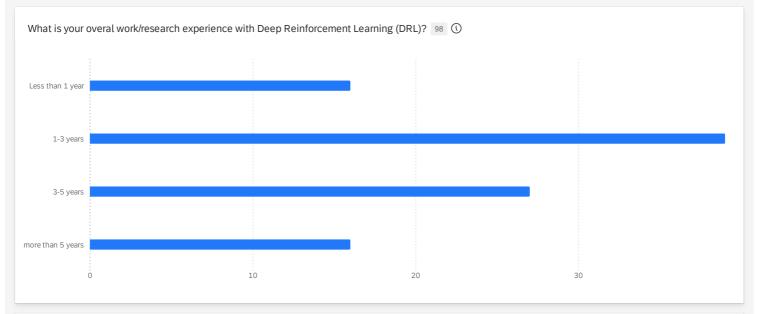
| Software developer |
|---------------------------|
| Researcher |
| Research Assistant |
| Data Scientist |
| Research Engineer |
| PhD student |
| Researcher |
| Software Engineer |
| Researcher |
| Product Manager |
| Research engineer |
| PhD student |
| Machine Learning Engineer |
| Phd student |
| undergraduate |
| PhD student |
| Manager |
| iOS Developer |
| Student |
| Phd |
| Master |
| Data Scientist |
| Research Associate |
| Bachelor |

| Researcher |
|--|
| Master's Student - Research Assistant |
| Master |
| Phd student |
| Associate Professor |
| Researcher |
| Bachelor Student |
| Master of Science in Machine Learning graduate student |
| PhD student |
| PhD student |
| Product owner |
| PhD Student |
| Developer |
| Developer |
| Research Associate |
| Researcher |
| Researcher |
| Senior Data Scientist |
| Machine Learning Engineer |
| Data Engineer |
| ML Engineer |
| Machine learning engineer |
| Machine Learning Engineer |
| Researcher |

| PhD student |
|---|
| Al Engineer |
| PhD student |
| Machine learning engineer |
| Master student |
| AI Scientist |
| Senior Data Scientist |
| Postdoc Fellow |
| Reseach Associate |
| Data Scientist |
| PhD student |
| Software Engineer |
| PhD student |
| Researcher |
| Professor |
| Master |
| researcher in the RoboFEI robotics team, graduate student |
| PhD candidate |
| Machine Learning Engineer |
| Researcher |
| Developer |
| Master Graduate |
| Full-stack software engineer |
| Data Scientist |

| Founder, Al startup (Cloudcraftz.com) |
|--|
| Assistant Professor |
| Deep Learning practitioner |
| Research Associate |
| Developer and Researcher |
| PhD Candidate |
| Developer |
| Unemployed, but I guess self-employed researcher |
| Master |
| Local Agent of Innovation |
| Master |
| Master |
| Master, Machine learning engineer |
| Research Scientist |
| Senior Data Scientist |
| Software Engineer |
| ML Research Engineer |
| PhD student |
| Research |
| Machine Learning Engineer |
| PhD Student Tech Manager & Senior Software Developer |
| PhD Student |
| Assistant professor |
| Developer/Researcher |



| What is your overal work/research experience with Deep Reinforcement Learning (DRL) | ? 98 🛈 | |
|---|------------|-------|
| Q3 - What is your overal work/research experience with Deep Reinforcement Learning (DRL)? | Percentage | Count |
| Less than 1 year | 16% | 16 |
| 1-3 years | 40% | 39 |
| 3-5 years | 28% | 27 |
| more than 5 years | 16% | 16 |

| What is your overal work/research experience with Deep Reinforcement Learning (DRL)? 98 (1) | | | |
|---|----------------------|--|--|
| Average | Minimum | Maximum | Count |
| 1.00 | 1.00 | 1.00 | 16 |
| 2.00 | 2.00 | 2.00 | 39 |
| 3.00 | 3.00 | 3.00 | 27 |
| 4.00 | 4.00 | 4.00 | 16 |
| | 1.00 2.00 3.00 | Average Minimum 1.00 1.00 2.00 2.00 3.00 3.00 | Average Minimum Maximum 1.00 1.00 1.00 2.00 2.00 2.00 3.00 3.00 3.00 |

| Python | | | |
|--------|--|--|--|
| Python | | | |
| | | | |

| Python |
|-------------------|
| C/C++, Python |
| C/C++, Pythn |
| Python C# C++ |
| Python |
| Java Swift python |
| python |
| Python |
| Python |
| Python |
| Python |
| Python |
| Python, C# |
| Python |
| Python, C++ |
| Python |
| C/C++, Python |

| Python |
|--------------------------|
| Python |
| Java, c/c++,python,scala |
| python, c# |
| Python |
| Python |
| Python, Rust |
| Python and Cython |
| Pyhton |
| Python |
| Python |
| Python |
| Python |
| Python C++ |
| Python |
| Python, C++ |
| Python, C++ |
| Python, Java |

| Python |
|------------------|
| Python C++ |
| Python |
| C++, Python |
| Python |
| Python |
| Python |
| Python |
| python |
| python |
| Python, C++ |
| Python, Makefile |
| Python |
| Python, C++, C# |
| Python |
| Python, C++ |

