



DEMYSTIFYING HEALTHCARE CLAIMS ADJUDICATION: HOW AUTOMATION REDUCES ERRORS AND SPEEDS UP PAYMENTS

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Abstract

Healthcare claims adjudication can refer to the set of operations that make up the lifecycle of an insurance reimbursement claim: eligibility verification, benefits and pricing determination, and patient liability determination. Manual processes used by payers to support this lifecycle are generally inefficient and result in increased operational error, variable coverage decisions, and long payment turnarounds, all of which may harm payers, patients, and providers. Automation technologies, like rule-based processing engines, real-time eligibility checks, and predictive analytics, help make the claims process more efficient by using consistent rules, spotting mistakes as they happen, and speeding up payment. Interoperability technologies, using APIs, make sure that adjudication systems regularly check and update claims with enrollment files, provider directories, and Advanced anomaly detection features. These can improve just-in-time claims processing by not only checking claims against rules but also spotting fraud and cleaning up claims before they are submitted, helping to catch denials. By automating the claims process for everyone involved—providers, patients, and payers—those managing the revenue cycle can look forward to getting paid faster, having fewer claims denied, providing clearer financial information to patients, and reducing the costs of handling each claim, which makes smart adjudication systems essential for efficient revenue cycles in today's healthcare.

Keywords: Claims adjudication Automation, Real-Time Eligibility Verification, Predictive Analytics In Healthcare, Health Information Exchange Interoperability, Revenue Cycle Management

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1. Introduction

Claims adjudication is the process of determining whether a claim (the formalized healthcare bill) submitted by a healthcare provider to a healthcare payer on behalf of a patient satisfies the requirements of the applicable benefit plan, determining the extent of the payer's payment obligation, and making payment or denying the claim. It is one of the most processing-intensive aspects of the insurance reimbursement cycle. Claims adjudication remains a foundational function but has historically been a highly fragmented process, with manual review dependencies and high error rates across high-volume processing.

Today's reimbursement architecture is a complex web of interactions as claims pass through a collection of provider billing systems, clearinghouses, payer adjudication systems, pharmacy benefit management (PBM) networks, and other entities before payment is made. Each of these points of transition introduces friction due to data inconsistency, eligibility mismatch, or coding errors, which slow down the payments cycle and increase the costs of claims resolution. An abundance of Health Information Technology (HIT) studies have shown that manual adjudication increases administrative costs, and that rework and resubmission are large components of revenue cycle overhead [1].

More recently, the focus has shifted to automation as the silver bullet to eliminate the inefficiencies introduced by the model, with automated adjudication platforms using automation engines (rule-based validation engines, real-time eligibility, and smart error detection) to reduce processing delays and errors. Artificial intelligence and machine learning can be incorporated into these adjudication processes to support predictive fraud identification and ambiguous claim detection, proactive fraud detection, dynamic pricing verification, and more [2]. This would signify an important model shift in the way claims are evaluated, adjudicated, and communicated to others in the reimbursement process.

The article describes the complex healthcare claims adjudication workflow, automation architecture, and stakeholders' outcomes with clever adjudication systems. The article is divided into six major sections on the workflow and automation architecture, error detection and interoperability infrastructure, and strategies that have been realized through its utilization across the provider-payer-patient continuum (e.g., the supply chain of healthcare).

2. The Claims Adjudication Lifecycle: Structural Complexity and Process Challenges

The healthcare claims adjudication lifecycle encompasses all business, technical, and legal activities necessary to adjudicate a health claim from submission to disposition. A claim is submitted by a provider or pharmacy as an electronic healthcare claim standard to the payer (or its clearinghouse) in the 837P (Professional) or 837I (Institutional) formats promulgated by HIPAA. The claim itself is adjudicated through a series of reviews to determine the validity of the claim, the type or level of benefit eligibility, and any reimbursement. The use of health information technology in the claim adjudication process has the potential for important savings for the healthcare system, including reducing paperwork and its associated administrative costs from paper-based, labor-intensive claims processing [1].

