

—— “2007-2017” ECUG 十年高峰盛会 ——

AI算法实现和云平台应用

2017.12.17

About Me

- ❖ HBase/Hadoop contributor
- ❖ OpenStack contributor
- ❖ Seagull(Docker Web UI) author
- ❖ TensorFlow/Kubernetes developer
- ❖ System architect at 4Paradigm



Agenda

- ◆ 人工智能与机器学习介绍
- ◆ 机器学习算法原理与实现
- ◆ 云机器学习平台架构实践

Agenda

- ◆ 人工智能与机器学习介绍
- ◆ 机器学习算法原理与实现
- ◆ 云机器学习平台架构实践

人工智能？

- ◆ 机械自动化 -> 人工智能自动工场
 - ◆ 字符串生成 -> 人工智能自动编程
 - ◆ 验证码识别 -> 人工智能黑客技术

机器学习介绍

- ✿ "A computer program is said to learn from **experience E** with respect to some class of **tasks T** and **performance measure P** if its performance at tasks in T, as measured by P, improves with experience E." - Mitchell
- ✿ 场景 + 指标 + 样本 -> 机器学习模型

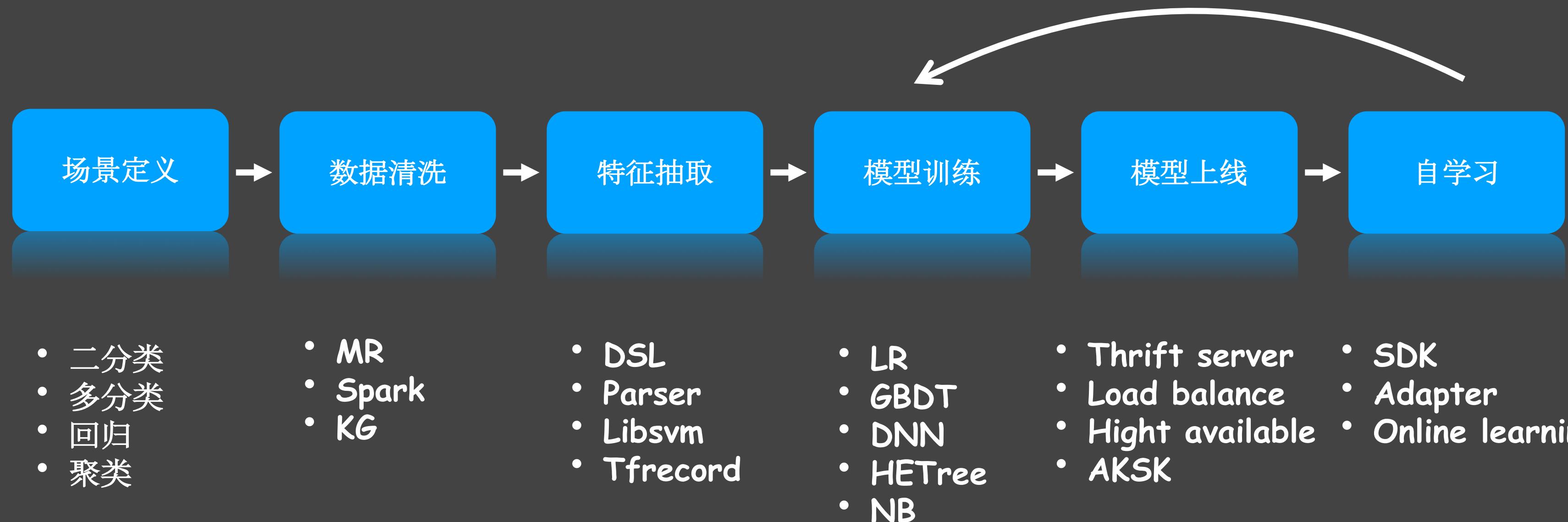
机器学习应用

- ♣ 1000张猫的图片，得到一个识别猫的模型
- ♣ 1000万盘围棋棋局，得到一个AlphaGo模型
- ♣ 历史的信用卡操作记录，得到一个反欺诈模型
- ♣

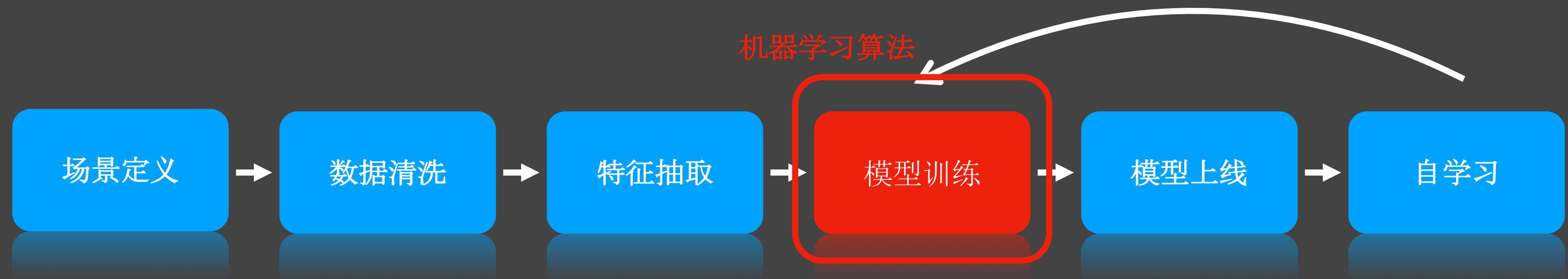
机器学习应用

- ♣ 1000张猫的图片，**得到一个识别猫的模型**
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- ♣ 所有的信用卡操作记录，**得到一个反欺诈模型**
- ♣

