Figure 2: Example pump sheet from Q2-Trak, linking installed components and failure observations, including visual evidence, to support equipment-performance evaluation.

Performance Tracking

The platform presents over a terabyte of production data in consistent and comparable formats, enabling pre- and post-installation trend evaluation for specialty components (e.g., [Q2 Flow](#), [HVS Valve](#), [WhaleShark](#) and [SharkTAC](#)). Users can confirm run-life improvements, validate production gains, and assess equipment selection choices using actual field data rather than assumptions or delayed physical inspections. This enables real-time evaluation of artificial lift performance throughout the run life of a component, as illustrated in the production trend dashboard shown in Figure 3.



Figure 3: Pump Metrics Dashboard in Q2-Trak showing time-aligned oil, water, and gas production with pump history overlays for a single well.

Real-Time Invoicing and Cost Visibility

As soon as a pump is entered into the system, cost records and well dashboards update automatically. Figure 4 illustrates how this real-time connection between equipment performance and financial impact allows operators to evaluate whether investments are delivering value and informs future equipment selection.

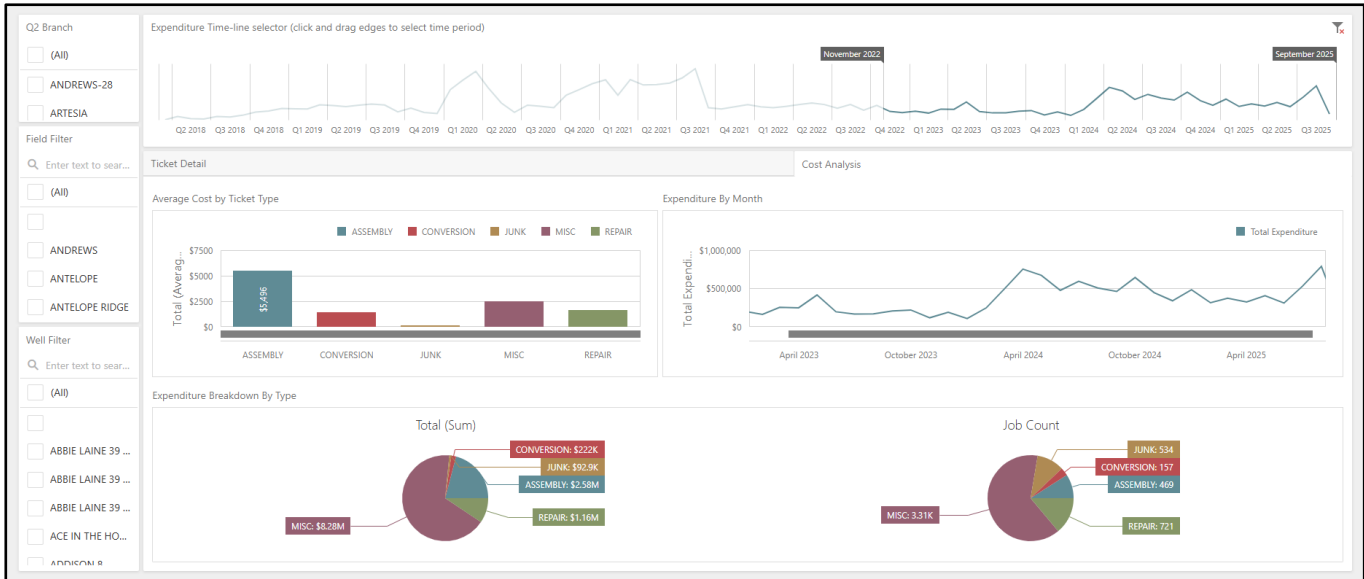


Figure 4: One of Q2-Trak's financial dashboards displaying expenditure trends, average cost by ticket type, and job count distribution, supporting real-time visibility into operational costs.

Cross-Functional Collaboration

Q2-Trak provides common visibility to customers and internal Q2 ALS teams, including engineering, operations, sales, IT, and accounting, ensuring decisions are based on shared, up-to-date information. This transparency fosters clearer communication, faster issue resolution, and more reliable decision-making across the artificial lift workflow.

Together, these capabilities transform fragmented information into meaningful operational understanding. Replacing uncertainty with visibility and bringing clarity to areas of well performance that have long remained unseen.