Once submitted, the claim passes through an eligibility determination process where the payer system verifies that the patient was enrolled with an active benefit plan on service dates, that the rendering provider was credentialed within the applicable network, and that the service was a covered benefit. Reasons why claims fail eligibility include outdated enrollment data files, credentialing lapses, and demographic mismatches that require provider intervention before resubmission. Given the ability of information technology to improve the safety and reliability of clinical and administrative decision-making, eligibility verification should be data-driven and real-time, not performed after the fact in batches [2].

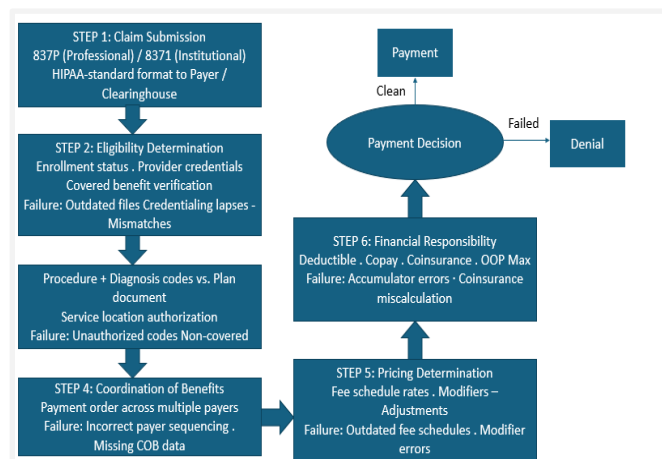


Fig. 1: Claims Adjudication Lifestyle [1, 2]

Benefit validation refers to the step in the process that determines whether the procedure and diagnosis codes are allowed by the plan document established for the service location. The benefit validation step also includes coordinating benefits with multiple payers so that each can determine order of payment and payment obligation. The pricing portion of the claim takes the assigned fee schedule rates or potential payment method and computes the allowed amount. The financial responsibility portion applies the payer payment obligation, patient deductible, copayment, and coinsurance to the allowed amount on the remittance advice to arrive at the net payment amount sent to the provider.

Manual adjudication workflows introduce systemic vulnerabilities at every phase. Human review for benefit determinations, with interpretation of complex plan language under time pressure, leads to divergent coverage determinations for substantively similar claims across plans. The cumulative effect of these factors is variability in claims processing times, claims outcomes, and claims processing costs. Automated adjudication systems seek to reduce these problems by providing shared, common spaces for storing adjudication rules and benefit plans and by standardizing the processing logic on all claims, not just a subset.

Stage	Process Activity	Key Data Elements Verified	Common Failure Causes	Workflow Type
Claim Submission	Electronic claim filing via 837P or 837I format	Provider NPI, patient demographics, service codes	Incorrect format, missing fields, invalid codes	Manual/Automated
Eligibility Determination	Verification of active plan enrollment on service date	Enrollment status, provider credentials, covered benefits	Outdated enrollment files, credentialing lapses, demographic mismatches	Batch / Real-Time
Benefit Validation	Authorization of procedure and diagnosis code combinations	Plan document rules, service location, benefit coverage	Unauthorized code pairs, non-covered services, COB conflicts	Rule-Based / Manual
Coordination of Benefits	Determination of payment order across multiple payers	Primary and secondary payer liability, plan provisions	Incorrect payer sequencing, missing COB data	Manual/Automated
Pricing Determination	Calculation of allowed amount using fee schedules	Contracted rates, modifiers, place-of-service adjustments	Outdated fee schedules, missing modifier logic	Automated / Manual
Financial Responsibility Calculation	Distribution of liability across payer, deductible, copay, and coinsurance	Deductible accumulator, out-of-pocket maximum, coinsurance rate	Incorrect accumulator data, coinsurance calculation errors	Automated
Remittance and Payment	Issuance of net payment and remittance advice to provider	Net payment amount, denial codes, adjustment reasons	Payment delays, incorrect denial coding	Automated

Table 1: Claims Adjudication Lifecycle Stages and Associated Process Characteristics [1, 2]

3. Automated Claims Processing Architectures: Rule Engines and Real-Time Validation

Automated claims adjudication is based on an architecture of rule engines defining requirements for payer benefit plans as logical conditions, which are used to evaluate claims against a normalized collection of coverage rules, exclusions, prior authorization requirements, fee schedule rules, and other parameters governing consistent, auditable adjudication decisions made against every claim. Rule-based CDSS applies. Clinical guidelines as condition-action rules reason over incoming clinical data against a knowledge base. CDSS are capable of autonomously rendering recommendations or actions based on such rule processing capabilities. These systems can help improve decision consistency and reduce the variability of outcomes in high-volume processing environments. This directly relates to claims processing, where claims need to be interpreted according to their policy rules with a high degree of consistency [3] .

