

# Chapter 1

## Domains

### RL

- Agent that performs actions at every time step
- World is not necessarily deterministic
- Actions might or might not provide a reward
  - Temporal credit assignment
    - Reward may occur after the action
  - Exploration vs exploitation
    - Trade off between trying new things vs using it already knows
    - Not really obvious how this works directly
      - Policy network is used
- Eg
  - Robot learning to walk
  - Chess

### Semi supervised

- Partially labelled / semi automatic

### Unsupervised

- Generative models
  - No labels
  - Synthesize new examples that are statistically indistinguishable from the training data
  - Might or might not explicitly describe the probability distribution over the input data

### Supervised

- Requires labeled input/output pairs for training

## Data Processing

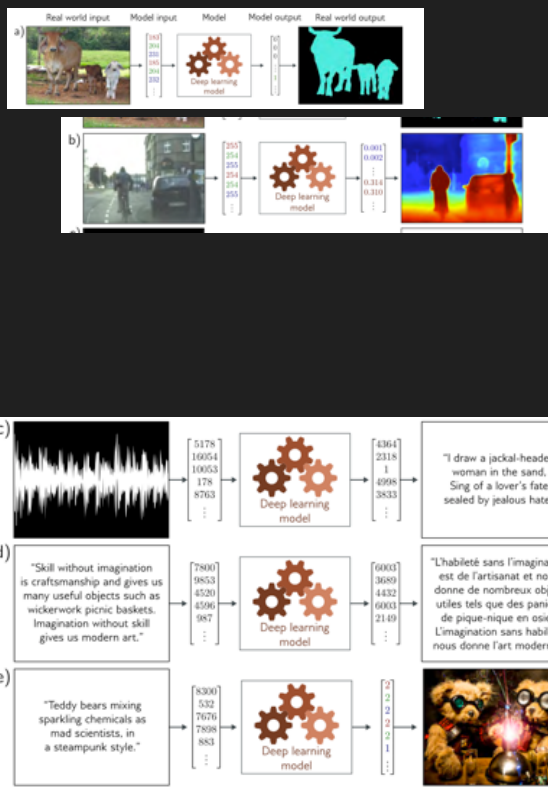
- Audio
  - Sampled at 44.1 kHz
  - 16 bit integers
- Images
  - Concatenated RGB values per pixel
- Chemical data
  - Geometric structure of molecules + additional information per atom

## Deep Neural Networks

- Output types (some)
  - Single real numbers
  - Multiple numbers
  - Probabilities

## Types of Problems

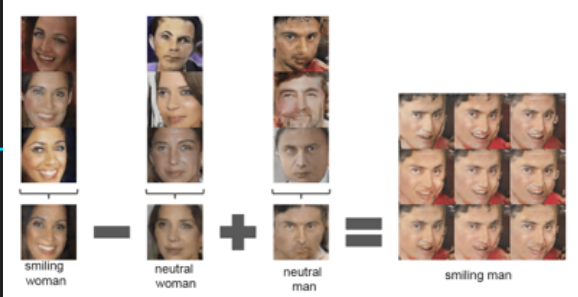
- Regression
  - Multivariate
- Classification
  - Binary
    - Segmentation
  - Multiclass
    - Segmentation



## Latent variables

### eg

- Describing each data using a smaller number of underlying values
- Real world images can be generated from random RGB pixel values
- A face is only a small subset of possibilities



(This is not from the book but I found it useful a few years ago)

### Uses

- Image interpolation
- Caption -> Image

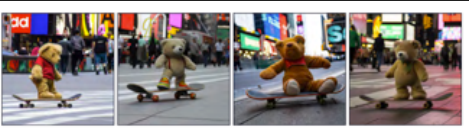


Figure 1.12 Multiple images generated from the caption "A teddy bear on a skateboard in Times Square" Generated by DALL-E2 (Ramesh et al., 2022).

Manifold hypothesis (also not from the book but massively helped me understand latent variables)

<https://www.youtube.com/watch?v=BePQBWPnYuE>

### Connects Supervised + unsupervised

Instead of a direct input -> output mapping

### Why

- Learn relationship between latent variables
- Need less text/image pairs
- Lower dimensional
- Add Randomness
- Generate multiple images easily

## Ethics

- Bias
  - Historical bias
- Explainability
  - Somewhat easy to produce local explanations
  - Still hard to build a complex transparent system
- Weaponising
  - Existential risk
- Concentrating power
  - Companies can be assholes
  - Automating jobs



Some models can have additional constraints (eg: "diamond" ring as opposed to just a ring)

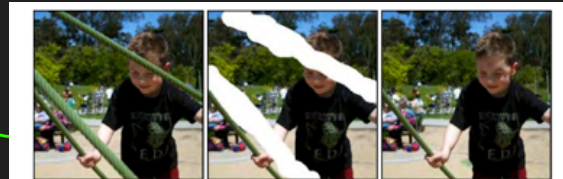


Figure 1.17 Inpainting. In the original image (left), the boy is obscured by metal calipers. These indistinct regions (center) are removed and the generative model synthesizes a new image (right) under the constraint that the remaining pixels must stay the same. Adapted from Subbar et al., (2020a).

- Image generation
  - Conditional
- Chat GPT
  - Text generation
  - Conditional text synthesis

Inpainting