Limitations & Considerations


Public production data includes a three-month reporting delay, meaning recent well activity may not immediately appear in Q2-Trak. While the platform provides structured access to production and equipment information, engineering judgment remains essential when evaluating performance outcomes. Artificial lift performance is influenced by reservoir factors, wellbore condition, and surface constraints beyond component selection, and these must be considered in assessment. As Q2-Trak continues to expand its data capabilities, system-supported performance evaluation will only evolve.

BUILDING THE VIEW: STUDY METHODOLOGY

A retrospective performance analysis was conducted for Company X to evaluate the impact of various traveling and standing cage designs installed in rod-lift wells. The objective of the study was to quantify production behavior and reliability trends before and after specialty component installation using existing operational data.

Q2-Trak was first used to extract historical records for pumps installed within the study region. The exported dataset included installation dates, invoice details, product descriptions, failure mechanism comments, and well identifiers such as API or UWI numbers. These identifiers were then provided to Enverus, who matched the equipment history to publicly available production data for the same wells. The merged dataset was delivered through Power BI and exported to Microsoft Excel for analysis, as outlined in Figure 5.

FROM Q2-TRAK → ENVERUS → POWER BI → EXCEL



Branch	API_UWI	Well	Product Category	Part Description	Pump Sheet #	Service Date	ServiceMonth	MaxServiceDate	MonthOffset	ProducingMonth	Prod_BOE	Prod_MCFE	LiquidsProd_BBL	GasProd_MCF
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-10	3/1/2021	16	96	16	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-9	4/1/2021	24	144	24	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-8	5/1/2021	22	132	22	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-7	6/1/2021	20	120	20	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-6	7/1/2021	23	138	23	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-5	8/1/2021	21	126	21	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-4	9/1/2021	21	126	21	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-3	10/1/2021	21	126	21	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-2	11/1/2021	20	120	20	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	-1	12/1/2021	50	300	50	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	0	1/1/2022	15	90	15	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	1	2/1/2022	18	108	18	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 175470	1/18/2022	1/1/2022	1/18/2022	2	3/1/2022	4	24	4	1/0/1900		
42-003-37626	MABEE RANCH 1	CAGE-STAINLESS	OVERSIZE,CAGE CLOSED BAR 175470	1/18/2022	1/1/2022	1/18/2022	3	4/1/2022	36	216	36	1/0/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-10	3/1/2021	248.67	1492	179	2/21/1901		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-10	4/473	120.67	724	69	11/5/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-10	4/462	92.67	556	66	6/8/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-9	4/1/2021	112	672	112	1/0/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-9	4/474	74.67	448	47	6/14/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-9	4/473	150.17	901	99	11/2/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-8	5/1/2021	126.67	760	99	6/14/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-8	4/485	100.33	602	77	5/19/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-8	4/473	120.67	724	69	11/5/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-7	6/1/2021	102.67	616	71	7/8/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-7	4/485	78.33	470	65	3/20/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-7	4/474	74.67	448	47	6/14/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-6	7/1/2021	112	672	75	8/9/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-6	4/486	66.83	401	61	2/4/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-6	4/485	100.33	602	77	5/19/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-5	8/1/2021	109	654	67	5/11/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-5	4/486	142.83	857	100	9/13/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED BARREL 1294543	3/14/2023	3/1/2023	3/14/2023	-5	4/485	78.33	470	65	3/20/1900		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED DOUBLE VALV 177914	1/28/2022	1/1/2022	1/28/2022	-4	9/1/2021	168.67	1018	96	3/17/1901		
42-003-40043	UNIVERSITY 5036 1	CAGE-STAINLESS	CAGE, CLOSED PIN END PLUN 309710	5/30/2023	5/1/2023	5/30/2023	-4	4/4927	42	252	35	2/11/1900		

Figure 5: Data workflow used in this study: records were extracted from Q2-Trak, production data was aligned in Enverus and Power BI, and the merged dataset was exported to Excel for analysis. The standardized Month Offset index (highlighted) enables aligned pre- and post-installation performance comparison across wells.

