AWS Introduction

Most Important AWS Topics/Services for Data Science

Here's a curated list of **essential AWS services** specifically for Data Science, with beginner-friendly explanations:

AWS Service	Why It Matters (Simple Explanation)
S3 (Simple Storage Service)	Like a big online USB drive. You use it to store datasets, CSVs, models, logs, etc.
EC2 (Elastic Compute Cloud)	Like renting a remote computer in the cloud where you can run Jupyter, Python code, train models, etc.
SageMaker	AWS's main tool for Data Science. Lets you build, train, deploy ML models without worrying about hardware or backend setup.
IAM (Identity & Access Management)	Like a lock & key system. Controls who can access your AWS stuff (important for security).
Lambda	Run Python code automatically when something happens (like when new data is uploaded). No server needed.
CloudWatch	Like a security camera + diary. It watches your AWS resources, tracks usage, errors, etc.
RDS (Relational Database Service)	Managed SQL databases in the cloud. Think MySQL/PostgreSQL without needing to install anything.
Glue	Helps move and transform data between systems. Useful for cleaning and preparing large datasets.
Athena	Query large CSV/Parquet/JSON files stored in S3 using SQL. No need to move data anywhere.
Step Functions	Like a flowchart engine. Helps automate workflows like "first clean data \rightarrow then train model \rightarrow then email results".

AWS Service	Why It Matters (Simple Explanation)
CloudFormation	Like a recipe. You write code to describe your AWS setup, and AWS builds it for you automatically.
ECR + ECS (Optional later)	For containerized apps. Helpful if you use Docker for model deployment. We'll revisit this later.

 Amazon S3 – Your cloud "hard drive" for storing datasets, models, and results.

Why? All data science starts with data, and S3 is where you'll store it.

2. **AWS Lambda** – Run code without managing servers (e.g., preprocess data automatically).

Why? Automate small tasks like cleaning data or triggering ML pipelines.

 Amazon SageMaker – Fully managed service to build, train, and deploy ML models.

Why? The heart of AWS ML—no need to set up servers or frameworks manually.

4. **AWS Glue** – Prepares and transforms data for analysis (ETL: Extract, Transform, Load).

Why? Real-world data is messy; Glue cleans it for you.

- 5. **Amazon EC2** Virtual computers in the cloud (for heavy computations). *Why?* When you need more power than your laptop for training big models.
- 6. **AWS IAM** Controls who can access your AWS resources (security first!). *Why?* Avoid accidental data leaks or unauthorized costs.
- 7. **Amazon Athena** Query data in S3 using SQL (no database setup needed). *Why?* Quickly analyze large datasets without moving them.
- 8. **Amazon QuickSight** Create dashboards and visualize your data. *Why?* Share insights with non-technical teams.
- AWS Step Functions Coordinate multiple AWS services into workflows.
 Why? Chain data preprocessing → training → deployment automatically.

10. Bedrock (for GenAl) – Access to foundation models like Claude, Llama 2.
Why? Add generative Al (text, images) to your apps without training from scratch.

Simple Questions

Which topic excites you the most? Or ask:

- "How do I store a CSV file in AWS?" → S3
- "How do I train a model without installing anything?" → SageMaker
- "How do I automate a data pipeline?" → Lambda + Glue

Common Bill Shock Scenarios (And Fixes)

Scenario	Why It Happens	How to Prevent
EC2 left running 24/7	Forgot to turn off a test server	Use AWS Instance Scheduler or set auto-termination timers
S3 bucket with public access	Accidentally exposed files	Enable S3 Block Public Access
Lambda functions spamming	Infinite loop in code	Set Concurrency Limits

Key Tips for Beginners

- Always enable MFA (Multi-Factor Auth): Prevents hackers from mining Bitcoin on your account.
- Delete unused resources: Check AWS Cost Explorer monthly for "ghost" services.
- 3. Use the Free Tier wisely: Stick to t2.micro EC2 instances and small S3 buckets.
- 4. Mock bills with the Pricing Calculator: AWS Pricing Calculator.

Reality Check

- AWS won't charge you without warning (you'll get alerts first).
- Small experiments (e.g., running a server for 1 hour) cost pennies.
- Major leaks usually happen due to misconfigured auto-scaling or public data transfers.

