

# LearnSphere

Generative AI-Powered Machine Learning System

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**Team: MetaMinds**

Ashwitha. Y

Shreshta. J

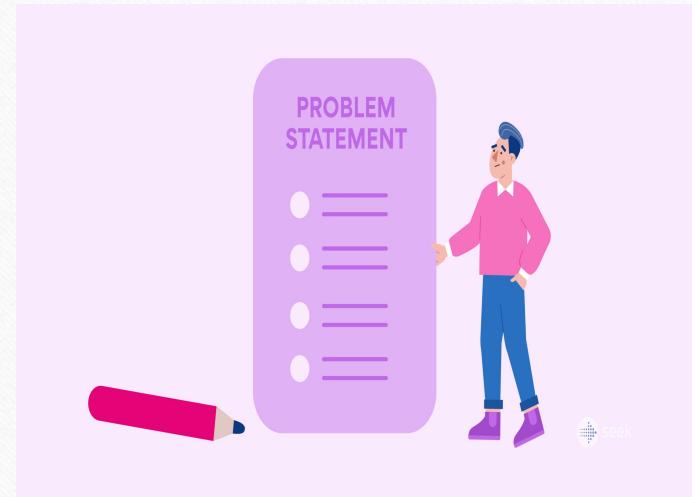
Sonu. V

Srujana. K

# Problem Statement

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- AI-driven learning platform delivers personalized ML education through text, code, audio, and visual explanations. It adapts content to user learning styles, making complex ML concepts easier to understand.



# Proposed System

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- Uses Generative AI for dynamic concept explanations
- Generates real-time code examples for better understanding
- Adapts content difficulty based on user performance
- Provides multi-modal learning (text, code, visual, audio explanations)
- Bridges the gap between theory and practical implementation

# Technologies and Tools

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## Backend:

- Flask
- Python
- Gemini, AI APIs

## Frontend:

- HTML, CSS, JavaScript

# Implementation



# Results

The screenshot shows the LearnSphere interface. On the left, a dark sidebar displays three steps: 1. Select Learning Level (set to Beginner), 2. Choose Topic (set to "What is Machine Learning?"), and 3. Select Mode (listing four options: Concept Explanation, Code Generation, Visual Learning Aids, and Audio Lessons). On the right, a light-colored main area is titled "Welcome to LearnSphere" and contains the text "Pick a level and topic to generate AI-powered ML content."

**LearnSphere**

1. Select Learning Level:

Beginner

2. Choose Topic:

What is Machine Learning?