得到机器学习应用



得到机器学习应用



- 二分类
- 多分类
- 回归
- 聚类

- MR
- Spark
- KG

- DSL
- Parser
- Libsvm
- Tfrecord

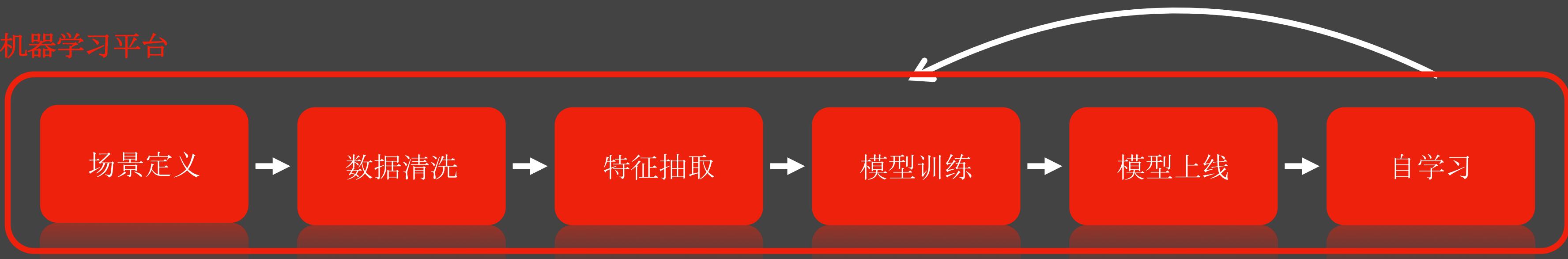
- LR
- GBDT
- DNN
- HETree
- NB

- Thrift server
- Load balance
- Hight available
- AKSK

- SDK
- Adapter
- Online learning

得到机器学习应用

机器学习平台



- 二分类
- 多分类
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- MR
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机器学习算法 - 逻辑回归

逻辑回归：简单而强大的机器学习算法，广泛用于推荐系统、CTR等场景

- ◆ 监督型模型
- ◆ 线性模型
- ◆ 高性能
- ◆ 可解析性强
- ◆ 可拓展性强

机器学习算法 - 逻辑回归

性别:男 性别:女 年龄:0-30 年龄:30-60 年龄:60-90 年龄:80+ 收入>50

样本1	1	0	0	1	0	0	1
样本2	0	1	0	0	0	1	0
...
LR模型	0.7	0.4	0.6	0.9	0.3	0.01	

机器学习算法 - 逻辑回归

性别:男 性别:女 年龄:0-30 年龄:30-60 年龄:60-90 年龄:80+ 收入>50

LR模型	0.7	0.4	0.6	0.9	0.3	0.01
-------------	------------	------------	------------	------------	------------	-------------

新样本	1	0	1	0	0	0
------------	---	---	---	---	---	---

计算得分	0.7	0	0.6	0	0	0
-------------	-----	---	-----	---	---	---

$$\text{activation}(0.7 + 0.6) = 1$$

机器学习算法 - 逻辑回归

```

import tensorflow as tf
import numpy as np
from tensorflow.examples.tutorials.mnist import input_data

# Train data
mnist = input_data.read_data_sets("./mnist/", one_hot=True)

# Hyper parameters
learning_rate = 0.1
batch_size = 10
train_epoch_number = 100
display_interval = 10

# The variables to compute
x = tf.placeholder(tf.float32, [None, 784])
y = tf.placeholder(tf.float32, [None, 10])
W = tf.Variable(tf.zeros([784, 10]), name="weight")
b = tf.Variable(tf.zeros([10]), name="bias")

pred = tf.nn.softmax(tf.matmul(x, W) + b)
loss = tf.reduce_mean(-tf.reduce_sum(y * tf.log(pred), reduction_indices=1))
train_op = tf.train.GradientDescentOptimizer(learning_rate).minimize(loss)

init_op = tf.initialize_all_variables()

tf.global_variables_initializer()