What Is the AWS Free Tier?

AWS offers a 12-month Free Tier for most services.

Examples:

- **S3**: 5 GB of storage free
- **EC2**: 750 hours/month of micro instance (t2.micro or t3.micro)
- SageMaker: 250 hours of ml.t2.medium
- Bedrock: limited free LLM tokens for some models



But: If you go beyond the free limits (like 6 GB S3 instead of 5 GB), then you're charged for the **extra 1 GB only.**

Pro Tips for New Users

Mistake	Prevention
Leaving EC2 running 24/7	Stop or terminate it after use
Choosing large instance types	Stick to t2.micro, t3.micro for Free Tier
Not deleting SageMaker notebooks	Stop the notebook to avoid surprise charges
Using services not in Free Tier	Double-check AWS Free Tier website

What is an EC2 Instance?

Imagine an EC2 instance as a "virtual computer" that you rent from AWS.

- It's like a remote laptop with CPU, RAM, storage, and OS (Linux/Windows).
- You control it entirely: install software, run code, or host websites.
- You pay only for the time it's running (per second/hour).
- You can log in, run code, train ML models, host apps, etc.

Real-Life Example:

Case: You want to train a model on 1 million rows of data but your laptop is too slow.

- You create an EC2 instance with more memory and CPU (or GPU).
- Upload your data and training script.
- Run the training on EC2.
- · Download the model weights.
- Stop the instance.

Cost? Just the hours you used it — could be ₹10 or ₹100 depending on size/time.

O Common Beginner Mistakes

Mistake	Fix
Leaving EC2 running	Stop it after use from AWS console
Picking big instance types	Start with Free Tier (t2.micro)
Not setting SSH key	You won't be able to login — always download the .pem key safely
Using EC2 to store data permanently	EC2 storage gets erased on termination — store important data in S3

Smart Strategy for Beginners

Use Case	What to Do
Learning EC2, Jupyter, Python setup	✓ Use t2.micro or t3.micro , stop it when done

Use Case	What to Do
Hosting website/chatbot (low traffic)	Use t3.micro, set a ₹500 budget alert, monitor usage
Hosting only during the day	✓ Use Lambda or Fargate (event-based, cheaper)
Need GPU for model training	✓ Use EC2 just for the few hours you train, then stop

S Cost Insight

Setup	Monthly Cost (est.)
Single t3.medium instance	₹1,300-₹1,600
Two t3.micro instances	₹1,200-₹1,500 combined
One g4dn.xlarge GPU instance (training only)	₹12,000-₹20,000+ 💸

Types of Cloud Infrastructure Services

- Infrastructure as a Service (laaS)
- Platform as a Service (PaaS)
- Software as a Service (SaaS)

1. laaS (Infrastructure as a Service)



Analogy: You get the land and foundation.

Your responsibility:

- Operating system (Linux, Windows)
- All your software (Python, libraries, specific versions)
- Applications (Jupyter notebooks, custom data pipelines)
- Network configuration within your virtual network
- Security patching of your OS and software

AWS's responsibility:

- Physical servers, networking hardware
- Virtualization layer (making one big server act like many small ones)
- Power, cooling, physical security of data centers

Key Characteristics:

- Maximum control: You decide everything.
- Flexibility: Tailor environments exactly to your needs.
- Cost-effective for variable loads: Pay only for what you use.
- Requires expertise: You need to know how to manage servers.

AWS Examples for Data Science:

- Amazon EC2 (Elastic Compute Cloud): Your virtual machine. Spin up a
 powerful EC2 instance with many CPUs or GPUs for training complex ML
 models. Install specific deep learning frameworks (TensorFlow, PyTorch)
 directly onto it.
- Amazon S3 (Simple Storage Service): Object storage for massive datasets. Store raw data lakes, processed data, model artifacts, and project files. Highly durable and scalable, no server to manage for S3 itself.

- Amazon EBS (Elastic Block Store): Persistent storage volumes attached to EC2 instances. Like a hard drive for your virtual server, good for databases or when your EC2 needs fast, dedicated storage.
- Amazon VPC (Virtual Private Cloud): Your isolated network in the cloud.
 Define your own IP addresses, subnets, and security rules. Essential for securing your data science environments.
- Why: Max control, highly flexible.