The combined dataset contained time-aligned production metrics including total liquid rate, oil and gas volumes, water cut, and flowing behavior indicators. Each well record also included a standardized Month Offset value, where zero represents the installation month of that specific cage. Negative values correspond to well production before installation, and positive values reflect production while the new cage was in operation. Most wells contained at least nine months of data before and after installation, providing a meaningful period for evaluation.

Production trends were evaluated in Excel using pivot tables (Figure 6). Liquid production in barrels per day was plotted for each cage type and normalized to account for differences in sample size and reservoir variability. Trendlines were applied to observe shifts in well

performance following component installation. To assess product prevalence and ensure appropriate weighting in interpretation, cage count distributions were also generated using pivot tables. Pie charts (Seen in Figure 10) illustrated the proportional use of each cage design within each operating district, helping contextualize performance results relative to deployment frequency.

Row Labels	Average of PDLiquids, BBLPerDAY	Count of DLiquids, BBL PerDAY	Row Labels	Average of PDLiquids, BBLPerDAY	Count of PDLiquids, BBL PerDAY	Row Labels	Average of PDLiquids, BBLPerDAY	Count of PDLiquids, BBL PerDAY
# CAGES - ONE PC	30.8593366	411	# CAGES S/REG	27.78971831	2538	# CAGES REGULAR	10.77395	200
-10	37.87925926	27	-10	35.10044775	134	-10	11.345	10
-9	34.75962963	27	-9	34.31477612	134	-9	10.945	10
-8	29.57185185	27	-8	32.72	134	-8	12.169	10
-7	30.89148148	27	-7	31.19469697	132	-7	11.909	10
-6	31.0762963	27	-6	29.33307087	127	-6	11.131	10
-5	30.23259259	27	-5	28.82837209	129	-5	11.723	10
-4	29.73703704	27	-4	29.71153846	130	-4	11.465	10
-3	28.83148148	27	-3	28.89827058	133	-3	10.505	10
-2	29.20962963	27	-2	28.98897838	127	-2	11.194	10
-1	26.73333333	27	-1	25.85763359	131	-1	10.33	10
0	29.24730769	26	0	25.32265625	128	0	10.032	10
1	30.2352	25	1	25.84874016	127	1	11.877	10
2	28.15052632	19	2	24.39976	125	2	10.446	10
3	32.18882353	17	3	25.06365079	126	3	10.267	10
4	32.42666667	15	4	23.74931496	127	4	10.015	10
5	26.63818182	11	5	23.54471074	121	5	10.228	10
6	31.26	9	6	25.39666387	119	6	10.059	10
7	37.57714286	7	7	25.33940171	117	7	10.128	10
8	29.38	6	8	25.82794872	117	8	10.027	10
9	28.60666667	6	9	24.67716667	120	9	9.276	10
# CAGE-OTHER	17.06036354	1898	# CAGE-MNL-HRDLND	13.52533762	311	# CAGE-STAINLESS	20.30756052	10162
-10	19.15075259	93	-10	13.55875	16	-10	24.49639525	536
-9	17.91882979	94	-9	13.72	16	-9	23.60824299	535
-8	17.57893617	94	-8	13.32875	16	-8	22.54682243	535
-7	17.47833333	96	-7	12.133125	16	-7	22.73076636	535
-6	17.57851485	101	-6	12.503125	16	-6	22.26900841	535
-5	18.27909091	99	-5	11.975	16	-5	21.35646729	535
-4	16.87348939	98	-4	11.96875	16	-4	20.39484112	535
-3	17.45126316	95	-3	11.95235294	17	-3	20.32162921	534
-2	16.14545455	99	-2	12.76705882	17	-2	19.70696629	534
-1	15.5821875	96	-1	13.10058824	17	-1	17.84433962	530
0	16.47257732	97	0	13.09235294	17	0	18.37485605	521
1	15.3859375	96	1	15.7325	16	1	19.62760314	509
2	16.8371875	96	2	14.654375	16	2	19.24860279	501
3	15.07880851	94	3	14.5175	16	3	19.02414634	482
4	16.26717391	92	4	16.194	15	4	19.0075052	481
5	15.94851064	94	5	14.65468667	15	5	18.78307056	471
6	16.18322581	93	6	14.63357143	14	6	18.72742004	469
7	18.26380435	92	7	13.88615385	13	7	18.68277056	462
8	17.3710989	91	8	11.69230769	13	8	18.98071895	459
9	18.6475	88	9	15.02153846	13	9	19.01604857	453
						Grand Total	21.04677176	15668

Figure 6: Pivot-table summary of average liquid production (BBL/day) by Month Offset and cage type, used to normalize production behavior across wells.