```

机器学习算法 - 逻辑回归

逻辑回归：简单而强大的机器学习算法，广泛用于推荐系统、CTR等场景

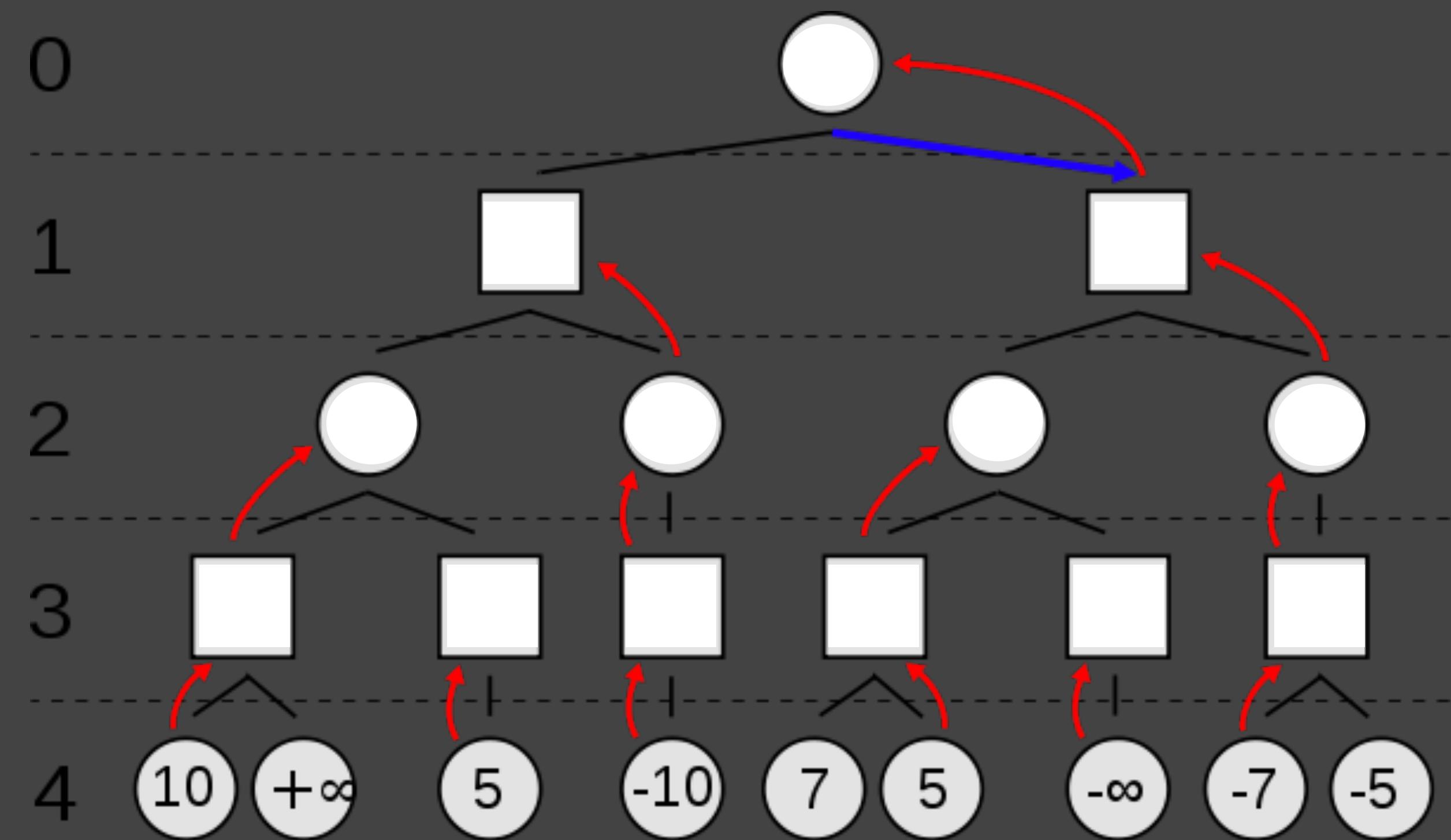
- ♣ 使用特征工程生成非线性Feature
- ♣ 模型为一维数组（长度与Feature维度相同）
- ♣ 定义Loss函数作为指标（如CrossEntropy）
- ♣ 使用梯度下降算法训练（如Adagrad优化器）
- ♣ 通过Parameter server实现分布式训练