2. PaaS (Platform as a Service)

Analogy: You get a partially built house.



Your responsibility:

- Your application code (e.g., Python script for a model API)
- Your data
- Configuration specific to your application

AWS's responsibility:

- All laaS components (servers, network, OS)
- Runtime environment (e.g., Python interpreter, Node.js)

- Middleware (web servers like Apache/Nginx, database software)
- Scalability and load balancing for the platform
- Security patching for the platform itself

Key Characteristics:

- Faster development/deployment: Focus on coding, not infrastructure.
- Managed environment: AWS handles most operational tasks.
- Built-in scalability: Often scales automatically or with minimal config.
- Less control/flexibility: Limited access to the underlying infrastructure.

AWS Examples for Data Science:

- Amazon RDS (Relational Database Service): Managed relational databases (PostgreSQL, MySQL, SQL Server, etc.). You don't install or patch the database server; AWS handles it. You just connect and query. Great for structured data.
- AWS Lambda (Serverless Compute): Run your code without managing servers. Upload a Python function to process data, trigger it when a new file lands in S3 or on a schedule. Pay only when your code runs. Perfect for eventdriven tasks.
- Amazon SageMaker: A comprehensive ML platform. Provides managed environments for:
 - Jupyter Notebooks (SageMaker Studio): Pre-configured, powerful notebooks for exploration and development.
 - Training Jobs: Run large-scale model training without setting up servers.
 - Model Endpoints: Deploy models as APIs for real-time predictions;
 SageMaker manages the hosting.
- AWS Elastic Beanstalk: Deploy web applications (like a Flask app serving an ML model) easily. AWS handles the underlying EC2 instances, load balancing, and auto-scaling.

• Why: Focus on coding, faster deployment.

3. SaaS (Software as a Service)

• **Analogy:** You rent a furnished apartment.



Your responsibility:

- Using the application.
- Your data within the application.

AWS's responsibility:

- Everything: the application itself, its data, runtime, middleware, OS, servers, network, storage.
- All maintenance, updates, security.

Key Characteristics:

- Easiest to use: Just log in and go.
- No maintenance burden: Provider handles everything.
- Subscription-based: Usually paid per user or per usage.
- Least control/customization: You use it as it's provided.

AWS Examples (less direct "infrastructure" but good to know):

- Amazon QuickSight: Business intelligence service. Connects to your data sources (like S3, RDS) and lets you build interactive dashboards and reports.
 You don't manage any servers for QuickSight; you just use the web interface.
- AWS Management Console: The web interface you use to manage all your AWS services. It's a SaaS application itself!
- Why: Easiest to use, no maintenance.

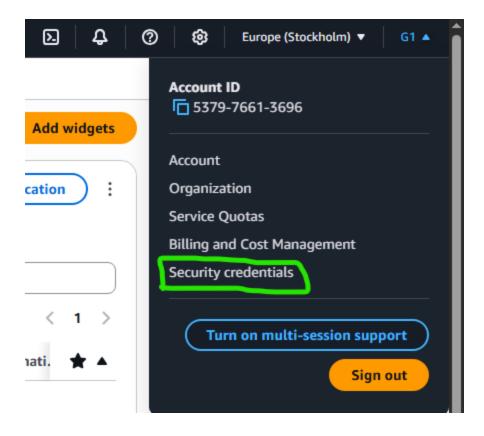
Bonus: Limit Costs Programmatically

You can even set up a Lambda function + CloudWatch Alarm to:

- Stop EC2 or SageMaker automatically when you hit budget limits
- Useful if you want to enforce hard spending stops

MFA (Multi-Factor Authentication) on your AWS Root Account

- 1. Login console as root user
- 2. Click your name in upper right hand corner → Security Credentials



3. Assign MFA



4. Choose Authenticator App

MFA device

Device options

In addition to username and password, you will use this device to authenticate into your account.



Passkey or security key

Authenticate using your fingerprint, face, or screen lock. Create a passkey on this device or use another device, like a FIDO2 security key.



Authenticator app

Authenticate using a code generated by an app installed on your mobile device or computer.

