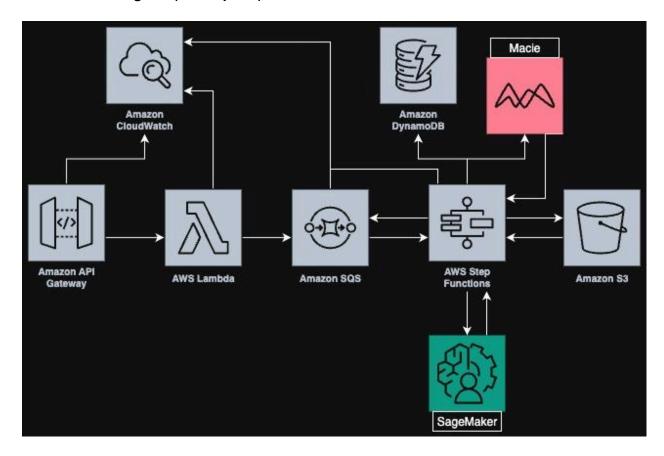
# **System Design: Extraction Service**

## a. High-Level Architecture

The system is a serverless, queue-based pipeline deployed in AWS GovCloud, using managed services and open-source LLMs to ensure compliance and scalability. Key components:

- API Gateway (REST API): Entry point for job submissions (e.g., POST /jobs with config and source path). Authenticated via IAM roles for multi-tenancy.
- **AWS Lambda (Job Manager):** Handles job creation, validation, and queueing. Stores job metadata in DynamoDB.
- Amazon SQS (Job Queue): Decouples job submission from processing. Standard queue for extraction tasks, dead-letter queue for failures.
- **AWS Step Functions:** Orchestrates the workflow (crawl → extract → deliver). Ensures idempotency and state tracking.
- **AWS S3 (Storage):** Stores input docs (S3/SharePoint sync), output JSON, and logs. Encrypted with SSE-KMS.
- Amazon SageMaker (LLM Inference): Hosts open-source LLM (e.g., BERT or LLaMA from Hugging Face) for metadata extraction. Deployed in GovCloud with custom endpoints.
- AWS Macie: Scans for PII before processing; flags sensitive docs for review.
- Amazon CloudWatch & X-Ray: Metrics, logs, and traces for observability. Custom dashboards for job status.
- AWS SDK (SharePoint Connector): Crawls SharePoint via GovCloud-compatible APIs (assumes OAuth or federated access).

# **Architecture Diagram (Conceptual):**



**Rationale:** Serverless minimizes ops overhead and scales automatically. SQS/Step Functions ensure reliable, idempotent processing. SageMaker supports open-source LLMs, avoiding third-party APIs. All services are GovCloud-compliant.

### b. Data Model

### Job Metadata (DynamoDB):

```
json
{
    "job_id": "uuid-1234", // Partition key
    "tenant_id": "agency-xyz", // Sort key for multi-tenancy
    "status": "pending | running | completed | failed",
    "config": {
        "fields": ["author_names", "publish_date", "abstract_summary",
    "code_snippets"],
        "source_type": "s3 | sharepoint",
        "source_path": "s3://bucket/prefix | https://sharepoint-url",
        "output_path": "s3://bucket/output"
},
```

```
"created_at": "2025-09-18T10:02:00Z",
   "updated_at": "2025-09-18T10:03:00Z",
   "doc_count": 1000,
   "failed_docs": ["doc1.pdf", "doc2.pdf"]
}
```

# Output JSON (S3 per doc):

```
json
{
  "doc_id": "doc1.pdf",
  "job_id": "uuid-1234",
  "tenant_id": "agency-xyz",
  "metadata": {
      "author_names": ["John Doe", "Jane Smith"],
      "publish_date": "2023-01-01",
      "abstract_summary": "This paper discusses AI advancements...",
      "code_snippets": ["def example(): ..."],
      "pii_flagged": false
},
      "status": "success | failed",
      "error": "null | parsing_failed"
}
```

**Rationale:** DynamoDB for low-latency job tracking; S3 for scalable, durable output storage. JSON schemas are simple, extensible, and normalized for downstream use.

### c. Sequence Diagrams

#### 1. Create Job $\rightarrow$ Crawl $\rightarrow$ Extract $\rightarrow$ Deliver Results

- 1. User via API Gateway submits POST /jobs with config and source path.
- 2. Lambda job manager validates request, stores job in DynamoDB, pushes to SQS.
- 3. Step Functions picks up a job, triggers Lambda to list docs from S3/SharePoint.
- 4. For each doc:
  - Macie scans for PII; flags if sensitive.
  - Lambda pulls doc from S3/SharePoint.
  - SageMaker extracts metadata based on config.
  - Lambda saves JSON to S3 output path.
- 5. Step Functions updates job status in DynamoDB.
- 6. API Gateway returns job status or notifies user via SNS email.

# 2. Retries, Backoff, Idempotency

- 1. If extraction fails, SQS retries 3x with exponential backoff of 1s, 2s, 4s.
- 2. Step Functions ensure idempotency via job id/doc id checks.
- 3. Failed docs move to dead-letter queue; logged in CloudWatch.
- 4. Lambda retry handler analyzes failures, updates DynamoDB, alerts if threshold exceeded.

### d. Scaling Strategy

- 100k Docs: Process in batches. SQS handles backpressure by buffering tasks.
- **Concurrency:** Lambda auto-scales up to GovCloud limits, ~1000 concurrent. SageMaker endpoints scale via instance count.
- **Backpressure:** SQS visibility timeout prevents overload; Step Functions throttles if SageMaker hits rate limits.
- Rate Limits: API Gateway throttles at 100 reg/s; configurable per tenant.
- Cost Caps: Monitor via CloudWatch Budgets; cap SageMaker instances at 10 to stay under \$1k/100k docs.
- Optimization: Pre-warm SageMaker endpoints; use spot instances for non-critical jobs.

#### e. Observability & Ops Runbook

### Observability:

- Metrics via CloudWatch: Processed docs, succeeded/failed, p50/p95 latency.
- **Logs:** CloudWatch Logs for Lambda, SageMaker, Step Functions. Structured JSON with job id/doc id.
- **Traces:** X-Ray for end-to-end request tracking.
- Dashboards: CloudWatch Dashboard showing throughput, error rate, cost, and job status.

### Runbook:

- **SLOs:** 99% API uptime, <1% failed extractions, <24h for 100k docs.
- Alerts: CloudWatch Alarms for >1% error rate, >\$500 spend, or >5min p95 latency.
- Actions:
  - High error rate → Check DLQ, scale SageMaker instances.
  - Cost overrun → Reduce batch size, use spot instances.
  - Job stuck → Restart Step Functions state machine; inspect logs.