Soup to Nuts of Protecting ML Innovations





Partner

Perkins Coie LLP



Today's Presenters



Sumedha Ahuja
Partner, Perkins Coie LLP

sahuja@perkinscoie.com

Agenda

- What is Al/ML
- II. Conducting effective disclosure meetings for ML innovations
- III. Challenges to patenting Al
 - Subject Matter Eligibility (Section 101)
 - 2. Written Description and Enablement (Section 112)
 - 3. Prior-art based rejections (Section 103)

Al is Everywhere



What is Al

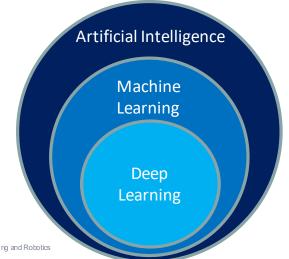


Source: Office of the Chief Economist, "Inventing AI", Number 5, October 2020

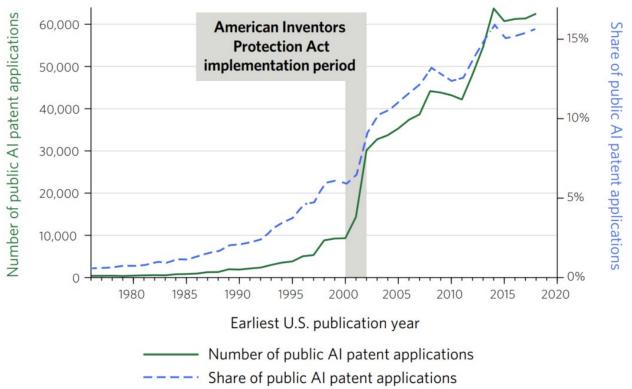
■ Many different definitions and subject to change in the future

"Software and/or hardware that can learn to solve complex problems, ... undertake tasks that require human-like ..., cognition, planning, learning, communication, or physical action" (NIST)

ML = Automatically deriving useful signals from data



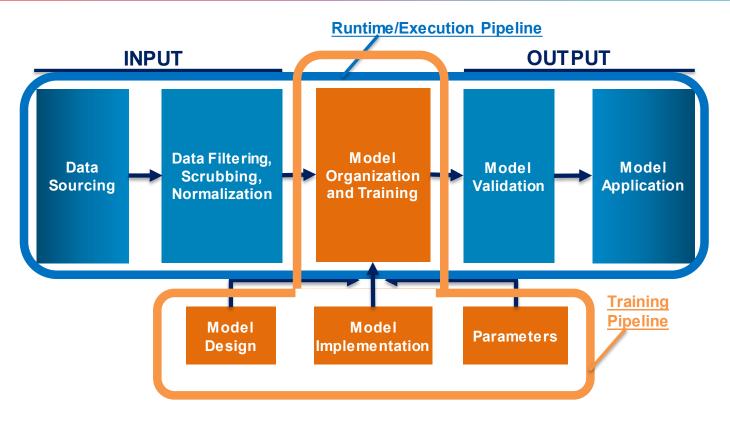
Substantial Growth of Al Patent Applications (1976-2018)



Comparing ML to Traditional Software

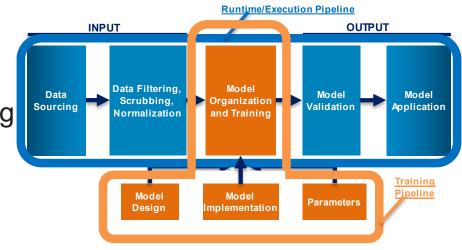
Traditional Programming Machine Learning Output Data Program Program

Pipeline View of an ML System



How to Conduct an Effective Disclosure Meeting

- 1. Location of invention
- 2. Problem being solved
- 3. Data collection and pre-processing
- 4. Model architecture and training
- 5. Post-processing steps
- 6. Output utilization



With Whom to Conduct the Disclosure Meeting

PARTICIPANTS

- High-level technical person
 CTO, CDO, Chief Scientist
- Mid-level technical person in key areas
- Sales person most responsible for product/product category



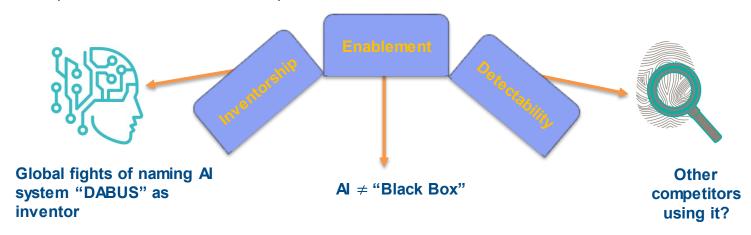
Draft Competitively

- Assess competitive value when prioritizing ML pipeline inventions:
 - Prioritize "leverage" technology that is (or will be) necessary for a competitor to compete with you
 - Carefully consider substitutions/design-around potential within the ML pipeline
 - Evaluate and claim unavoidable requirements of competitor ML pipeline solutions (even if different from your own solution)
 - Include substantial discussion of the practical application(s) (i.e., what you enable/accomplish by using the ML pipeline)