A complementary evaluation was performed on field-recorded failure comments to identify recurring issues by cage design. These comments, which include descriptions such as “beat-out,” “leaked,” “ball worn,” “cage worn,” and “valve worn,” were extracted from pump pull documentation and processed using Copilot to automatically categorize and quantify occurrences of these keywords. The results were then normalized as a percentage of total recorded comments for each cage type to support unbiased comparison of reliability across designs, as summarized in Figure 7.

Product Category	Count with Keywords	Total Comments	Percentage	Percentage (%)
CAGE-CRBNSTEEL	3	4	0.75	75
CAGE-MNL-HRDLND	16	21	0.7619	76.19
CAGE-OTHER	37	83	0.4458	44.58
CAGE-S/S INSERT	1	1	1	100
CAGE-SS-HRDLNED	5	17	0.2941	29.41
CAGE-STAINLESS	206	611	0.3372	33.72
CAGES - ONE PC	14	68	0.2059	20.59
CAGES MONEL IG	0	5	0	0
CAGES MONEL REG	15	36	0.4167	41.67
CAGES REGULAR	8	10	0.8	80
CAGES S/REG	80	144	0.5556	55.56

Figure 7: Summary table showing the percentage of failure-related comments associated with each cage type, reflecting relative reliability differences across configurations.

Together, these steps created a consistent framework for evaluating whether specialty cage designs influence production trajectory and component durability under real-world operating conditions. The next section presents the results of this analysis, including production performance outcomes and observed differences in failure-related comment frequency between cage types.

REVEALING PRODUCTION AND RELIABILITY TRENDS: RESULTS & INSIGHTS

The objective of this analysis was to compare Q2 ALS's one-piece Q2 Flow cage with several traditional cage designs in Company X's Permian Basin operations, using aligned production and reliability data (Figure 8).



Figure 8: Q2 Flow one-piece cage (left) versus a conventional one-piece cage (right), showing differences in internal flow geometry.

Production Analysis

Across 1,811 cage installations reviewed in Company X’s study area, roughly 22% utilized standard double-valve cage designs, which are known to increase pressure drop by up to 70% when paired with API standard balls. As reflected in the normalized production curves, wells operating with these traditional cage types generally showed a continued decline in liquid production after installation, indicating increasing flow restriction through the pump over time (Figure 9).

