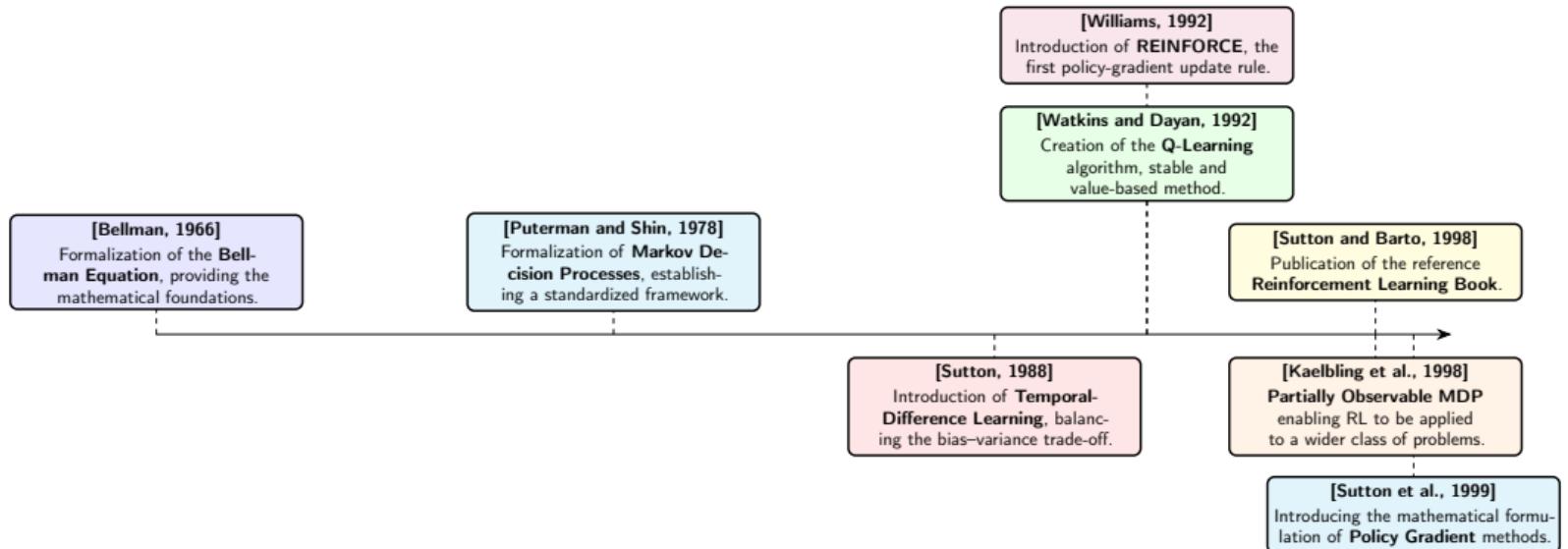
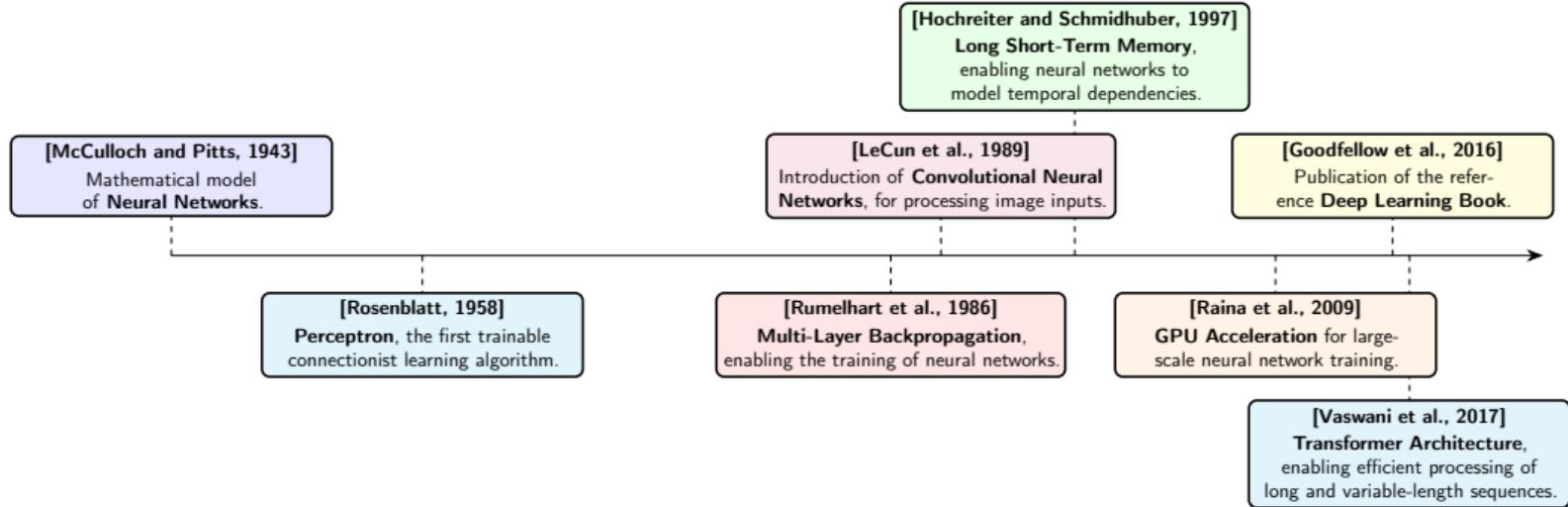
