Generative Al

What is Generative Al?

- **Definition**: Generative Al refers to machine learning models that create new content (text, images, videos, code) based on data they've seen.
- How It Works: Models take a prompt (input) and generate an output similar to learned data.
- **Types of Inputs/Outputs**: Prompts can be text, images, or other data types, with outputs like text, images, or modified content (e.g., removing a person from an image).
- Applications: Used in sales, marketing, finance, healthcare, and more for tasks like drafting emails, analyzing data, and automating processes.
- **Impact on Jobs**: Al won't replace jobs entirely, but it will change workflows. Understanding these tools will enhance their use in daily life.
- Course Goals: Learn how generative AI is trained, its legal/ethical implications, and its societal impact.

Generative AI in the Machine Learning Landscape

- **Discriminative Models**: Focus on classifying data (e.g., identifying if an image is a puppy or a bagel). They answer closed-ended questions by learning from labeled data.
- **Generative Models**: Unlike discriminative models, they can generate new content based on training data (e.g., creating a new puppy image). They "imagine" data, not just classify it.
- **Integrating Models**: Combining discriminative, generative models, and other techniques creates more effective AI systems that can produce high-quality, creative outputs.
- Generative Adversarial Networks (GANs): A special type of generative AI where two models (generator and discriminator) compete and learn from each other to improve content generation over time (e.g., generating realistic images).
- Artificial General Intelligence (AGI): The long-term goal is to create an AI that possesses human-like
 intelligence, including reasoning, creativity, social skills, and cognitive abilities like sight and language.
- Choosing the Right Tool: Discriminative models are useful for prediction and classification tasks, while generative Al is great for content creation (e.g., text, images). AGI, in the future, could perform complex human tasks autonomously.

The Evolution of Generative Al

- **Generative Al's Rise in 2023**: Generative Al became mainstream in 2023, with products like ChatGPT achieving 100 million users within two months, a milestone faster than social media platforms.
- Key Drivers of Development:
 - **Computing Power**: Models required 100 million times more computational power than a decade ago, enabled by GPUs, TPUs, and cloud computing. Parallelization allowed larger, more complex models.
 - Data Availability: The explosion of data and synthetic data creation techniques enhanced training capabilities for generative models.
 - Competition: Commercial and political competition among Big Tech and governments pushed faster development.

Model Innovations:

- Generative Adversarial Networks (GANs): Introduced in 2014, GANs significantly improved the quality of generated content by having generator and discriminator models compete and learn from each other.
- Transformers: Transformers revolutionized text processing by analyzing relationships between words, improving contextual understanding, and generating coherent responses (e.g., educational chatbots).
- Reinforcement Learning with Human Feedback (RLHF): RLHF incorporated user feedback to refine models through trial and error, improving content based on ratings from real users. Companies like Midjourney use this feedback loop to enhance image generation.

Model Design and Data Collection for Generative Al

Development Process Overview

Developing a generative AI model involves four key steps:

- **Research & Design**: Decide on the model architecture.
- **Data Collection & Preparation**: Gather and preprocess the training data.
- **Model Training**: Train the model on the prepared data.
- **Model Evaluation**: Assess the model's performance and fine-tune as necessary.

Stable Diffusion Case Study

Stable Diffusion, a 2022 image generation model, provides an example of the development process:

- **Purpose**: Image generation from text prompts.
- **Architecture**: Diffusion model, which generates images by transforming noise.
- **Resources**: Required hundreds of GPUs running for 150,000 hours, costing around \$600,000.

Model Design and Data Collection for Generative Al

Data Collection for Generative Models

- Generative AI models need massive datasets (e.g., Stable Diffusion required 2 billion images, about 100,000 GB of data).
- The data must be **diverse** to ensure it covers a wide range of scenarios and outputs.
- Preprocessing is necessary to adjust the data to the correct format and improve quality (e.g., resizing images).

Privacy & Security Considerations

- Large datasets often contain user-generated content that may have personally identifiable information (PII).
- Developers must anonymize or aggregate data and implement security measures to prevent unauthorized access or misuse.
- Failure to address these issues could lead to copyright and ownership concerns.

Model Training for Generative Al

Training Components:

- Hardware: Determines the speed of training (e.g., personal laptop, local GPUs, or server farms).
- **Time**: Affected by dataset size, model complexity, and training iterations.
- **Cost**: The expense associated with hardware and training time.

Advanced Training Techniques:

- Transfer Learning and Fine-tuning: Pre-trained models are adapted for new tasks, saving time and resources.
- Reinforcement Learning with Human Feedback (RLHF): User feedback is used to refine model responses and improve alignment with user preferences.
- Embeddings: Data entities are transformed into compact representations, improving model understanding and efficiency.