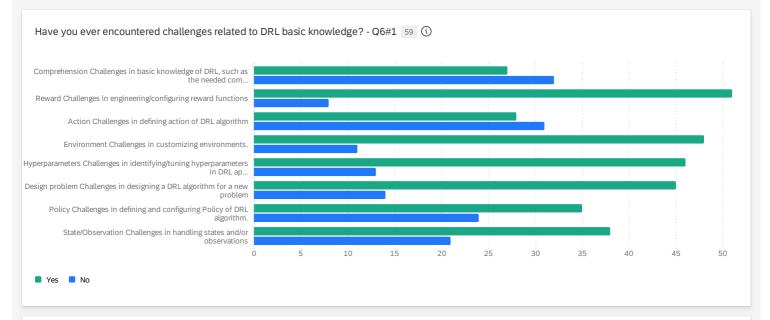
| Python |
|-------------------------|
| C++, Python |
| Python |
| Python |
| Python, Javascript |
| Python and C++ |
| Python, C++ |
| Python |
| Mostly python, some C++ |
| Python |
| C++, Python |
| Python |
| Python |
| Python |
| Python |
| Python |
| Python |
| python |
| C++ / Python |
| |
| Tensorflow, keras |
| Tensorflow |
| Pytorch |
| Pytorch |

| PyTorch, Transhou, Stable-baselines 3 |
|---|
| PyTroch, TesnorFlow |
| PyTorch |
| PyTorch |
| PyTorch |
| PyTorch, Tensorflow and keras |
| Tensorflow Keras PyTorch Tensorforce RLLib CleanRL Others |
| PyTorch + OpenAI Gym library |
| Tensorflow, PyTorch |
| PyTorch, Jax |
| Pytorch |
| PyTorch |
| Gym |
| PyTorch |
| Tensorflow, keras, PyTorch, Keras-rl, Python-rl |
| Tensorflow keras |
| pytorch |
| Pytorch, Tensorflow, Keras |
| Tensorflow, keras, pytorch, stable-baselines 3, clean-rl |
| Tensorflow, keras, PyTorch |
| PyTorch, PyTorch-RL |
| Keras, PyTorch |
| PyTorch, Unity ML-Agents |
| PyTorch, stable baselines 3, OpenAI gym, Open AI baselines, tf agents |

| Tensorflow, keras, PyTorch |
|---|
| Tensorflow, PyTorch |
| PyTorch |
| Tensorflow, PyTorch |
| Tensorflow, keras, spacy |
| Tensorflow, keras, PyTorch |
| Pytorch |
| Jax, PyTorch |
| Tensorflow, keras |
| PyTorch, Tensorflow |
| Tensorflow、keras、PyTorch |
| Tensorflow,kereas,pytorch |
| Tensorflow, keras |
| PyTorch |
| PyTorch |
| PyTorch, Jax |
| Tensorflow, Keras, Keras-RL, Baselines, StableBaselines, TRFL |
| PyTorch |
| PyTorch |
| Tensorflow, keras, PyTorch |
| PyTorch, Gym |
| Tensorflow, Keras, pytorch. |
| Pytorch Tensorflow JAX |
| PyTorch, Tensorflow |

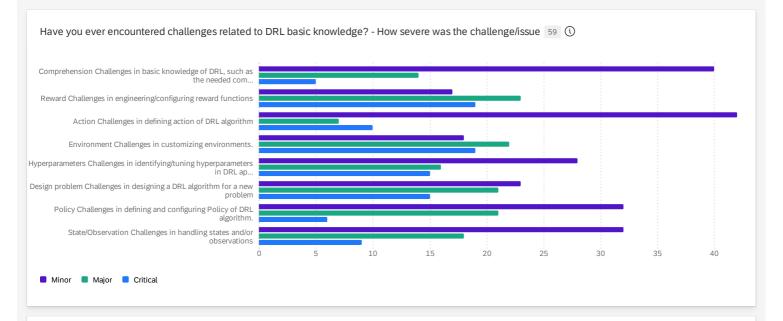
| PyTorch, TensorFlow |
|---|
| PyTorch, Tensorflow |
| PyTorch, Tensorflow |
| PyTorch, some Tensorflow |
| PyTorch, Tensorflow |
| PyTorch |
| Tensorflow, Keras, Pytorch |
| PyTorch, Tensorflow |
| Tensorflow, keras, keras-rl |
| PyTorch |
| Keras, PyTorch, Jax |
| PyTorch |
| pytorch |
| Tensorflow, keras, PyTorch, ROS, Gazebo |
| Pytorch |
| Tensorflow |
| PyTorch |
| PyTorch, Keras |
| Tensorflow |
| PyTorch, Keras, Tensorflow |
| TensorFlow and PyTorch |
| Pytorch, keras |
| Tensorflow, PyTorch, Python-rl |
| Tensorflow, PyTorch, gym |