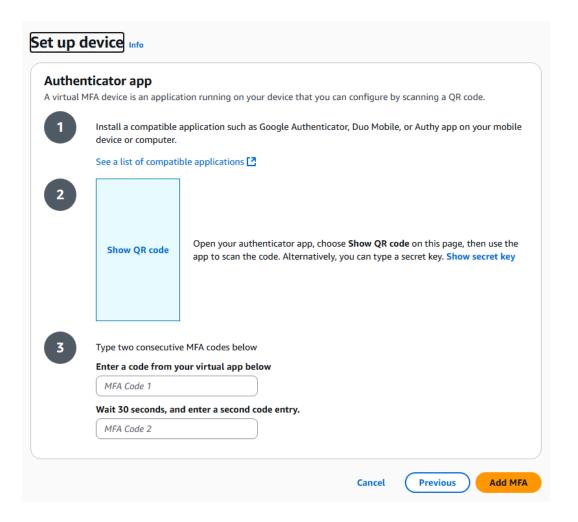


Hardware TOTP token

Authenticate using a code generated by Hardware TOTP token or other hardware devices.

5. Set up device

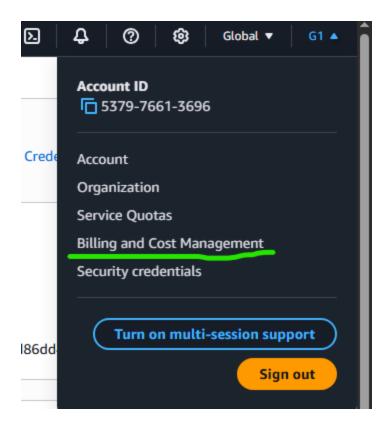
- Show QR code
- Scan on Google Auth App

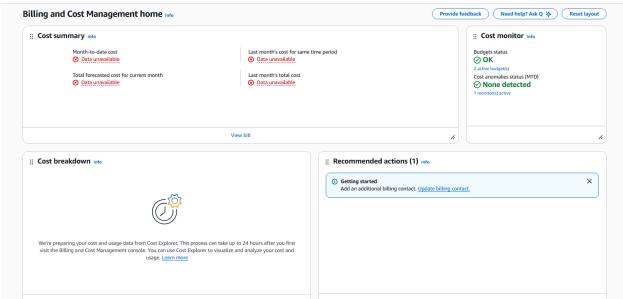


- Enter 2 codes
- Click Add MFA

Check the Bill

• Login to console (Root)

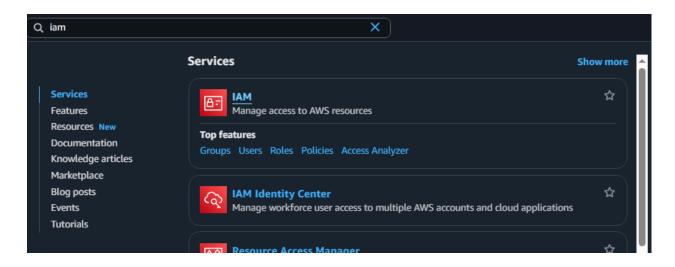




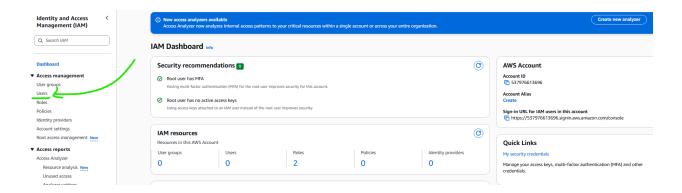
► Create an IAM User (!!Extremely Fookin' IMP)