Model Evaluation for Generative Al

Purpose of Evaluation:

- Measures progress during training.
- Enables comparison of models to identify the best fit for specific tasks.
- Benchmarks Al performance against human capabilities to understand strengths and application areas.

Evaluation Methods:

- Quantitative Metrics: Includes discriminative model evaluation metrics (e.g., accuracy) and generative model-specific metrics (e.g., realism, diversity).
- Human-Centered Metrics: Involves comparisons with human performance and subjective evaluation to assess nuanced aspects of content quality.

Model Evaluation for Generative Al

Types of Evaluation:

- **Discriminative Model Evaluation**: Measures performance on well-defined tasks, but may not be suitable for subjective or creative outputs (e.g., beauty in art).
- **Generative Model-Specific Metrics**: Tailored for generative AI tasks, focusing on criteria like novelty and diversity, but they have limitations in capturing all aspects of generated content.
- Human Performance Comparison: Al performance is compared to human performance on standardized tests, providing insight into practical applications but may not be fully accurate due to differing strengths.
- Human Evaluation: The gold standard for evaluating Al-generated content, but it is slow, costly, and introduces human bias.
- Turing Test: A classic human evaluation method where AI passes if a human evaluator cannot distinguish its responses from those of a human. However, the test has been criticized for its potential to confuse unintelligent human behavior with intelligent AI outputs.

Evaluating and Mitigating Social Bias in Al Models

Understanding Social Bias:

 Social bias refers to systematic unfairness or partiality in AI models, often impacting certain groups disproportionately. It arises from various sources, including training data and model assumptions, and can result in harmful societal consequences if not addressed.

Sources of Bias:

- Bias in Data: Imbalanced or unrepresentative training data can lead to skewed Al outputs, perpetuating underrepresentation or misrepresentation of certain groups.
- Bias in Models: Assumptions or optimization choices during model design can result in narrow objectives that produce biased outcomes, even with unbiased data.
- Bias in Use: Users applying generative Al in harmful or malicious ways can lead to biased or unfair results, even when the model itself is designed to be neutral.

Evaluating and Mitigating Social Bias in Al Models

Identifying Bias:

- Representation Analysis: Examines whether the model uses different language or behaves differently when referring to distinct groups, indicating potential bias.
- **Fairness Metrics**: Algorithms that assess equal treatment, opportunity, and accuracy across different groups can help detect subtle biases.
- Human Audits: Involves manual review of model outputs to identify and evaluate bias, providing insights that may be overlooked by automated methods.

• Mitigating Bias:

- Diversifying Training Data: Ensure a broad and balanced representation of groups to counter underrepresentation.
- Model Adjustments: Prioritize diverse data during training to improve representation of marginalized or underrepresented groups.
- Adversarial Training: Use separate models to detect bias during training and adjust the generative Al
 accordingly.
- **Continuous Evaluation**: Regularly assess models for biases and make updates based on emerging anti-bias techniques. Engaging diverse stakeholders throughout development helps identify and address bias early on.

Copyright, Ownership, and Privacy in Generative Al

- Ownership of Al-Generated Content: Determining ownership is complex—whether it's the prompt creator, Al
 developer, or original artists. Al's growing independence challenges traditional IP laws.
- **IP Law and AI**: Existing copyright laws are designed for human creators, but AI's role in content creation forces adaptations in the legal system.
- Best Practices:
 - Check copyright status of Al training data.
 - Ensure AI doesn't use copyrighted content without rights.
 - Consult legal experts and stay updated on IP laws.
- Privacy:
 - Review terms of use and data handling practices.
 - Data shared may be used for model training.
 - Consider using local servers for privacy.
- **Industry Norms**: Different industries respond differently to Al. Creative sectors are cautious, while medical research embraces Al's potential.
- **Evolving Regulations**: Al laws vary globally, with privacy regulations like the EU's requiring compliance, regardless of developer location. Stay informed about legal changes.

Responsible Generative Al Applications

- **Ethical Considerations**: Beyond bias and ownership, it's crucial to address ethical concerns regarding Al usage.
- Malicious Use: Al can be exploited to manipulate society:
 - **Deepfakes**: Synthetic media that misrepresents reality, often for defamation or manipulation.
 - **Misinformation Campaigns**: Al can generate misleading content to sway opinions.
 - **Enhanced Hacking**: All can be used to access critical infrastructure maliciously.
- Detection and Prevention:
 - Human-in-the-loop: Ensuring human review of Al-generated content.
 - Harm Prevention: Blocking harmful content, such as violence or hate speech.
 - **Regular Updates**: Continuously review and update Al models to prevent misuse.
- Access Control: Implementing measures like Know Your Customer (KYC) to verify user identity and prevent illicit
 activities.
- Prompt and Response Monitoring: Developers should monitor for harmful prompts and ensure responses adhere to ethical guidelines.
- **Applications**: Watermarks on Al-generated content can identify its source, but malicious actors may attempt to remove them, necessitating law enforcement intervention.
- **Communication and Feedback**: Developers should engage stakeholders, provide clear usage guidelines, and gather regular feedback to improve AI systems responsibly.