| TensorFlow, Keras |
|--|
| PyTorch |
| Pytorch |
| PyTorch, Keras |
| Mainly PyTorch. Used all the others but they're either too high level of abstraction where you're not really doing anything intelligent (i.e. PyTorchLightning), or if they are a bit closer to the metal, they're just not as convenient as PyTorch |
| PyTorch, stable-baselines |
| Stable Baselines |
| Mostly keras-rl |
| PyTorch, TensorFlow, Keras, Cherry-RL |
| Tensorflow, Keras, PyTorch |
| Pytorch |
| pytorch, jax, haiku, acme, other proprietary frameworks |
| keras-rl |
| Tensorflow, Keras |
| PyTorch, Ray + RLlib, OpenAI baselines, Unity ML-agents |
| PyTorch |
| PyTorch |
| Tensorflow, Pytorch |
| Tensorflow, Keras and PyTorch |
| PyTorch Stable-baselines3 RLlib |
| torch, tensorflow, keras |
| PyTorch/ Tensorflow |



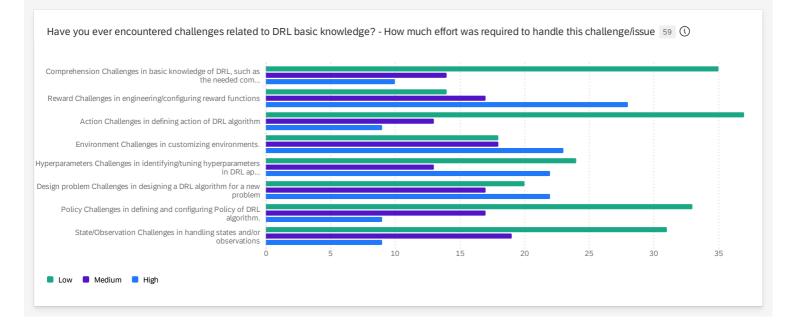
| Have you ever encountered challenges related to DRL basic knowledge? - Q6#1 59 (i) | | |
|--|-----|----|
| Q6#1 | Yes | No |
| Comprehension Challenges in basic knowledge of DRL, such as the needed com | 27 | 32 |
| Reward Challenges in engineering/configuring reward functions | 51 | 8 |
| Action Challenges in defining action of DRL algorithm | 28 | 31 |
| Environment Challenges in customizing environments. | 48 | 11 |
| Hyperparameters Challenges in identifying/tuning hyperparameters in DRL ap | 46 | 13 |
| Design problem Challenges in designing a DRL algorithm for a new problem | 45 | 14 |
| Policy Challenges in defining and configuring Policy of DRL algorithm. | 35 | 24 |
| State/Observation Challenges in handling states and/or observations | 38 | 21 |
| | | |

| Q6#1 | Average | Minimum | Maximum | Cour |
|--|---------|---------|---------|------|
| Comprehension Challenges in basic knowledge of DRL, such as the needed com | 1.54 | 1.00 | 2.00 | Ę |
| Reward Challenges in engineering/configuring reward functions | 1.14 | 1.00 | 2.00 | Ę |
| Action Challenges in defining action of DRL algorithm | 1.53 | 1.00 | 2.00 | Ę |
| Environment Challenges in customizing environments. | 1.19 | 1.00 | 2.00 | Ę |
| Hyperparameters Challenges in identifying/tuning hyperparameters in DRL ap | 1.22 | 1.00 | 2.00 | Ę |
| Design problem Challenges in designing a DRL algorithm for a new problem | 1.24 | 1.00 | 2.00 | Ę |
| Policy Challenges in defining and configuring Policy of DRL algorithm. | 1.41 | 1.00 | 2.00 | Ę |
| State/Observation Challenges in handling states and/or | 1.36 | 1.00 | 2.00 | į |



| Have you ever encountered challenges related to DRL basic knowledge? - How severe was the challenge/issue 59 (1) | | | | | |
|--|-----------------------------------|---|--|--|--|
| Minor | Major | Critical | | | |
| 40 | 14 | 5 | | | |
| 17 | 23 | 19 | | | |
| 42 | 7 | 10 | | | |
| 18 | 22 | 19 | | | |
| 28 | 16 | 15 | | | |
| 23 | 21 | 15 | | | |
| 32 | 21 | (| | | |
| 32 | 18 | , | | | |
| | Minor 40 17 42 18 28 23 32 | Minor Major 40 14 17 23 42 7 18 22 28 16 23 21 32 21 | | | |

| Have you ever encountered challenges related to DRL basic knowledge? - How severe was the challenge/issue 59 🛈 | | | | |
|--|---------|---------|---------|-------|
| How severe was the challenge/issue | Average | Minimum | Maximum | Count |
| Comprehension Challenges in basic knowledge of DRL, such as the needed com | 1.41 | 1.00 | 3.00 | 59 |
| Reward Challenges in engineering/configuring reward functions | 2.03 | 1.00 | 3.00 | 59 |
| Action Challenges in defining action of DRL algorithm | 1.46 | 1.00 | 3.00 | 59 |
| Environment Challenges in customizing environments. | 2.02 | 1.00 | 3.00 | 59 |
| Hyperparameters Challenges in identifying/tuning hyperparameters in DRL ap | 1.78 | 1.00 | 3.00 | 59 |
| Design problem Challenges in designing a DRL algorithm for a new problem | 1.86 | 1.00 | 3.00 | 59 |
| Policy Challenges in defining and configuring Policy of DRL algorithm. | 1.56 | 1.00 | 3.00 | 59 |
| State/Observation Challenges in handling states and/or observations | 1.61 | 1.00 | 3.00 | 59 |
| | | | | |