It's Global service

1. Go to IAM



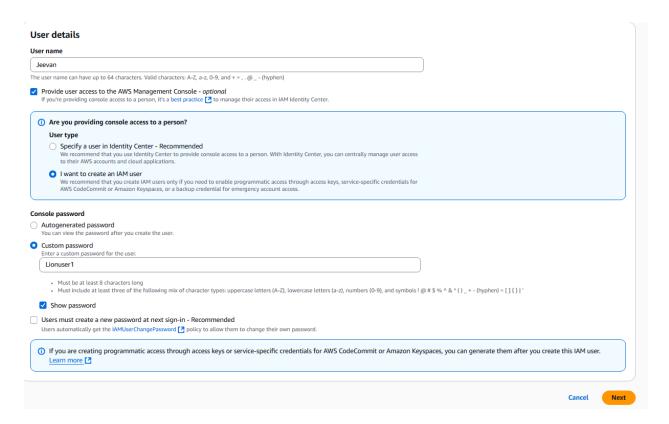
2. Click Users



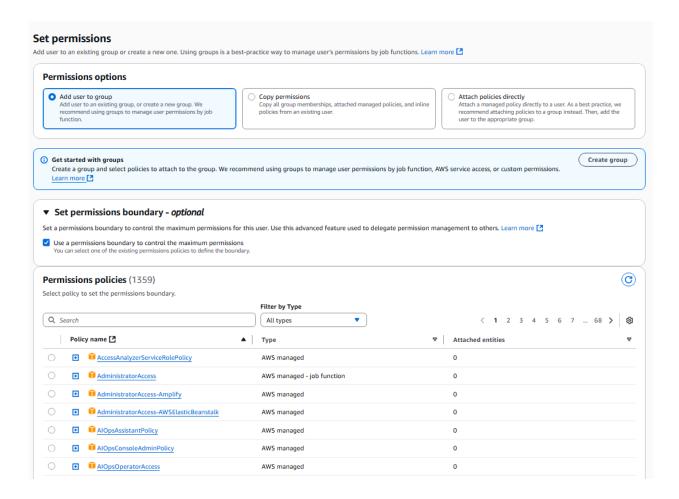
3. Create User



- 1. Tick: Provide user access to the AWS Management Console
- 2. I want to create an IAM user
- 3. Custom password
 - Enter Password
- 4. Untick: Users must create a new password at next sign-in Recommended

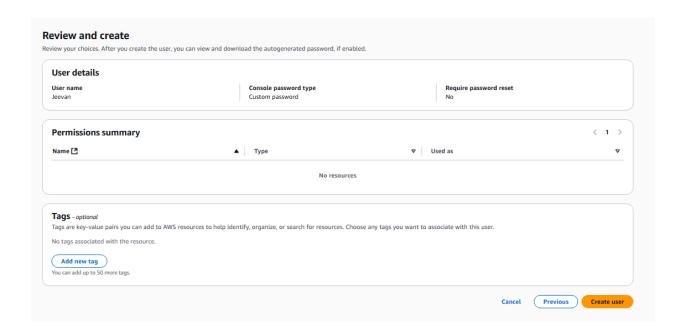


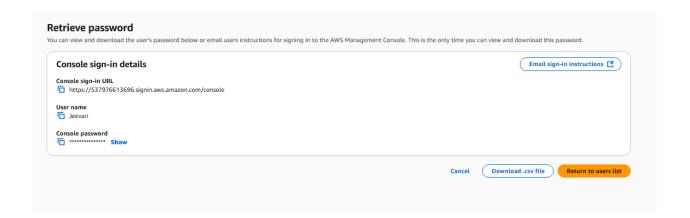
4. Set Permissions



For now, we'll proceed without any permissions.

5. Create User





Console sign-in URL

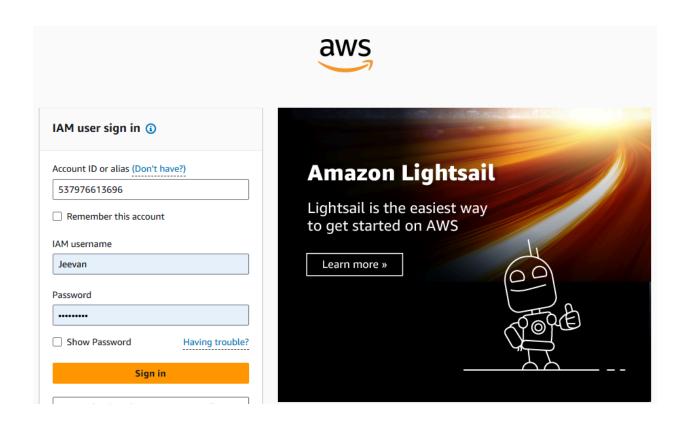
https://537976613696.signin.aws.amazon.com/console