3. Select Mode:

Concept Explanation

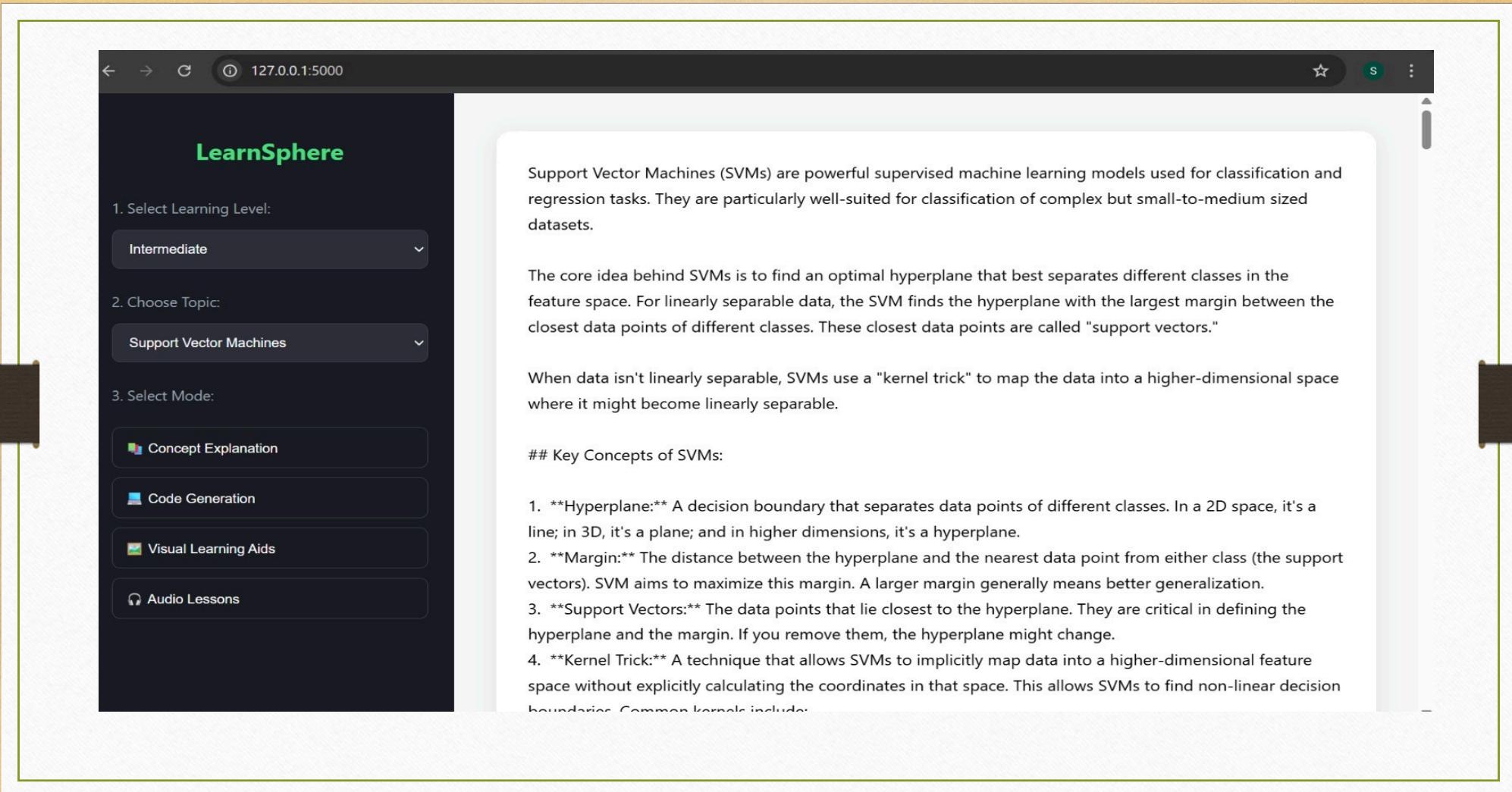
Code Generation

Visual Learning Aids

Audio Lessons

Welcome to LearnSphere

Pick a level and topic to generate AI-powered ML content.



**LearnSphere**

1. Select Learning Level:

Intermediate

2. Choose Topic:

Support Vector Machines

3. Select Mode:

Concept Explanation

Code Generation

Visual Learning Aids

Audio Lessons

Support Vector Machines (SVMs) are powerful supervised machine learning models used for classification and regression tasks. They are particularly well-suited for classification of complex but small-to-medium sized datasets.

The core idea behind SVMs is to find an optimal hyperplane that best separates different classes in the feature space. For linearly separable data, the SVM finds the hyperplane with the largest margin between the closest data points of different classes. These closest data points are called "support vectors."

When data isn't linearly separable, SVMs use a "kernel trick" to map the data into a higher-dimensional space where it might become linearly separable.

## Key Concepts of SVMs:

1. **Hyperplane:** A decision boundary that separates data points of different classes. In a 2D space, it's a line; in 3D, it's a plane; and in higher dimensions, it's a hyperplane.
2. **Margin:** The distance between the hyperplane and the nearest data point from either class (the support vectors). SVM aims to maximize this margin. A larger margin generally means better generalization.
3. **Support Vectors:** The data points that lie closest to the hyperplane. They are critical in defining the hyperplane and the margin. If you remove them, the hyperplane might change.
4. **Kernel Trick:** A technique that allows SVMs to implicitly map data into a higher-dimensional feature space without explicitly calculating the coordinates in that space. This allows SVMs to find non-linear decision boundaries. Common kernels include:

The screenshot shows the LearnSphere interface. It has a dark header with the title 'LearnSphere'. Below it, there are three steps for selecting content:

1. Select Learning Level: A dropdown menu set to 'Intermediate'.
2. Choose Topic: A dropdown menu set to 'Support Vector Machines'.
3. Select Mode: A list of four options: 'Concept Explanation', 'Code Generation', 'Visual Learning Aids', and 'Audio Lessons'. 'Concept Explanation' is highlighted with a blue border.

This example will cover:

1. \*\*Synthetic Data Generation:\*\* To easily visualize the decision boundary.
2. \*\*Data Preprocessing:\*\* Scaling features, which is crucial for SVMs.
3. \*\*Model Training:\*\* Using 'SVC' (Support Vector Classifier).
4. \*\*Model Evaluation:\*\* Metrics like accuracy, confusion matrix, and classification report.
5. \*\*Hyperparameter Tuning:\*\* Using 'GridSearchCV' to find optimal 'C' and 'gamma' values.
6. \*\*Visualization:\*\* Plotting the decision boundary.

```
'''python
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.svm import SVC
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import classification_report, accuracy_score, confusion_matrix
from sklearn.datasets import make_classification, load_iris

# Set a random seed for reproducibility
np.random.seed(42)

# --- 1. Synthetic Data Generation for Visualization ---
```

# LearnSphere

1. Select Learning Level:

Beginner

2. Choose Topic:

Types of Machine Learning

3. Select Mode:

Concept Explanation

Code Generation

Visual Learning Aids

Audio Lessons

## AI Voice Lesson Ready

II 0:08 / 4:54

Download MP3

## Audio Lesson Ready

Playing lesson for **Types of Machine Learning...**