A major feature of automated adjudication infrastructure is real-time eligibility verification. Payer enrollment systems allow adjudication systems to look up a patient's enrollment status upon submission of the claim. Information may include benefit details such as deductible accrual status, out-of-pocket maximum status, and coordination of benefits information for a patient's plan. For healthcare applications, real-time eligibility allows providers and payers to assess a patient's eligibility, cost-sharing, and network status prior to a service being rendered or a claim being submitted, which greatly reduces the number of eligibility-related claim denials that require manual analysis and resubmission [4].

Pre-edits (automated data validation) can occur upstream of the rule engine. This ensures claims entering the adjudication pipeline contain the required structural, syntactical, and semantic components necessary for claims to process correctly. Front-end edits check claims for the presence of required fields, code pairs, date logic, and the validity of the NPI. Claims that fail any of these front-end edits will be rejected back to the submitting provider before ever entering the adjudication queue, preventing incomplete or malformed claims from tying up adjudication resources. Smart edit libraries overlay clinical logic edits that examine combinations of procedure codes for medical necessity, gender-specific procedures, and age-appropriate services in addition to basic structural validation.

Automated payment calculation modules complete the adjudication cycle more rapidly by implementing pricing logic without human involvement. Fee schedule lookup engines provide the contracted rates for rendering providers that are effective on the date of service requested and calculate the allowed amount for each component with appropriate modifiers, place of service adjustments, and multiple procedure reductions. Together these automated processes compose a claims processing model that handles the majority of claims without human involvement and makes claims decisions on an exception basis, reserving human review only for those claims that require additional information and expertise [3].

Processing Stage	Validation Type	Failure Action	Human Involvement
Structural Pre-Edit	Syntactic	Rejected to provider before queue entry	None
Clinical Logic Pre-Edit	Semantic	Flagged with clinical edit reason	None
Real-Time Eligibility Check	Administrative	Eligibility denial issued to provider	None
Rule Engine Benefit Validation	Policy	Claim denied with denial code	None
Fee Schedule Pricing Lookup	Financial	Exception routed for manual rate review	Exception cases only
Payment Calculation	Financial Responsibility	Exception routed for manual review	Exception cases only
Human Review Queue	Complex Adjudication	Manual decision rendered	Required

Table 2: Rule Engine and Real-Time Validation Stages: Failure Actions and Human Involvement Levels [3, 4]

4. Smart error detection and predictive analytics for claims processing

Predictive analytics applied to the healthcare claims adjudication process is an evolution from rule-based automation. In place of a process that compares claims against business rules and denies them when one or more rules are not met, a predictive approach applies analytics to determine claim error likelihood, predict adverse conditions, and recommend corrective actions in advance of adjudication. Predictive applications for claim submissions, comparing them with historical patterns found within claims processing, having already had the statistical profile of bad claims before the fact. Anomaly detection patterns on transaction streams using clustering, density estimation, and classification-based methods enable the detection of abnormal data points and outliers from established behavioral trends or patterns and can be applied directly in classifying bad or fraudulent claims within mass adjudication pipelines [5].

Beyond retrospective analysis, predictive analytics tools operate in real-time to provide point-of-care insights and alerts and trigger early interventions to reduce risk or reinforce appropriate behavior. For claims management, predictive analytics assesses information upon receipt, including provider billing patterns, plausibility of procedure-diagnosis pairs, authorization status, and patient utilization patterns, to create risk scores that can optimize claims assignment and prioritize processing based on risk. Demonstrated success in other domains for healthcare predictive analytics has included increasing efficiency of operations for when to focus resources and attention on cases most in need of intervention instead of an even allocation of review effort across all submissions [6].