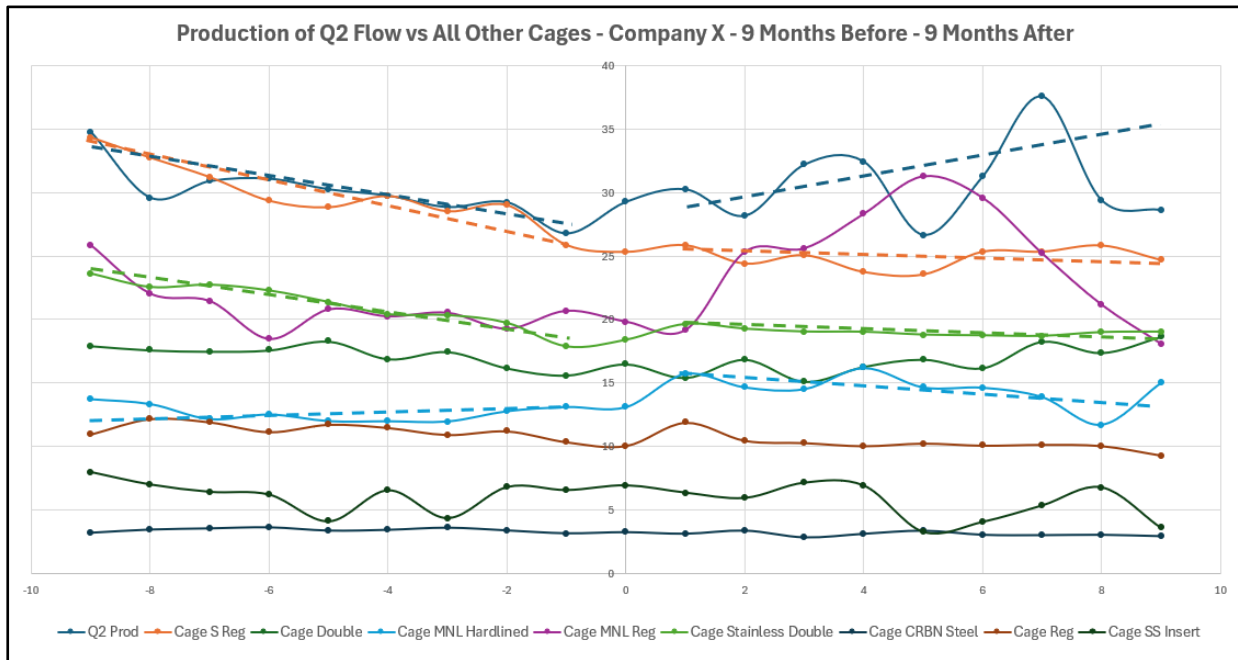


Figure 9: Normalized production trends for Q2 Flow (Dark Blue) and conventional cage designs across Company X wells.

In contrast, wells equipped with the Q2 Flow one-piece cage demonstrated a distinct positive shift in production trajectory following installation. Although fewer of these cages were deployed in this district (26 installations, Figure 10), the month-offset results showed a clear improvement in slope direction relative to baseline performance. Where other cage types continued gradual decline, the Q2 Flow curve recovered and increased after installation, indicating more stable deliverability under similar reservoir and operating conditions.

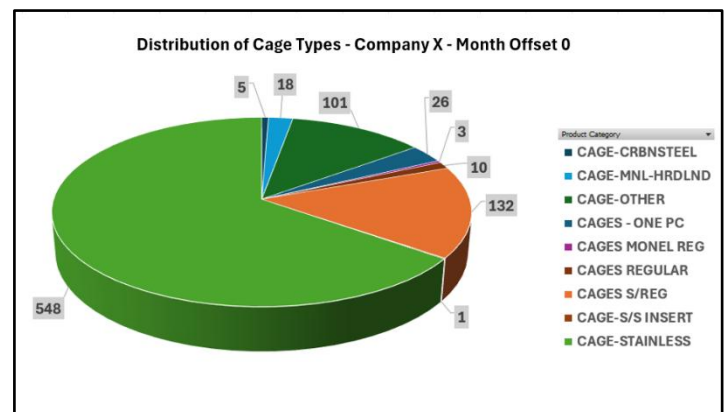


Figure 10: Distribution of cage types at Company X at installation month, including 26 Q2 Flow one-piece cages.

This measurable divergence in production behavior suggests that reduced pressure drop and improved flow efficiency associated with the Q2 Flow one-piece cage are reflected in real-world well performance. Even with a smaller deployment count, the strength and consistency of the upward trajectory position Q2 Flow as a notably higher-performing option compared to the standard configurations included in this evaluation.

Failure Analysis

Across all cage types evaluated, the Q2 Flow one-piece cage demonstrated the lowest incidence of reported cage- or valve-related issues, with only 21% of comments referencing these concerns. In comparison, several conventional configurations exhibited substantially higher failure-related percentages, including stainless-steel cages at 34%, and Monel regular cages at 42%, indicating more frequent wear- and seal-related challenges. Additional reliability outcomes for all cage designs are summarized in Figure 11.