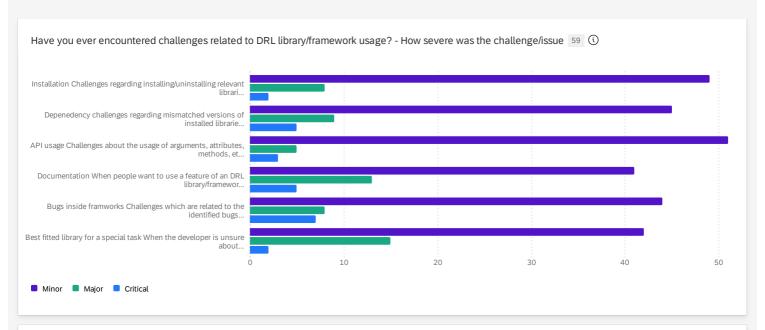
| Have you ever encountered challenges related to DRL basic knowledge? - How much effort was required to handle this challenge/issue 59 ① | | | | |
|---|----------------------------------|---|--|--|
| Low | Medium | High | | |
| 35 | 14 | 10 | | |
| 14 | 17 | 28 | | |
| 37 | 13 | 9 | | |
| 18 | 18 | 23 | | |
| 24 | 13 | 22 | | |
| 20 | 17 | 22 | | |
| 33 | 17 | 9 | | |
| 31 | 19 | 9 | | |
| | 35 14 37 18 24 20 | 35 14 14 17 37 13 18 18 24 13 20 17 33 17 | | |

| How much effort was required to handle this challenge/issue | Average | Minimum | Maximum | Cou |
|---|---------|---------|---------|-----|
| Comprehension Challenges in basic knowledge of DRL, such as the needed com | 1.58 | 1.00 | 3.00 | ! |
| Reward Challenges in engineering/configuring reward functions | 2.24 | 1.00 | 3.00 | į |
| Action Challenges in defining action of DRL algorithm | 1.53 | 1.00 | 3.00 | í |
| Environment Challenges in customizing environments. | 2.08 | 1.00 | 3.00 | ! |
| Hyperparameters Challenges in identifying/tuning hyperparameters in DRL ap | 1.97 | 1.00 | 3.00 | į |
| Design problem Challenges in designing a DRL algorithm for a new problem | 2.03 | 1.00 | 3.00 | į |
| Policy Challenges in defining and configuring Policy of DRL algorithm. | 1.59 | 1.00 | 3.00 | ! |
| State/Observation Challenges in handling states and/or | 1.63 | 1.00 | 3.00 | į |



| Have you ever encountered challenges related to DRL library/framework usage? - Q7#1 59 ① | | |
|--|-----|----|
| Q7#1 | Yes | No |
| Installation Challenges regarding installing/uninstalling relevant librari | 24 | 35 |
| Dependency challenges regarding mismatched versions of installed librarie | 31 | 28 |
| API usage Challenges about the usage of arguments, attributes, methods, et | 22 | 37 |
| Documentation When people want to use a feature of an DRL library/framewor | 29 | 30 |
| Bugs inside framworks Challenges which are related to the identified bugs | 29 | 30 |
| Best fitted library for a special task When the developer is unsure about | 30 | 29 |

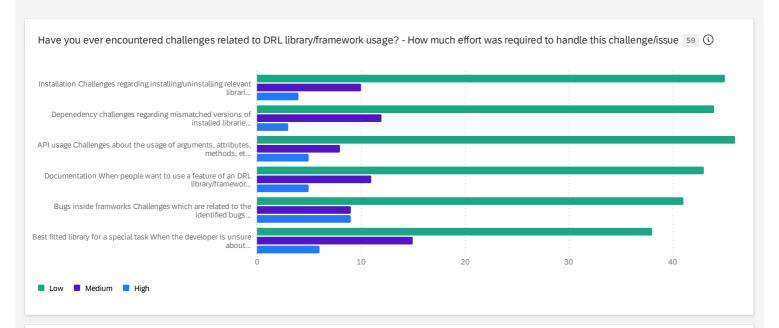
| 7#1 | Average | Minimum | Maximum | Cour |
|---|---------|---------|---------|------|
| nstallation Challenges regarding installing/uninstalling elevant librari | 1.59 | 1.00 | 2.00 | Ę |
| Depenedency challenges regarding mismatched versions of installed librarie | 1.47 | 1.00 | 2.00 | ! |
| API usage Challenges about the usage of arguments, ittributes, methods, et | 1.63 | 1.00 | 2.00 | ! |
| Documentation When people want to use a feature of in DRL library/framewor | 1.51 | 1.00 | 2.00 | ! |
| Bugs inside framworks Challenges which are related to he identified bugs | 1.51 | 1.00 | 2.00 | ! |
| Best fitted library for a special task When the developer sunsure about | 1.49 | 1.00 | 2.00 | Ę |