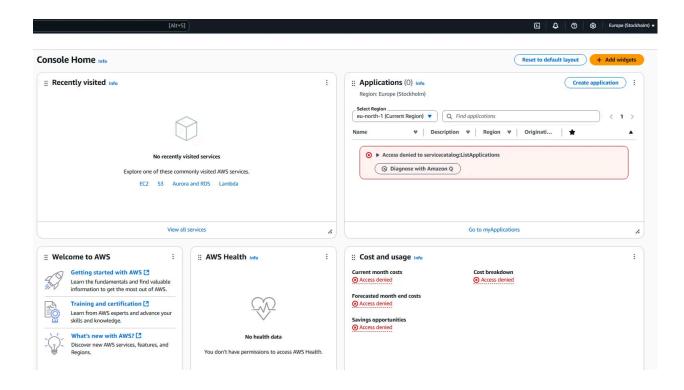
User name

Jeevan

6. Use the link & Sign-in

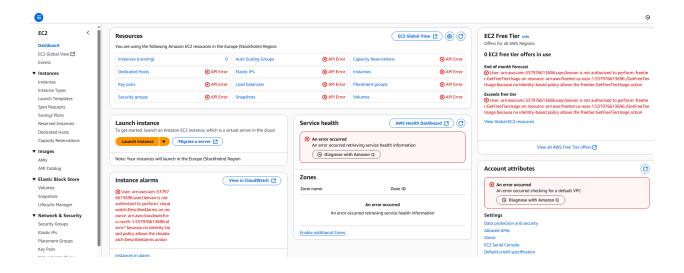
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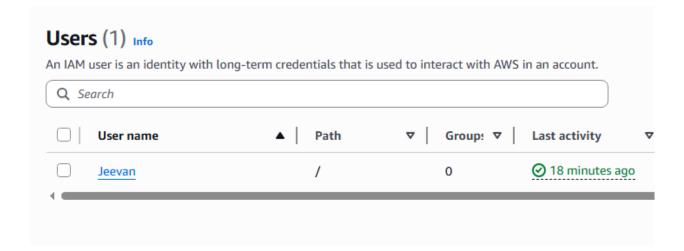


You do not have access to anything.

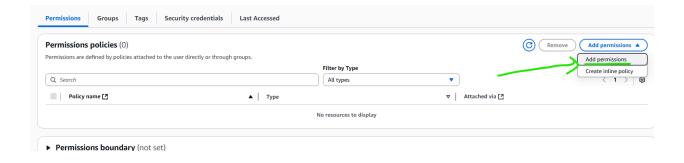


7. Give permissions

- Login your root account
- IAM → Select user



· Add permissions



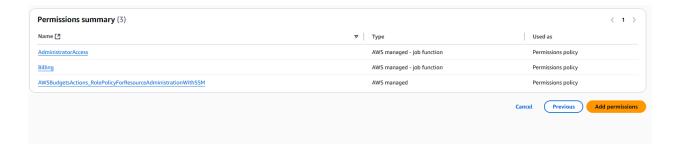
- · Attach policies directly
- Select the permissions:

AdministratorAccess,

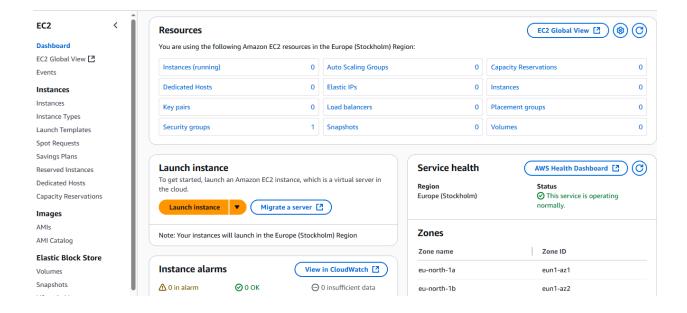
Billing,

 $AWS Budgets Actions_Role Policy For Resource Administration With SSM$

• Add permissions



8. Now login to your IAM User





You'll see the EC2 instance



Even with AdministratorAccess, an IAM user does NOT have access to the aws-portal (billing) or other root-specific actions by default.