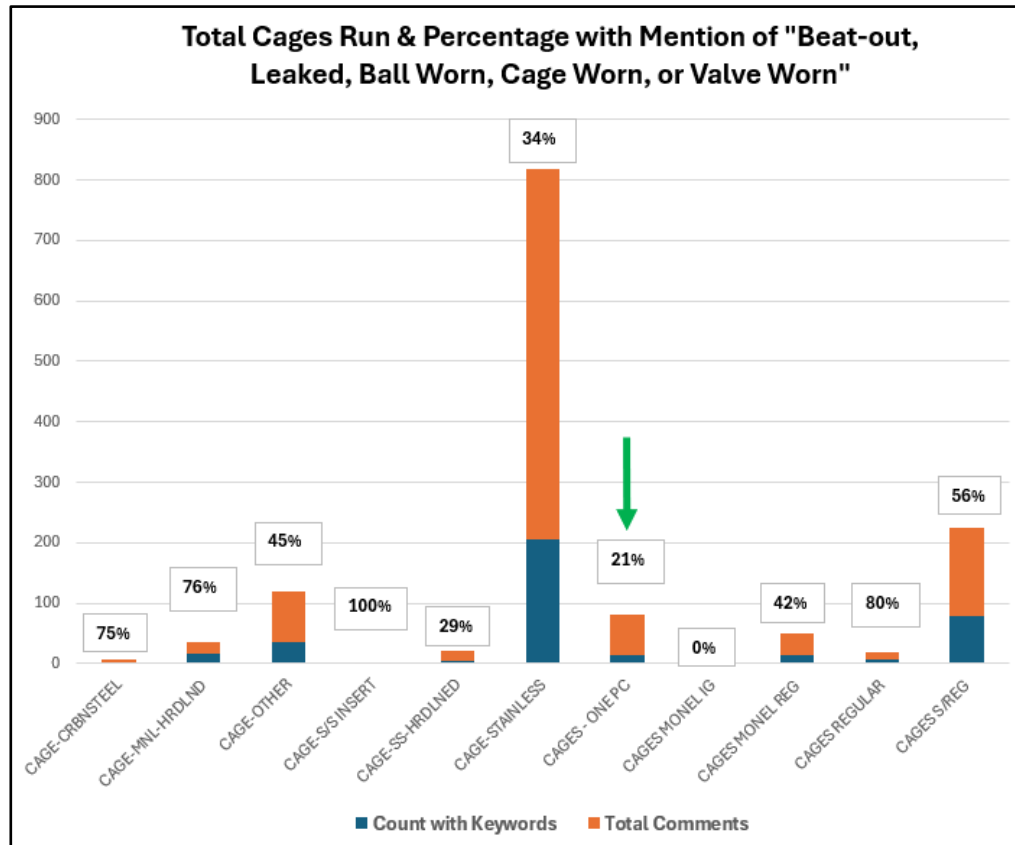


Figure 11: Percentage of failure-related comments by cage type, showing the Q2 Flow one-piece cage with the lowest incidence at 21%, compared to significantly higher rates in other configurations.

Despite its smaller deployment count within the dataset, Q2 Flow maintained both lower failure comment frequency and reduced severity of leakage and wear indicators relative to traditional designs. These findings reflect a meaningful reliability advantage, with fewer reported mechanical degradation issues during service life.

Interpretation

The alignment between improved production trajectory and reduced failure incidence suggests that the Q2 Flow cage supports both sustained deliverability and enhanced component durability under comparable operating conditions.

These insights were enabled through Q2-Trak's ability to align production history with equipment configuration and run-life documentation across Company X's asset base. By bringing these datasets into a single comparable structure, operators were able to verify value from specialty components using actual field performance rather than assumptions or delayed well interventions.

CLARITY BENEATH THE SURFACE: CONCLUSIONS

This case study demonstrated how integrated access to production behavior, equipment configuration, and run-life documentation can support clearer performance evaluation in rod-lift operations. By aligning existing operational data within Q2-Trak, Q2 ALS engineers were able to compare field performance across multiple cage designs for Company X, revealing measurable differences in both production trajectory and reliability outcomes.

These findings reinforce the importance of visibility when evaluating specialty artificial-lift components. Without unified access to production behavior during run life, operators are left uncertain whether equipment changes are truly improving outcomes. Q2-Trak addresses this gap by enabling decisions to be supported with field-based evidence rather than delayed diagnostics or assumptions. When performance clarity replaces operational uncertainty, operators gain confidence that their investments are working below the surface, even when they cannot see into the well directly.

The implications for the industry extend beyond this study. As datasets expand and platform adoption increases, artificial-lift decisions can become faster, more proactive, and more financially accountable. Continued development has the potential to incorporate greater automation and AI-assisted analytics, further enhancing performance benchmarking and supporting operational decision-making. These opportunities represent a path toward increasingly efficient optimization across wells.

By illuminating the connection between equipment selection and production outcomes, Q2-Trak brings long-observed well behavior into view and provides operators a clearer path toward sustained artificial-lift performance.

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