| Have you ever encountered challenges related to DRL library/framework usage? - How severe was the challenge/issue 59 (1) | | | | |
|--|-------|-------|----------|--|
| How severe was the challenge/issue | Minor | Major | Critical | |
| Installation Challenges regarding installing/uninstalling relevant librari | 49 | 8 | 2 | |
| | | | | |

| How severe was the challenge/issue | Minor | Major | Critical |
|--|-------|-------|----------|
| Dependency challenges regarding mismatched versions of installed librarie | 45 | 9 | 5 |
| API usage Challenges about the usage of arguments, attributes, methods, et $% \label{eq:polyage} % \label{eq:polyage}$ | 51 | 5 | 3 |
| Documentation When people want to use a feature of an DRL library/framewor | 41 | 13 | 5 |
| Bugs inside framworks Challenges which are related to the identified bugs | 44 | 8 | 7 |
| Best fitted library for a special task When the developer is unsure about | 42 | 15 | 2 |
| | | | |

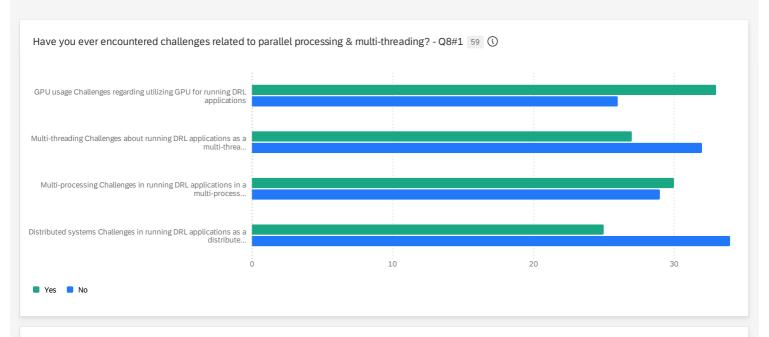
| How severe was the challenge/issue | Average | Minimum | Maximum | Cour |
|--|---------|---------|---------|------|
| installation Challenges regarding installing/uninstalling relevant librari | 1.20 | 1.00 | 3.00 | Ę |
| Depenedency challenges regarding mismatched versions of installed librarie | 1.32 | 1.00 | 3.00 | į |
| API usage Challenges about the usage of arguments, attributes, methods, et | 1.19 | 1.00 | 3.00 | į |
| Documentation When people want to use a feature of an DRL library/framewor | 1.39 | 1.00 | 3.00 | ţ |
| Bugs inside framworks Challenges which are related to the identified bugs | 1.37 | 1.00 | 3.00 | į |
| Best fitted library for a special task When the developer s unsure about | 1.32 | 1.00 | 3.00 | 5 |



| Have you ever encountered challenges related to DRL library/framework usage? - How much effort was required to handle this challenge/issue 59 (1) | | | | |
|---|-----|--------|------|--|
| How much effort was required to handle this challenge/issue | Low | Medium | High | |
| Installation Challenges regarding installing/uninstalling relevant librari | 45 | 10 | 4 | |
| Dependeency challenges regarding mismatched versions of installed librarie | 44 | 12 | 3 | |

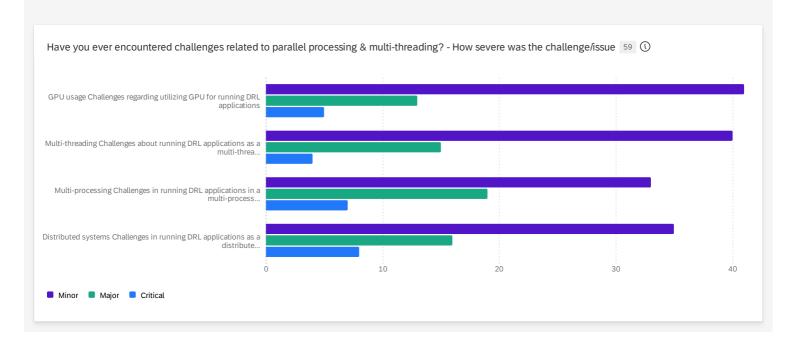
| How much effort was required to handle this challenge/issue | Low | Medium | High |
|--|-----|--------|------|
| API usage Challenges about the usage of arguments, attributes, methods, et | 46 | 8 | 5 |
| Documentation When people want to use a feature of an DRL library/framewor | 43 | 11 | 5 |
| Bugs inside framworks Challenges which are related to the identified bugs | 41 | 9 | 9 |
| Best fitted library for a special task When the developer is unsure about | 38 | 15 | 6 |
| | | | |