机器学习算法 - AlphaZero

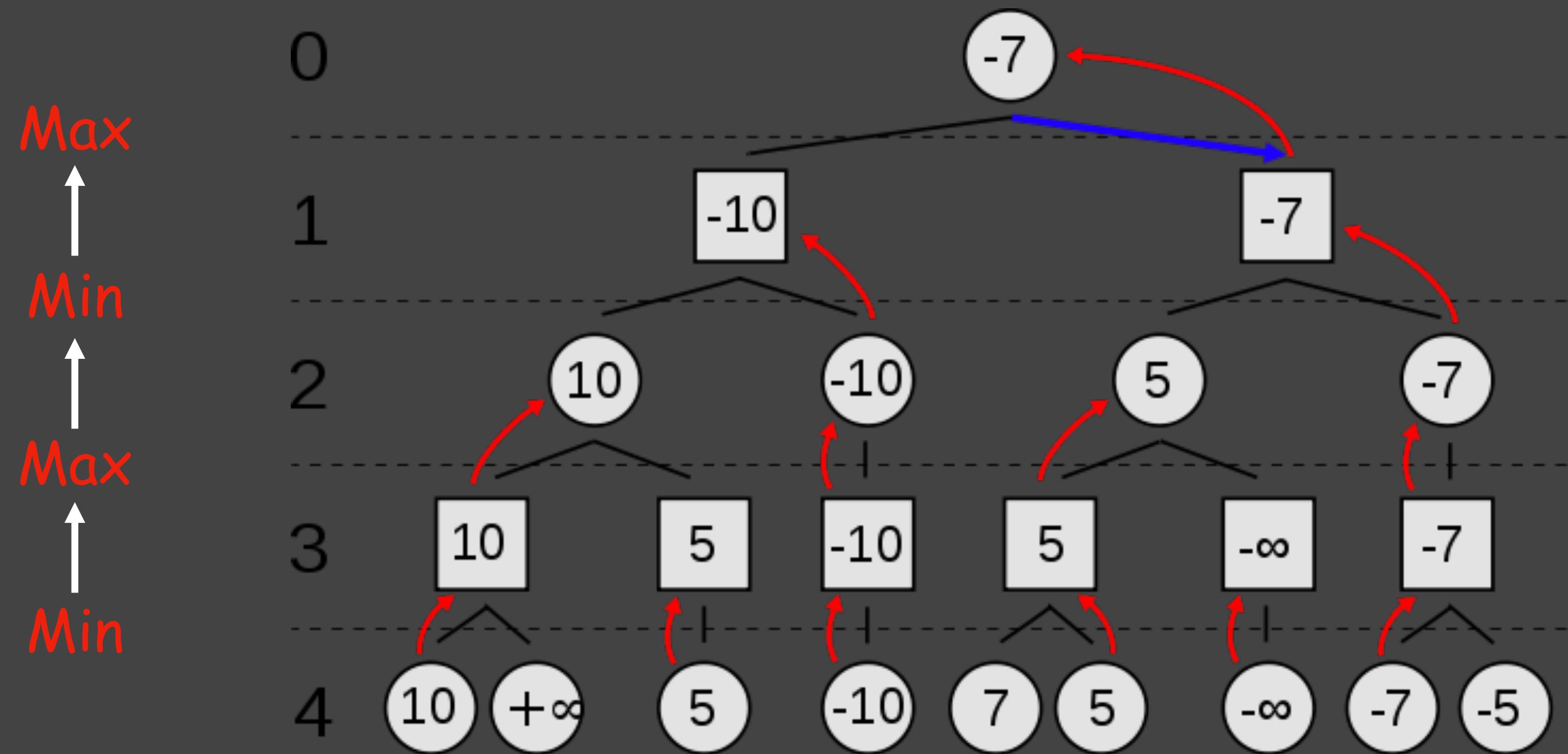
AlphaZero: 基于蒙特卡罗树搜索和神经网络的机器学习模型

- ◆ 基于蒙特卡罗树结构定义围棋规则（非自学习理解）
- ◆ 有监督学习，学习蒙特卡罗树搜索结果
- ◆ 使用Policy gradient（增强学习）生成更优的样本数据
- ◆ 模型为ResNet（输入Feature输出Probability）
- ◆ 适用于Combination game（零和、完全信息、无随机）