Patenting Al – Legal Challenges

For patent protection, an invention

- √ statutory (35 U.S.C. § 101)
- ✓ new
- ✓ useful
- √ non-obvious (35 U.S.C. § § 102, 103)

















Subject Matter Eligibility, Written Description, and Enablement for Machine Learning Inventions

Business Methods Partnership Meeting September 13, 2022 Steven D. Lawrenz Seed IP Law Group LLP















Subject Matter Eligibility

- Common ML claim types:
 - Process (at any stage(s) of ML pipeline)
 - Structure (of neural network)
 - Data structure:
 - Trained model
 - Training observations
 - Scoring observations













Subject Matter Eligibility

- Where available, Machine Learning Model ("MLM")
 architecture claims may be less likely to be rejected than
 MLM training and application claims:
 - Network architecture: structure of single network
 - Macro-architecture: organization of multiple networks working together
 - Micro-architecture: new type of nodes or node-combinations













Subject Matter Eligibility – Certain Methods of Organizing Human Activity

Ineligible

Generally Eligible

Practical Advice:



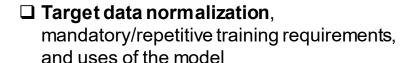
Neural network architecture; applied classification systems



☐ Avoid end results and business



Data processing; applied learning



Managing Behavior Relationships or Interactions Autonomous vehicles; loT; virtual assistants

☐ **Discuss technical difficulties** faced by existing technologies



Recommendation systemsProductivity/workflow solutions

 Describe practical or real-world applications of the claims with specificity

Financial transactions













Subject Matter Eligibility – Mental Processes and Mathematical Concepts

Ineligible

Generally Eligible

Observations, evaluations, judgements, opinions

Steps incapable of being performed by "pen and paper"



Applied formulas, equations, algorithms



- Signal processing
- Normalization (in the abstract)
- Weighing determination probabilities
- Weighing activation of nodes
- Solutions imitating/simulating human behavior

Practical Advice:

- □ Avoid behavior and decision-making capable of being performed entirely in a human's mind
- □ Explain necessity of a digital solution, and specify hardware components
- ☐ Generalize the ML model within the claim, to focus upon input/output novelty
- ☐ Patent Offices narrowly construe claimed algorithms (in general)
- □ Algorithmic claims may be necessary for certain inventions (e.g., codecs, standards, etc.)













Written Description and Enablement

- Claimed aspects:
 - Very little guidance exists about the level of disclosure needed for machine learning claims recited training, storing, or applying a MLM
 - A conservative approach is to, where pursuing these claims, include such detail as:
 - model architecture—either a diagram, or an incorporation by reference of an article effectively describing a common architecture used
 - data dimensionality, and other details of training observation contents and organization
 - any nonstandard aspects of training scheme











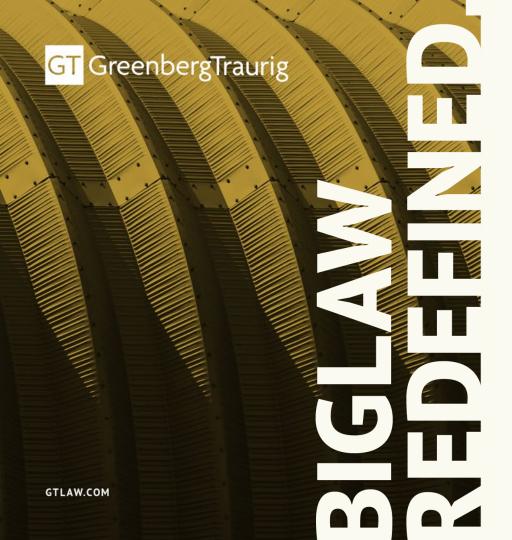


Written Description and Enablement

- Aspects initially unclaimed:
 - It may be helpful to be able to add details during examination, particularly in response to eligibility rejections:
 - further detail about existing stage(s) of ML pipeline
 - detail about additional stage(s) of ML pipeline
 - Including well-crafted boilerplate content describing the entire
 ML pipeline at a reasonable level of detail can provide support for such amendments



Seed



Rejections under 35 U.S.C. 103 for claims reciting Machine-Learning

GT GreenbergTraurig



Barry J. Schindler
Co-Chair, Global Patent
Prosecution Group
Greenberg Traurig, LLP
500 Campus Drive, Suite 400
Florham Park, NJ 07932
Work: 973.360.7944
Cell: 973.5194.944
schindlerb@gtlaw.com