| rary/mamework asage: Trow | much enone was required to | handle this challenge/issue | 59 (1) |
|---------------------------|---------------------------------------|--|--|
| Average | Minimum | Maximum | Count |
| 1.31 | 1.00 | 3.00 | 59 |
| 1.31 | 1.00 | 3.00 | 59 |
| 1.31 | 1.00 | 3.00 | 59 |
| 1.36 | 1.00 | 3.00 | 59 |
| 1.46 | 1.00 | 3.00 | 59 |
| 1.46 | 1.00 | 3.00 | 59 |
| | Average 1.31 1.31 1.31 1.36 1.46 | Average Minimum 1.31 1.00 1.31 1.00 1.31 1.00 1.36 1.00 1.46 1.00 | Average Minimum Maximum 1.31 1.00 3.00 1.31 1.00 3.00 1.31 1.00 3.00 1.36 1.00 3.00 1.46 1.00 3.00 |



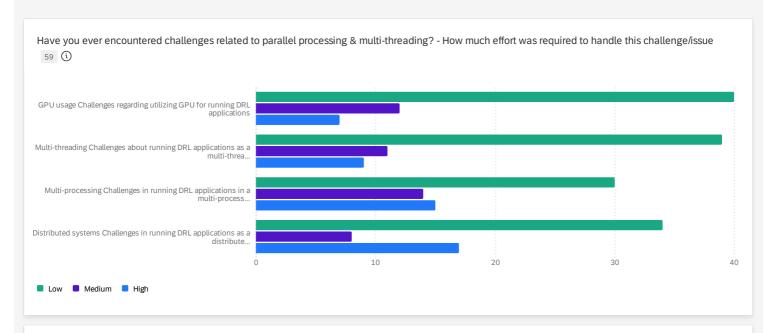
| Have you ever encountered challenges related to parallel processing & multi-threading? - Q8#1 59 🛈 | | |
|--|-----|----|
| Q8#1 | Yes | No |
| GPU usage Challenges regarding utilizing GPU for running DRL applications | 33 | 26 |
| Multi-threading Challenges about running DRL applications as a multi-threa | 27 | 32 |
| Multi-processing Challenges in running DRL applications in a multi-process | 30 | 29 |
| Distributed systems Challenges in running DRL applications as a distribute | 25 | 34 |
| | | |

| 28#1 | Average | Minimum | Maximum | Cour |
|--|---------|---------|---------|------|
| GPU usage Challenges regarding utilizing GPU for running DRL applications | 1.44 | 1.00 | 2.00 | Ę |
| Multi-threading Challenges about running DRL applications as a multi- hrea | 1.54 | 1.00 | 2.00 | Ę |
| Multi-processing Challenges in running DRL applications in a multi-process | 1.49 | 1.00 | 2.00 | į |
| Distributed systems Challenges in running DRL applications as a distribute | 1.58 | 1.00 | 2.00 | į |



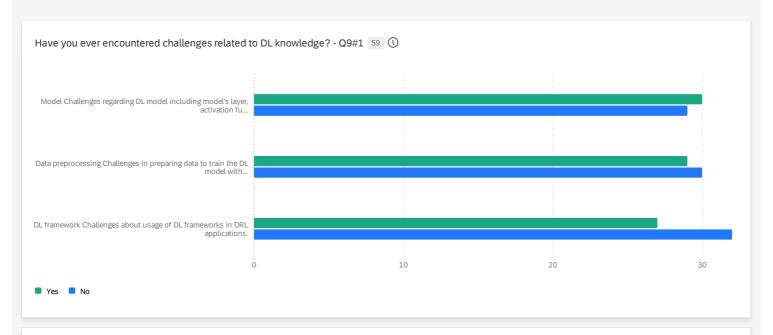
| ave you ever encountered challenges related to parallel processing 8 | a multi-threading? - How severe | was the challenge/issue 59 (i) | |
|--|---------------------------------|--------------------------------|----------|
| How severe was the challenge/issue | Minor | Major | Critical |
| GPU usage Challenges regarding utilizing GPU for running DRL applications | 41 | 13 | 5 |
| Multi-threading Challenges about running DRL applications as a multi-threa | 40 | 15 | 4 |
| Multi-processing Challenges in running DRL applications in a multi-process | 33 | 19 | 7 |
| Distributed systems Challenges in running DRL applications as a distribute | 35 | 16 | 8 |

| How severe was the challenge/issue | Average | Minimum | Maximum | Cou |
|--|---------|---------|---|-----|
| | , | | TOWN THE STATE OF | |
| GPU usage Challenges regarding utilizing GPU or running DRL applications | 1.39 | 1.00 | 3.00 | ! |
| Multi-threading Challenges about running DRL applications as a multi-threa | 1.39 | 1.00 | 3.00 | ! |
| Multi-processing Challenges in running DRL applications in a multi-process | 1.56 | 1.00 | 3.00 | į |
| Distributed systems Challenges in running DRL applications as a distribute | 1.54 | 1.00 | 3.00 | ! |