机器学习算法 - MinMax

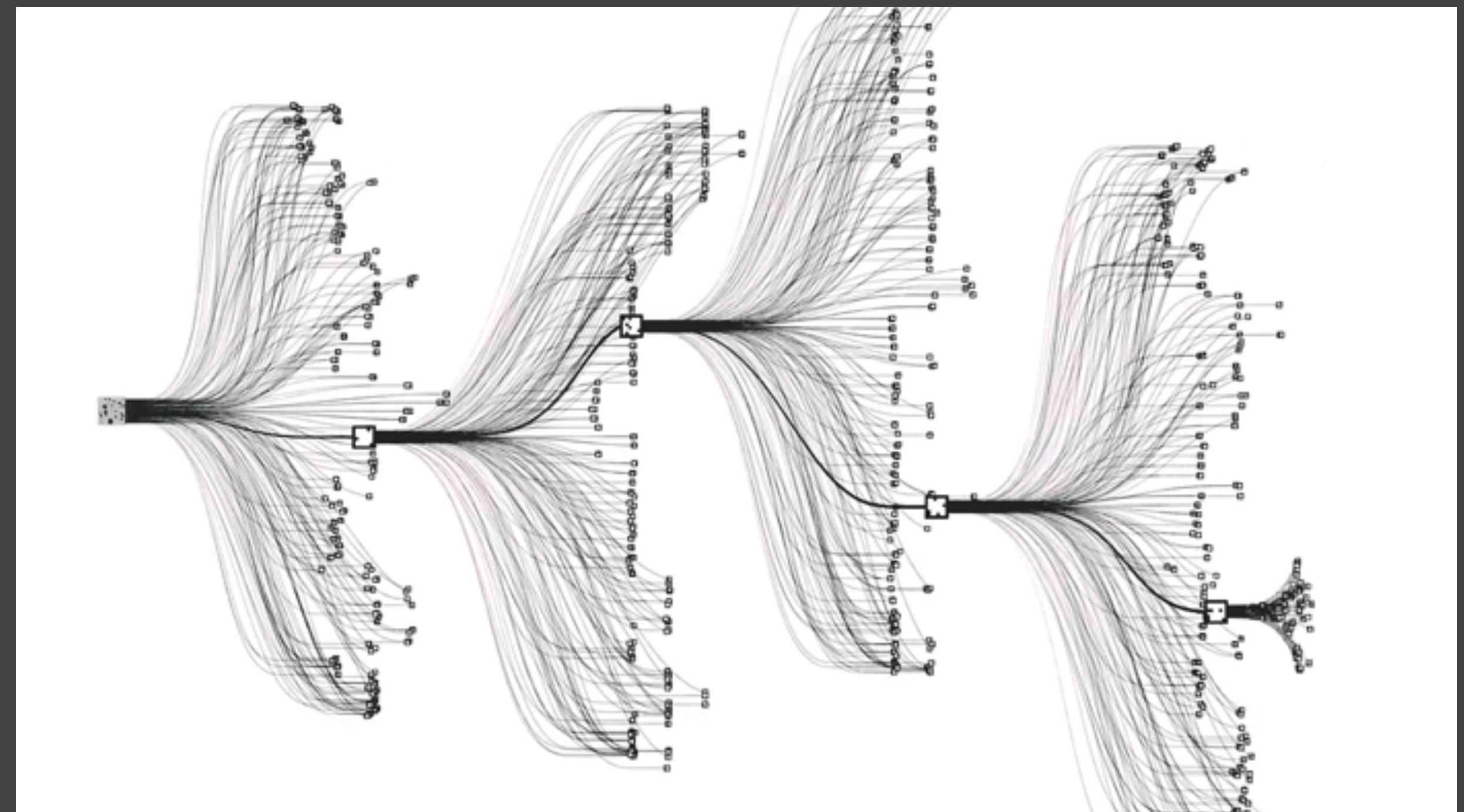


机器学习算法 - MinMax

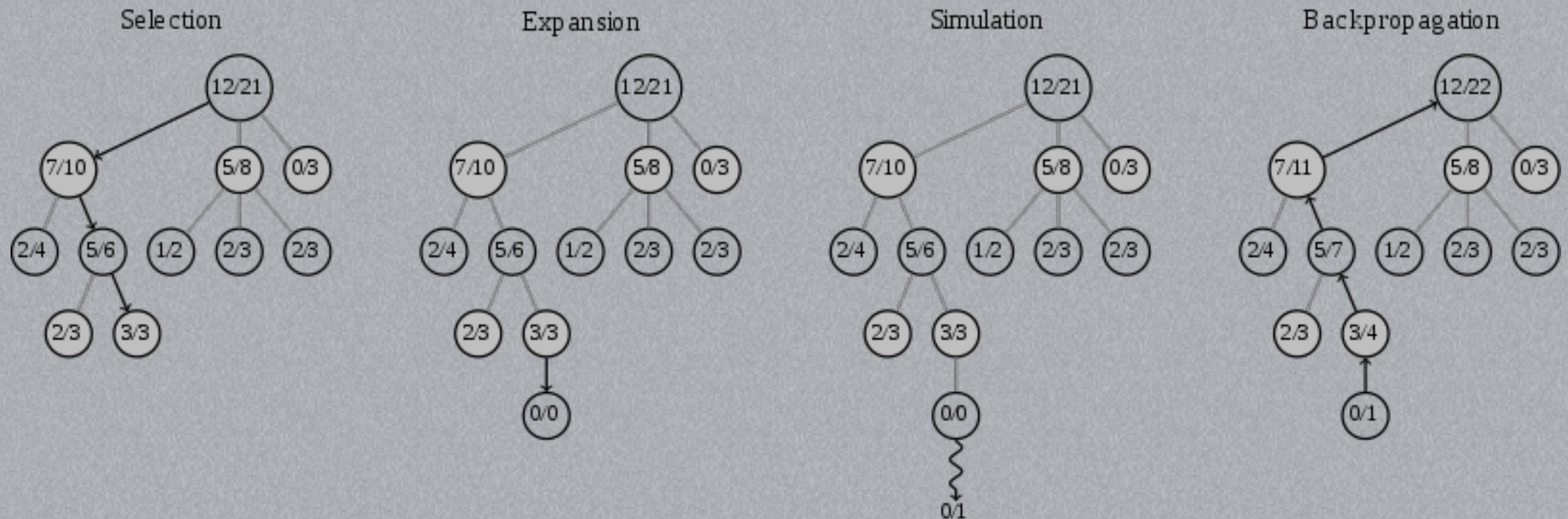


机器学习算法 - MCTS

- ❖ Exploration
- ❖ Exploitation

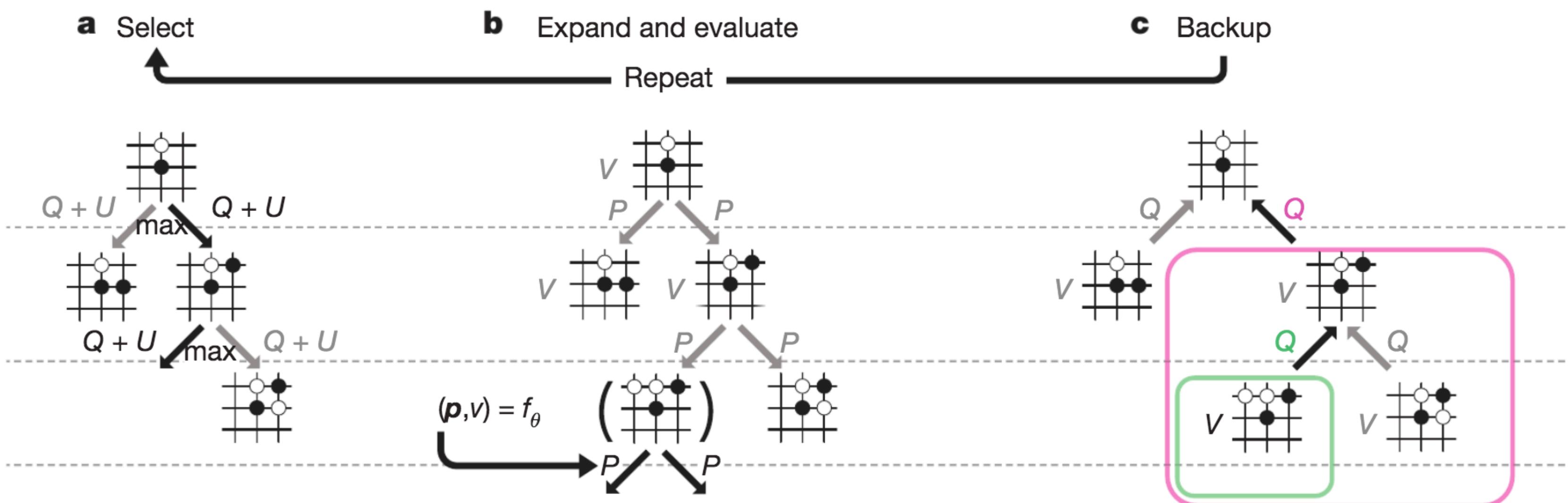


机器学习算法 - MCTS



机器学习算法 -

AlphaZero



$$(p, v) = f_\theta(s) \text{ and } l = (z - v)^2 - \pi^T \log p + c \|\theta\|^2$$

机器学习算法 - AlphaZero

- ✿ “UCB value”: $Q + U$
- ✿ Q : $\text{Sum_of_} V(\text{children}) / N$
- ✿ V : quality value, from neural network
- ✿ N : visit times
- ✿ U : $C * P * \text{"vanila UCB"}$
- ✿ C : Constant parameter to control exploration/exploitation
- ✿ P : probability to play, from neural network

机器学习算法 - AlphaZero

- ✿ “UCB value”: $Q + U$
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机器学习算法 - 自动求导

- ♣ Op求导 (Add/Minus/Multiple/Divide/Square...)
- ♣ 链式法则
- ♣ 求偏导
- ♣ 梯度下降 (Learning rate)
- ♣ Python operator override

机器学习算法 - 自动求导

on GitHub
<https://github.com/tobegin3hub/miniflow>

```
class AddOp(Op):
    """
    The addition operation which has only two inputs. The input can be
    primitive, ConstantOp, PlaceholderOp, VariableOp or other ops.
    """

    def __init__(self, input1, input2, name="Add"):
        super(AddOp, self).__init__(name)

        if not isinstance(input1, Op):
            self._op1 = ConstantOp(input1)
        else:
            self._op1 = input1

        if not isinstance(input2, Op):
            self._op2 = ConstantOp(input2)
        else:
            self._op2 = input2

        self._graph = graph.get_default_graph()