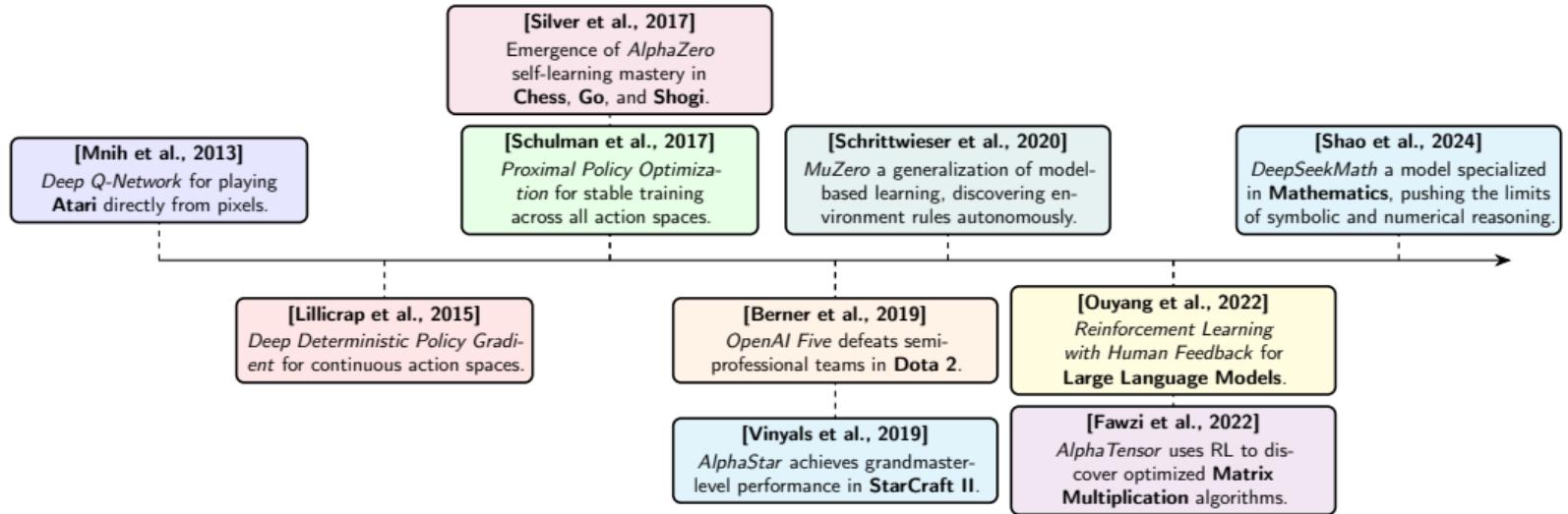