Machine learning models can be applied to claims analytics databases, which are based on historic adjudicated claims. These predictive models are suitable for finding non-linear relationships between the characteristics of a claim and payments made for it. Anomaly detection algorithms applied to claims can ease quality assurance and fraud detection. Statistical anomaly detection identifies claims that deviate from the expected pattern of billing for a given provider, specialty, or service category. Anomalies include potential coding errors, duplicates, and data entry errors. Unsupervised learning, such as clustering algorithms, can be used to detect new fraud patterns that cannot be captured by existing rules of known fraud patterns [5].

Pre-submission scrubbing tools, usually integrated into provider billing systems, use predictive analytics to scan the provider side of the claims workflow in order to identify and correct potential denials before the claim is sent to the payer. Potential reasons for denial may include missing authorization documentation and discrepancies between claim documentation and coding. By identifying these problems before submission, predictive scrubbing can eliminate much of the claim-denial-resubmission cycle that drives up administrative costs and delays payment. The difference in cost becomes much more pronounced in high-volume billing environments, where it costs many times more to resubmit a denied claim than it does to prevent the denial at the point of first submission [6].

Analytics Capability	Method / Technique	Application in Claims	Output
Historical Pattern Matching	Statistical profiling of past adjudicated claims	Identifies submissions matching known error or denial profiles	Risk-flagged claim for review
Anomaly Detection	Clustering, density estimation, classification	Detects outliers in transaction streams deviating from billing norms	Fraudulent or erroneous claim flag
Real-Time Risk Scoring	Predictive model scoring at point of receipt	Assesses billing patterns, procedure-diagnosis plausibility, authorization status	Risk score for claims routing
Machine Learning – Non-linear Modeling	Supervised learning on historical claim datasets	Identifies complex relationships between claim attributes and payment outcomes	Predictive denial probability score
Unsupervised Fraud Detection	Clustering algorithms on unlabeled claims data	Detects novel fraud schemes outside known rule-based patterns	New fraud pattern alert
Pre-Submission Scrubbing	Predictive analytics integrated into provider billing systems	Identifies missing authorization and coding discrepancies before submission	Denial prevention alert to provider

Table 3: Predictive Analytics and Anomaly Detection Techniques: Methods, Claims Applications, and Processing Outputs [5, 6]

5. Interoperability, API integration and performance monitoring frameworks

The performance of automated claims adjudication systems is heavily dependent on these systems being able to communicate with the collection of other systems involved in the healthcare reimbursement process, such as provider electronic health record systems, clearinghouse transaction systems, pharmacy benefit management systems, and enrollment systems, in real time, to obtain the information required to adjudicate a claim correctly. To realize a move towards patient-centered, interoperable health information systems, the infrastructure needs to be in place to enable such data flows to cross both organizational and technological boundaries, and allow for the rich data processing necessary for automated adjudication to take place [7].

Interoperability frameworks have further accelerated the use of APIs across the health system, enabling seamless data exchange and integration between different health information systems. FHIR-based APIs can safely share real-time decisions about prior authorizations from management systems, information about drug coverage from pharmacy benefit managers, and access to provider directories that show the current status of credentialing and network participation. In addition, health information exchange using standards-based data, which allows clinical and administrative data to flow across organizational boundaries via structured interfaces, improves care coordination and administrative efficiencies, providing yet another example of the case for investing in the interoperability of adjudications [8].

Performance monitoring dashboards are an operational control layer of automated adjudication and are used by system administrators and operations staff to monitor the state of the claims processing system. Dashboards are typically visual displays of key performance indicators (KPIs) powered by data aggregated from multiple points in the adjudication pipeline. Common KPIs include number of claims adjudicated per period, rate of auto-adjudications, denial rates, first-pass resolution rates, and latency. This ties to the ability to detect and remediate issues before the point at which the performance of payment cycle time is materially negatively impacted when the performance intelligence is real-time. . A requirement is the health information exchange (HIE) infrastructure that provides for connected data sharing across disparate systems and interoperability architecture that enables adjudication performance management [9].

Automated adjudication systems include audit trail functions, both for business reasons and to meet regulatory requirements. Each adjudication function (rule evaluation, price lookup, coordination of benefits determination, or payment determination) creates a tamper-proof event log. This audit trail supports post-hoc verification, dispute resolution, and regulatory compliance functions. Also, data from the audit trail could be used as input data to train the machine learning algorithms to learn the effectiveness of adjudication rules over time and create a feedback loop whereby operational adjudication data feeds into better automated decisions over time [7].