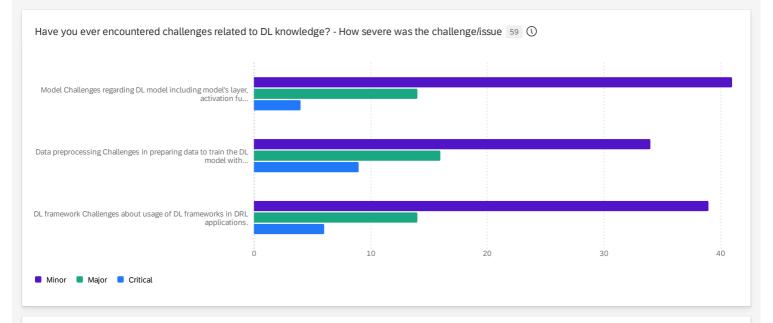
| Have you ever encountered challenges related to parallel processing & m $_{59}$ $\textcircled{1}$ | nulti-threading? - How much e | ffort was required to handle this challenge/ | issue |
|---|-------------------------------|--|-------|
| How much effort was required to handle this challenge/issue | Low | Medium | High |
| GPU usage Challenges regarding utilizing GPU for running DRL applications | 40 | 12 | 7 |
| eq:Multi-threading Challenges about running DRL applications as a multi-threa | 39 | 11 | 9 |
| Multi-processing Challenges in running DRL applications in a multi-process | 30 | 14 | 15 |
| Distributed systems Challenges in running DRL applications as a distribute | 34 | 8 | 17 |

| Have you ever encountered challenges related to pa | rallel processing & multi-threa | ading? - How much effort was r | equired to handle this challen | ge/issue |
|--|---------------------------------|--------------------------------|--------------------------------|----------|
| How much effort was required to handle this challenge/issue | Average | Minimum | Maximum | Count |
| GPU usage Challenges regarding utilizing GPU for running DRL applications | 1.44 | 1.00 | 3.00 | 59 |
| Multi-threading Challenges about running DRL applications as a multi-threa | 1.49 | 1.00 | 3.00 | 59 |
| Multi-processing Challenges in running DRL applications in a multi-process | 1.75 | 1.00 | 3.00 | 59 |
| Distributed systems Challenges in running DRL applications as a distribute | 1.71 | 1.00 | 3.00 | 59 |



| Yes | No |
|-----|-------|
| 30 | 29 |
| 29 | 30 |
| 27 | 32 |
| | 30 29 |

| ive you ever encountered challenges related to I | | | | |
|--|---------|---------|---------|------|
| 9#1 | Average | Minimum | Maximum | Cour |
| Model Challenges regarding DL model ncluding model's layer, activation fu | 1.49 | 1.00 | 2.00 | Ę |
| Data preprocessing Challenges in preparing data to train the DL model with | 1.51 | 1.00 | 2.00 | Ę |
| DL framework Challenges about usage of DL frameworks in DRL applications. | 1.54 | 1.00 | 2.00 | 5 |



| Have you ever encountered challenges related to DL knowledge? - $\mbox{\sc How}$ | severe was the challenge/issu | ie 59 🛈 | |
|--|-------------------------------|---------|----------|
| How severe was the challenge/issue | Minor | Major | Critical |
| Model Challenges regarding DL model including model's layer, activation fu | 41 | 14 | 4 |
| Data preprocessing Challenges in preparing data to train the DL model with | 34 | 16 | 9 |
| DL framework Challenges about usage of DL frameworks in DRL applications. | 39 | 14 | 6 |

| ave you ever encountered challenges related to D | L knowledge? - How severe wa | as the challenge/issue 59 (i) | | |
|--|------------------------------|-------------------------------|---------|-------|
| How severe was the challenge/issue | Average | Minimum | Maximum | Count |
| Model Challenges regarding DL model including model's layer, activation fu | 1.37 | 1.00 | 3.00 | 59 |
| Data preprocessing Challenges in preparing data to train the DL model with | 1.58 | 1.00 | 3.00 | 59 |
| DL framework Challenges about usage of DL frameworks in DRL applications. | 1.44 | 1.00 | 3.00 | 59 |



| ave you ever encountered challenges related to DL knowledge? - Hov | w much effort was required to han | dle this challenge/issue 59 (i) | |
|--|-----------------------------------|---------------------------------|------|
| How much effort was required to handle this challenge/issue | Low | Medium | High |
| Model Challenges regarding DL model including model's layer, activation fu | 38 | 16 | į |
| Data preprocessing Challenges in preparing data to train the DL model with | 38 | 14 | : |
| DL framework Challenges about usage of DL frameworks in DRL applications. | 38 | 16 | į |

| , | J | ort was required to handle this | 3 | |
|--|---------|---------------------------------|---------|------|
| How much effort was required to handle this challenge lissue | Average | Minimum | Maximum | Coun |
| Model Challenges regarding DL model including model's layer, activation fu | 1.44 | 1.00 | 3.00 | 59 |
| Data preprocessing Challenges in preparing data to train the DL model with | 1.47 | 1.00 | 3.00 | 5 |
| DL framework Challenges about usage of DL frameworks in DRL applications. | 1.44 | 1.00 | 3.00 | 5 |