Lennie A. Bersh
Shareholder
Greenberg Traurig, LLP
500 Campus Drive, Suite 400
Florham Park, NJ 07932
Work: 973.443.3536
Cell: 212.767.9141
bershl@gtlaw.com

GT GreenbergTraurig

Prima Facie Obviousness Based On ML Teachings: Rejections under 35 U.S.C. 103 for claims reciting Machine-Learning Models (MLMs)

- MPEP is silent
- Guiding Principles:
 - A ML Pipeline has particular objective(s) (e.g., solving a particular problem)
 - The Goal of MLM design, training and re-training is to satisfy such objective(s)

- Guiding Principles (cont.):
 - Three key factors determine the accuracy of MLMs:
 - fit (structure, inputs, outputs) of MLM: selecting a right MLM out of hundreds and hundreds MLMs
 - completeness of input data (e.g., feature vectors): the degree to which the number of data points required to reach a defined accuracy threshold has been provided
 - sufficiency of training data
 - For example, if MLM is trained to identify cars from an image, and the input data consists of photos of airplanes, the model would not know what a car looks like. Such MLM will not provide good results.

 - Multiple-Model techniques require a defined relationship between MLMs

- Common *Prima Facie* Obviousness Rationales (MPEP 2143):
 - Combine prior art MLMs according to known methods to yield predictable results; or
 - Substitute of one prior art MLM for another prior art MLM to obtain predictable results; or
 - Modify prior art MLM(s) "to arrive at the claimed invention" with "reasonable expectation of success"

- Meeting prima facie based on a combination of prior art MLMs according to known methods to yield predictable results:
 - Identify a known method on how to combine:
 - Output of MLM(1) as input for MLM(2); or
 - Apply a voting to outputs from different MLMs to obtain a common output; or
 - Apply a weighting function to numerical outputs from different MLMs to obtain a common score

AND

- Provide evidentiary support and/or technical reasoning as to why results would be predictable e.g.:
 - How would the combination of prior art MLMs be trained to achieve an objective of either reference? or
 - In case when the combination is based on output of MLM(1) as input for MLM(2), how would output of MLM(1) meet the completeness for input data for MLM (2) and be related to output of MLM(2)?

- Substitute of MLM(1) for another MLM(2) to obtain predictable results:
 - Provide evidentiary support and/or technical reasoning for the substitution – e.g.:
 - Why would MLM(1) be fit to achieve the same objective(s) of MLM(2)? or
 - Could MLM(1) be trained with inputs of MLM(2)? i.e., what is/are difference(s)/similarity(ies) between inputs of MLM(1) and MLM(2)?

- Modify a MLM "to arrive at the claimed invention" with "reasonable expectation of success":
 - Provide evidentiary support and/or technical reasoning for the modification e.g.:
 - Why would the modified MLM still fit so as to achieve the original objective(s)? or
 - How would the modified MLM be trained with original and/or modified input(s) to achieve the original objective(s)?

PTAB decisions:

- Examiner was affirmed when references in the substitutionbased combination taught that their MLMs consumed similar inputs and were trained to achieve similar objectives
- Examiner was reversed in the modification-based combination when the Board concluded that there was a lack of explanation as to why one would modify one MLM based on a structure of another MLM when those models provided unrelated outputs (i.e., models were designed for unrelated objectives)

Examiners' Obviousness "Toolbox"

- Combine prior art MLMs according to known methods to yield predictable results; or
- Substitute of one prior art MLM for another prior art MLM to obtain predictable results; or
- Modify prior art MLM(s) "to arrive at the claimed invention" with "reasonable expectation of success"

Suggested Analytical Framework -- Genus-Species Analysis

- MPEP 2144.08: Obviousness of Species When Prior Art Teaches Genus
 - "In the case of a prior art reference disclosing a genus, Office personnel should make findings as to:
 - (A) the structure of the disclosed prior art genus and that of any expressly described species or subgenus within the genus;
 - (B) any physical or chemical properties and utilities disclosed for the genus, as well as any suggested limitations on the usefulness of the genus, and any problems alleged to be addressed by the genus;
 - (C) the predictability of the technology; and
 - (D) the number of species encompassed by the genus taking into consideration all of the variables possible."

Genus-Species Analysis (MPEP 2144.08) -- Not to re-invent a proverbial wheel

- Potential factors to consider:
 - the structure of disclosed prior art MLM genus and that of any expressly described MLM species or subgenus within the MLM genus;
 - any similarity in design, training, and/or objective(s) addressed by the MLM genus/subgenus;
 - the number of MLM species encompassed by the MLM genus taking into consideration all of the variables possible.