        self._graph.add_to_graph(self)

    def forward(self):
        result = self._op1.forward() + self._op2.forward()
        return result

    def grad(self, partial_derivative_opname=None):
        result = self._op1.grad(partial_derivative_opname) + self._op2.grad(
            partial_derivative_opname)
        return result

    def __str__(self):
        result = self._op1.__str__() + self._op2.__str__()
        return result
```

```
class PowerOp(Op):
    def __init__(self, input, power, name="Power"):
        super(PowerOp, self).__init__(name)

        if not isinstance(input, Op):
            self._op = ConstantOp(input)
        else:
            self._op = input

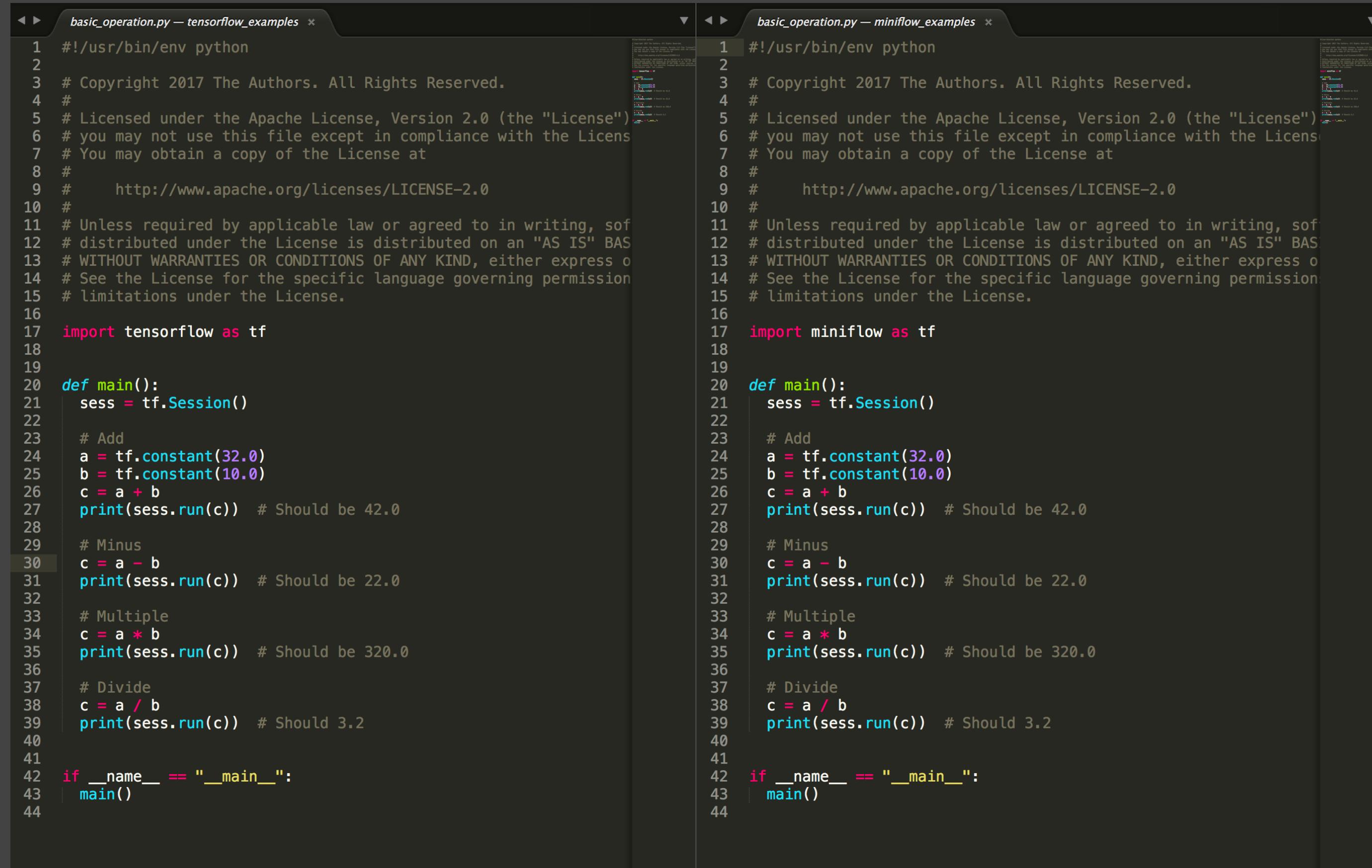
        self._power = power
        self._graph = graph.get_default_graph()
        self._graph.add_to_graph(self)

    def forward(self):
        result = pow(self._op.forward(), self._power)
        return result

    def grad(self, partial_derivative_opname=None):
        if isinstance(self._op, PlaceholderOp) or isinstance(self._op, ConstantOp):
            # op is the constant
            grad = 0
        elif isinstance(self._op, VariableOp):
            # op is the variable
            grad = self._power * pow(self._op.forward(), self._power - 1)
        else:
            # op is other complex operation and use chain rule
            grad = self._power * pow(self._op.forward(), self._power - 1) * self._op.grad(partial_derivative_opname)

        return grad
```