| Have you ever encountered any challenges/issues related to DRL that have not been mentioned in this survey? If yes, could you please describe them? |
|---|
| Train in simulation, and deploy in the real world enviroment |
| Another very important challenge is to adapt an existing environment/game for a DRL framework: if the environment is not created with the intention of using it for DRL, adapting it to this framework (e.g. make the env asking for a request, wait until the net has a response, batch observations in case of multi-agent env, etc) is a very critical challenge, that requires a very high level of effort and knowledge to solve. |
| Reward function drift, unstable learning due to shifts in data distributions, and more costly (time and expenses) to train a DRL agent compared to supervised learning. |
| Maybe is mentioned, but I want to add the problem of when is better to use off policy vs on policy algorithms |
| Industrial application, tendency |
| Redesigning problems to fit the (PO)MDP formulation can often be a struggle when working with folks that are fairly rigid about how they view the problem. This is mostly only an issue with critical controls tasks. |
| I work with highly random data that is hard for agents to learn. Finding features is difficult |
| The challenge related to Mujoco. Since DeepMind bought Mujoco and made it open source, there has been major challenges that requires high effort to design new projects with the open source Mujoco, especially when you start a new project and want to use some old codes (or old repositories) using older version of Mujoco. |
| Yes, Training - To realize when to stop training to efficient result |
| No, the survey covers all DRL topics well. |
| Local optimization methods, such as stochastic gradient descent, find a local optimum. Thus, when optimising multiple times, one can obtain different solutions. Analyzing the (quality) of solutions has been a challenge. |
| Practical applications such as robots or speficic device for control are important problem for RL, cause the only RL exist in book or simulation, it cannot attract more researchers and funding for RL. So I think sim-to-real, or pre-trained model learned by RL is important for future work. |
| Default parameters used in common libraries are just wrong so comparing evaluations by default results in inaccuracies that authors don't report, OR authors paper conflicts with his code and now there's 2 "valid" implementations floating around of this model, the one they released, and the one they discussed in the paper, and they can sometimes be hugely different in terms of implications, even though they both achieve similar results usually. |
| - Lack of active exploration capabilities in most DRL algorithms, beside random exploration which is not enough for some domains. (thus this does not fit the existing "Hyperparameters Tuning" category even though it mentions exploration, since no amount of tuning the epsilon decay or gaussian noise would fix the issue) - sim 2 real gap for applied DRLI guess this kind of falls into the reward/environment categories, but sometimes the world is too complex/slow to be efficiently AND accurately approximated anyway, so robustness/transfer is required - difficulty in interpreting the logs and diagnosing which part of all the RL machinery is broken (reward training? value training? policy training? representation? exploration? something else?) |
| no other challenges/issues encountered |
| How to optimize models to run in GPUs, and when this is best. We have found that sometimes CPUs run RL code faster than GPU, and we were uncertain as to why this was happening. |
| Model interpretability may require domain expertise. |
| Best wishes for your research |
| rllib is the worst RL library for research |
| |

| I often answered with "NO" due to the lack of experience in DRL. I only did a few learning projects with it and took a few month part of at a project at work. Most problems where |
|--|
| Hyperparameter-Tuning tried to solve it automatically with Ray-Tune or other Tools. Why do I write this comment? I hope you can better classify my answers with this background |
| information |

the design of the state space took me a lot of time, when testing new methods and mixing them with some different methods

Thank you for contacting me to survey the DRL. I am learning from many excellent studies by many DRL researchers.

Authors of RL papers deliberately make their math obtusely confusing when the entire mathematical framework for RL has been generalized for decades, this makes papers harder to follow but they think it makes them sound smart. other comments: why would you make the survey like this lol it's so unnecessarily confusing, if I answer no I still have given answers to the other questions? this isn't that hard come on now.

I focus on application of DRL algorithms and biggest usage of time is figuring out the "optimal" way to create a customized environment, how to represent the state-action space, and how to describe non-sparse rewards such that the agent learns a behaviour that is aligned with the objective.

While there are many tutorials to start working on RL, a few issues are shared between many. 1. Diversity- researchers use RL in different domain. However, many tutorials cover the RL for a very specific domain/application. For example, many tutorials use Cart Pole as their example. But, different researchers might like to see more serious examples related to their field. 2. Hands-on- Broadly speaking, machine learning and deep learning tutorials are distinguished between two themes: 1. Research-based and 2. development-based. Tutorials from category one cover a wide range of background that is needed to learn ML/DeepL in depth. The second category focuses on teaching how to use the existing techniques without having to learn the mathematical background. Doing so, they help their audience spend their time only as much as required for their purpose. However, many RL tutorials combine some background knowledge with hands-on, which often leads to these two undesirable situations: 1- they cover none of the background and hands-on in depth and with variety 2- audiences have to spend a lot of time to read things that may not be necessary for their purpose

RL is hard! :)

I didn't use parallel processing & multi-threading in my DRL research...

As mentioned in the dependency section, the most frustrating part of working with RL currently is all the breaking API changes. Especially with gym(nasium) consistently changing their API it effects all the downstream libraries.