Connected System	Integration Method	Data Exchanged	Adjudication Function Supported
Provider EHR Platform	FHIR-based API	Clinical and administrative records	Claim validation and medical necessity review
Clearinghouse Transaction System	EDI / API	Claim transaction data, rejection reports	Claim submission and pre-edit processing
Pharmacy Benefit Management System	FHIR-based API	Formulary data, coverage determinations	Drug benefit validation and pricing
Authorization Management System	FHIR-based API	Real-time prior authorization decisions	Benefit validation and authorization compliance
Payer Enrollment Repository	Real-time API query	Enrollment status, plan details, COB data	Eligibility determination
Provider Directory Service	API	Credentialing status, network participation	Provider eligibility and network validation
HIE Infrastructure	Standards-based interface	Clinical and administrative data across organizations	Care coordination and cross-system adjudication support

Table 4: Interoperability and API Integration: Connected Systems, Exchange Methods, and Adjudication Functions Supported [7, 8]

6. Stakeholder concerns and calculated implications of automated adjudication systems

Automated claims adjudication can deliver different but complementary benefits across the ecosystem of healthcare beneficiaries: healthcare providers, patients, employers, and payers. For health care providers, automated claims adjudication impacts the revenue cycle most considerably by reducing days in accounts receivable (DAR) as claims are adjudicated and paid more quickly. Automated adjudication systems can also process clean claims virtually in real time, eliminating multi-day queues in the manual adjudication environment. The application of information technology to the transactional workflow of hospitals and health systems has been shown to improve service quality and reduce administrative cycle time. The compression of cycle time in the automated adjudication environment leads to improved provider cash flow [10].

Denial rate reduction is another advantage of adjudication automation for providers. Automated systems apply the same rules to all claims and are not subject to the human discretion that causes inconsistent coverage determinations. Automated adjudication systems thereby provide a predictable set of denials, which can be better managed through more precise optimization of billing practices. Automating adjudications for patients will make coverage decisions more transparent and timely, alleviating uncertainty about claims in process and providing better estimates of patient responsibility. A great deal of patient-oriented research has focused on the role of coordinated and transparent administrative processes in fostering patient trust and relieving the burden of having to navigate fragmented care and billing systems [11].

Beyond its role in the industry, adjudication automation presents a business case to payers and self-insured employers due to reduced administrative cost per claim, better data for actuarial analysis, and improved fraud detection. The cost of adjudicating a claim on an automated adjudication pathway is substantially lower than the equivalent cost for manual adjudication. With more structured data flowing from automated adjudication systems, there are new opportunities for population health analytics and plan design, enabling data-driven benefits designs that effectively balance cost control with access. Health information exchange (HIE) infrastructure enables improvement in administrative efficiencies and care coordination as a result of implementation using appropriate standards and governance, including operating rules and administrative simplifications. Automated adjudication is a major enabler of sustainable operational performance when cleverly employed [12].

In summary, the early and active contribution of automated adjudication offers an opportunity for healthcare insurance operations to differentiate themselves. The combination of rule-based automation, predictive analytics, real-time interoperability, and performance measurement will improve the accuracy of claims, speed of reimbursement, and satisfaction of all stakeholders across the entire healthcare reimbursement ecosystem [10].

Conclusion

Automated healthcare claims adjudication workflows leverage rule-based processing, real-time eligibility verification, smart error tracking, and interoperability via application programming interfaces (APIs) to convert the labor- and error-intensive claims processing workflow from an exception-ridden, fragmented, and error-prone process to a standardized, high-quality operational process. Predictive analytics, anomaly detection, and other analytics capabilities go beyond just denying claims reactively and focus on proactively correcting and resolving them, as well as finding resolution for previously unresolvable claims before they are submitted. Standards-based interoperability infrastructure positions adjudication systems as part of a data-sharing ecosystem to deliver real-time, accurate decisions for providers, payers, and pharmacy benefit manager (PBM) use cases. Faster provider revenue cycles, improved coverage transparency for patients, and reduced administrative costs for payers suggest that clever adjudication automation delivers more than just process efficiency. This is a critical investment in a reliable, accurate, and sustainable healthcare reimbursement system to provide the confidence, consistency, and financial predictability that all involved in the care continuum need.

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