机器学习算法 - 自动求导



```
basic_operation.py -- tensorflow_examples *
1 #!/usr/bin/env python
2
3 # Copyright 2017 The Authors. All Rights Reserved.
4 #
5 # Licensed under the Apache License, Version 2.0 (the "License")
6 # you may not use this file except in compliance with the License.
7 # You may obtain a copy of the License at
8 #
9 #     http://www.apache.org/licenses/LICENSE-2.0
10 #
11 # Unless required by applicable law or agreed to in writing, software
12 # distributed under the License is distributed on an "AS IS" BASIS,
13 # WITHOUT WARRANTIES OR CONDITIONS OF ANY KIND, either express or
14 # implied. See the License for the specific language governing
15 # permissions and limitations under the License.
16
17 import tensorflow as tf
18
19
20 def main():
21     sess = tf.Session()
22
23     # Add
24     a = tf.constant(32.0)
25     b = tf.constant(10.0)
26     c = a + b
27     print(sess.run(c)) # Should be 42.0
28
29     # Minus
30     c = a - b
31     print(sess.run(c)) # Should be 22.0
32
33     # Multiple
34     c = a * b
35     print(sess.run(c)) # Should be 320.0
36
37     # Divide
38     c = a / b
39     print(sess.run(c)) # Should be 3.2
40
41
42 if __name__ == "__main__":
43     main()

basic_operation.py -- miniflow_examples *
1 #!/usr/bin/env python
2
3 # Copyright 2017 The Authors. All Rights Reserved.
4 #
5 # Licensed under the Apache License, Version 2.0 (the "License")
6 # you may not use this file except in compliance with the License.
7 # You may obtain a copy of the License at
8 #
9 #     http://www.apache.org/licenses/LICENSE-2.0
10 #
11 # Unless required by applicable law or agreed to in writing, software
12 # distributed under the License is distributed on an "AS IS" BASIS,
13 # WITHOUT WARRANTIES OR CONDITIONS OF ANY KIND, either express or
14 # implied. See the License for the specific language governing
15 # permissions and limitations under the License.
16
17 import miniflow as tf
18
19
20 def main():
21     sess = tf.Session()
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```

机器学习算法 - 自动求导

on GitHub
<https://github.com/tobegit3hub/miniflow>

- ✿ TensorFlow vs MiniFlow(22x speedup)
- ✿ 12.59 > 0.16 | 12.61 > 0.16 | 66.58 > 3.01

```
→ add_operation git:(master) ✘ ./tensorflow_add.py
Benchmark scenario: add operation, epoch: 100000
Run time(s): 12.5948340893
→ add_operation git:(master) ✘ cd ../multiple_operation
miniflow_multiple.py    tensorflow_multiple.py
→ multiple_operation git:(master) ✘ ./tensorflow_multiple.py
Benchmark scenario: multiple operation, epoch: 100000
Run time(s): 12.6196570396
→ multiple_operation git:(master) ✘ cd ../linear_regression
miniflow_linear_regression.py    tensorflow_linear_regression.py
→ linear_regression git:(master) ✘ ./tensorflow_linear_regression.py
Benchmark scenario: linear regression, epoch: 100000
Run time(s): 66.5865750313
```