RLLib: Industry-Grade, Scalable Reinforcement Learning

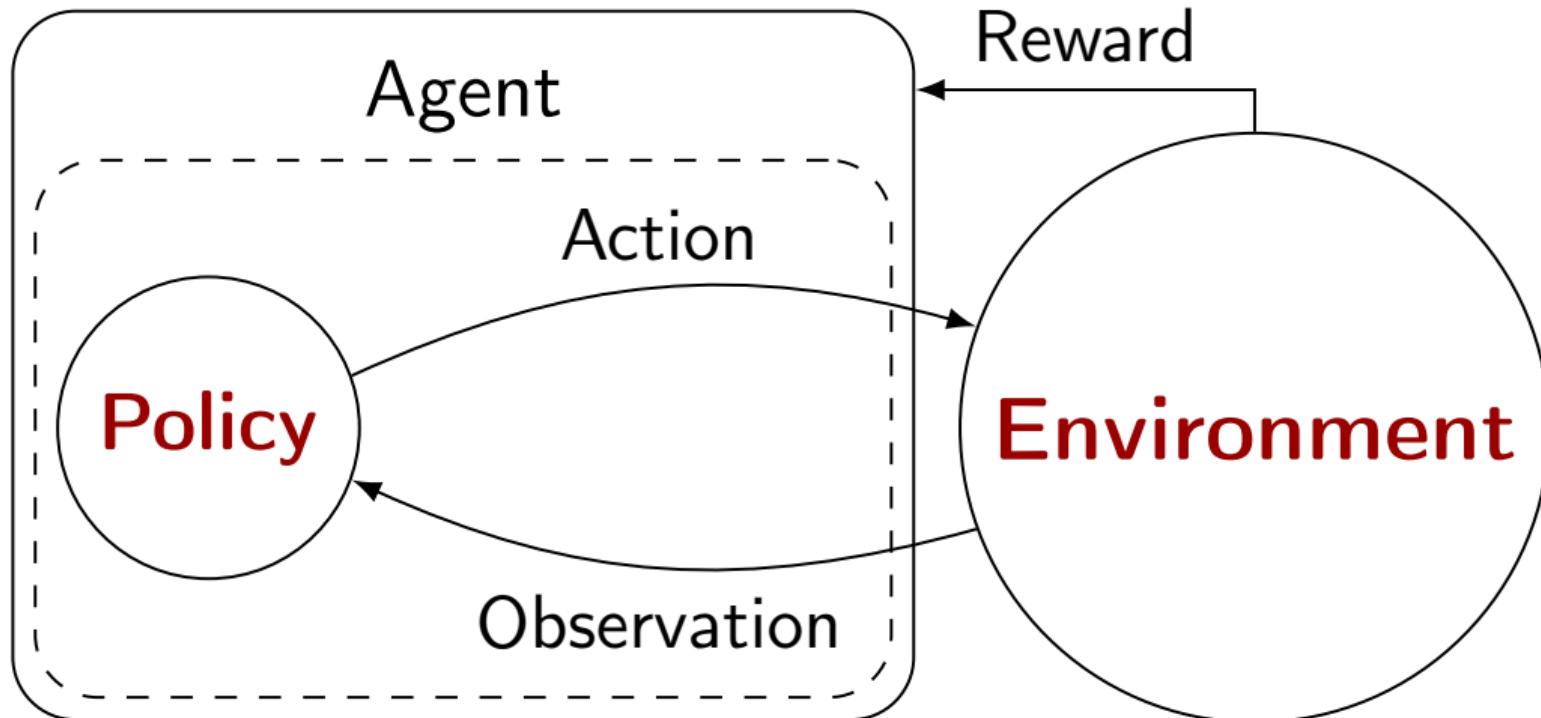
Maxime Alaarabiou







RL Algorithm



Gymnasium:

```
1 import gymnasium as gym
2
3 env = gym.make("ALE/SpaceInvaders-v5", render_mode="rgb_array")
4 obs, info = env.reset()
5 action = env.action_space.sample()
6 obs, reward, terminated, truncated, info = env.step(action)
```

PettingZoo:

```
1 from pettingzoo.atari import space_invaders_v2
2
3 env = space_invaders_v2.env(render_mode="rgb_array")
4 env.reset()
5 for agent in env.agent_iter():
6     obs, reward, term, trunc, info = env.last()
7     action = env.action_space(agent).sample() if not term else None
8     env.step(action)
```

Environment Configuration	
game_name	'SpaceInvaders-v5'
repeat_action_probability	0.05
frameskip	5
resize_observation_shape	(64, 64)
convert_to_grayscale	True
reward_scale_factor	0.05
frame_stack_len	4
normalize_observation	True
observation_numpy_type	np.float16
Policy Architecture Configuration	
architecture	CnnPPO
configuration_cnn	[(16,4,2),(32,4,2),(64,4,2),(128,4,2)]
configuration_hidden_layers	[512,256,128]
activation_function_class	LeakyReLU
use_layer_normalization_cnn	True
use_share_cnn	True
Reinforcement Learning Configuration	
algorithm_name	'PPO'
rollout_fragment_length	2048
train_batch_size	2048 * 8
minibatch_size	2048
lambda_gae	0.95
kullback_leibler_coefficient	0.5
clip_policy_parameter	0.1
clip_value_function_parameter	10
entropy_coefficient	0.01
number_epochs	10
learning_rate	0.00015
gradient_clip	100.0
gradient_clip_by	'global_norm'

Why choose RLlib?

Open-source

Scalable

DQN, DDPG, PPO, Dreamer

PyTorch, TensorFlow

Automatic Plotting

Automatic Checkpointing and Restart

Multi-agent, Hierarchical

Advanced Callback System

Exploration, Curriculum Learning, Custom RL Algorithms

```
1 config = PPOConfig()
2 config.environment("SpaceInvaders")
3 config.rl_module(
4     model_config={
5         "conv_filters": [
6             [32, [8, 8], 4],
7             [64, [4, 4], 2],
8             [64, [3, 3], 1]
9         ],
10    }
11 )
12
13 algo = config.build()
14 for _ in range(5):
15     pprint(algo.train())
16
17 pprint(algo.evaluate())
18 algo.stop()
```

Thank you for your attention

Questions, Comments, Ideas?

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