Artificial General Intelligence (AGI)

What is AGI?

AGI is a form of generative AI with human-like intelligence, capable of reasoning across domains, having social skills, thinking creatively, and possessing various forms of perception. It can complete human tasks and exceed human abilities in many areas, presenting both significant benefits and risks.

Pros of AGI:

- **Productivity Boom**: Automation of intellectual work and enhanced human capabilities could drive economic growth and free up leisure time.
- **Research Advancements**: AGI could revolutionize fields like medicine, education, and scientific research.
- Global Problem Solving: It could tackle complex issues such as climate change, energy sources, and supply chain management.
- **Human Enrichment**: Personalized AGI assistants could improve lives through wisdom, care, and efficiency.

Cons of AGI:

- **Economic Disruption**: Automation may lead to job loss or significant changes in existing jobs.
- **Malicious Use**: AGI could be misused if it doesn't align with human values, posing risks to humanity.
- **Existential Threat**: The most extreme risk is AGI deciding to harm or subjugate human society.

Artificial General Intelligence (AGI)

Safety Debate:

 Due to the risks, some propose halting Al development. However, like nuclear physics, Al development is likely to continue with increased regulation.

• Controlling AGI Outcomes:

- **Hard Constraints**: Implement physical and digital limitations, such as isolation or an off switch.
- Alignment Strategies:
 - **Iterative Development**: Start with limited AI releases to identify and address issues.
 - Constitutional AI: Codify human values and ethics, using feedback to improve alignment over time.
 - **Multi-Stakeholder Engagement**: Include diverse perspectives in the development process to ensure balanced ethical considerations.

Government Intervention:

- Regulation: Governments must ensure safety through Al safety regulations, oversight, and transparency.
- International Collaboration: Global cooperation is necessary to ensure that AGI development benefits humanity and adheres to shared guidelines.

Bringing New Al into Old Workflows

Al vs. Human Jobs

Al can assist but not fully replace humans. While Al offers speed and knowledge, it lacks common sense and real-world experience.

• Al's Advantages and Limitations

- Advantages: Fast, deep knowledge, and cost-effective.
- Limitations: Prone to errors (hallucinations, bias), lacks common sense, and requires adaptation in workflows.

Al in Workflows

- Augmentation: All assists with tasks (e.g., video creation).
- o **Co-Creation**: Humans and Al collaborate (e.g., building presentations).
- Replacement: Al fully automates tasks (e.g., monitoring equipment).

• Steps for Al Integration

- Identify Opportunity: Find areas AI can assist.
- **Decompose Process**: Break down workflows to see where Al fits.
- Test & Scale: Trial and scale Al solutions, like generating game art.

A New Way of Working

View Al as a partner for idea generation and routine tasks. Integration takes time, like onboarding a new team member.

Progress in Generative Al

- **Collaborative Effort**: Universities, governments, open-source communities, and companies all contribute to generative AI progress, each playing a unique role.
- Universities: Lead in research and training experts, collaborating with industry and government on key Al
 projects.
- Governments and Civic Organizations: Set regulations and provide funding, with civic institutions like CIFAR creating essential datasets.
- Open-Source Communities: Offer accessible tools and models like Stable Diffusion, fueling innovation but also facing challenges with maintenance and misuse risks.
- Startups and Large Companies: Drive commercialization, integrating Al into products, publishing research, and acquiring talent for innovation.
- The Openness Challenge: Companies must balance openness to attract feedback with the risk of losing competitive edge or enabling misuse.
- Development Boundaries: Al progress is fueled by decreasing hardware costs and research, but limited by technological limits, regulations, and resource shortages.

Preparing for a Future of Generative Al

- Do More with Less: Generative AI enables small teams to achieve what larger teams once did, boosting productivity in fields like healthcare and game development.
- **The Al Divide**: Unequal access to Al tools and broadband, along with the need for Al literacy, will give more affluent individuals and nations an advantage.
- **Education and Jobs**: Al will augment tasks and reshape education, focusing on Al usage over memorization. Governments will need to bridge the Al divide and address job displacement.
- Media and Entertainment: Al speeds up creative tasks and enables large-scale media personalization, raising challenges in distinguishing real from Al-generated content.
- **Science and Technology**: Al accelerates discoveries, like Alphafold's rapid protein folding, but human direction remains essential for innovation.
- Values: As Al advances, society will need to confront ethical questions about Al's rights and trustworthiness in decision-making.