```
| → add_operation git:(master) ✘ ./miniflow_add.py
| Benchmark scenario: add operation, epoch: 100000
| Run time(s): 0.163687944412
| → add_operation git:(master) ✘ cd ../multiple_operation
| miniflow_multiple.py    tensorflow_multiple.py
| → multiple_operation git:(master) ✘ ./miniflow_multiple.py
| Benchmark scenario: multiple operation, epoch: 100000
| Run time(s): 0.165066957474
| → multiple_operation git:(master) ✘ cd ../linear_regression
| miniflow_linear_regression.py    tensorflow_linear_regression.py
| → linear_regression git:(master) ✘ ./miniflow_linear_regression.py
| Benchmark scenario: linear regression, epoch: 100000
| Run time(s): 3.01005196571
```

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- ♣ 云机器学习平台架构实践

云机器学习平台 - 应用流程



云机器学习平台 - 架构设计

- ♣ TensorFlow LR支持10亿维稀疏模型，10万亿维呢？
- ♣ 开源框架与自研框架（机器学习算法）的集成
- ♣ 异构计算集群（CPU、GPU、虚拟机、云平台）的支持
- ♣ 机器学习工作流的支持

云机器学习平台 - 架构设计

IaaS

- ❖ AWS VMs
- ❖ AliCloud VMs
- ❖ AliCloud VMs

PaaS

- ❖ Google CloudML
- ❖ Xiaomi CloudML
- ❖ AliCloud PAI

“MLaaS”

- ❖ Azure ML Studio
- ❖ 4Paradigm prophet

图像处理

自然语言处理

语音处理

推荐系统

先知机器学习平台

API / SDK / WEB / Command-line

模型训练

模型服务

数据管理

特征抽取

Distributed
TensorFlow

Distributed MXNet

模型评估

自学习服务

Kubernetes Executor

Yarn / Hadoop Executor

CPU集群

GPU集群

FPGA集群

公有云

私有云

其他服务器

云机器学习平台 - “Google CloudML”

```

FROM ubuntu:16.04
RUN apt-get update && apt-get install -y --no-install-recommends \
    build-essential \
    curl \
    libfreetype6-dev \
    libpng12-dev \
    libzmq3-dev \
    pkg-config \
    python \
    python-dev \
    rsync \
    software-properties-common \
    unzip \
    && \
    apt-get clean && \
    rm -rf /var/lib/apt/lists/*
RUN pip --no-cache-dir install \
    http://storage.googleapis.com/tensorflow/linux/cpu/tensorflow
COPY jupyter_notebook_config.py /root/.jupyter/
COPY notebooks /notebooks
COPY run_jupyter.sh /
EXPOSE 6006
EXPOSE 8888
WORKDIR "/notebooks"
CMD ["/run_jupyter.sh", "--allow-root"]
CWD ["/root"]
WORKDIR "/notebooks"

```

Step 1: Build docker image

```

EXPOSE 8888
EXPOSE 6006
COPY run_jupyter.sh \
COPY notebooks /notebooks \
COPY run_jupyter.sh /root/run_jupyter \

```

```

@csrf_exempt
def train(request):
    if request.method == 'POST':
        # Get request data
        body = json.loads(request.body)
        job_name = body.get('job_name')
        job_module_name = body.get('module_name')
        job_trainer_uri = body.get('trainer_uri')
        job_args = body.get('job_args')
        job_train_data_paths = body.get('trainer_data_paths')
        job_eval_data_paths = body.get('job_eval_data_paths')
        job_output_path = body.get('output_path')
        job_master_spec = body.get('master_spec')

        # Generate Kuberentes resources
        time_suffix = datetime.datetime.now().strftime("%m%d%H%M")
        kubernetes_resource_name = job_name + "-" + time_suffix
        train_container_run_commands = ["/run_user_module.py",
                                         job_trainer_uri, job_module_name]

        client = KubernetesClient().get_client()

```

```

client = KubernetesClient().get_client()
# Create a deployment object
train_container = Container(name=kubernetes_resource_name,
                            image=job_trainer_uri,
                            command=train_container_run_commands,
                            ports=[{"port": 6006}], ...
deployment = Deployment(name=kubernetes_resource_name,
                        replicas=1,
                        template=train_container,
                        spec={"containers": [train_container]})

# Create a service object
service = Service(name=kubernetes_resource_name,
                  selector=kubernetes_resource_name,
                  ports=[{"port": 8888}], ...

```

```

deployment_object = {
    "apiVersion": "extensions/v1beta1",
    "kind": "Deployment",
    "spec": {
        "replicas": 1,
        "template": {
            "spec": {
                "containers": [
                    {
                        "name": kubernetes_resource_name,
                        "image": docker_image,
                        "command": container_run_commands,
                        "resources": {
                            "requests": {
                                "cpu": limit_cpu,
                                "memory": limit_memory,
                            }
                        }
                    }
                ]
            }
        }
    }
}

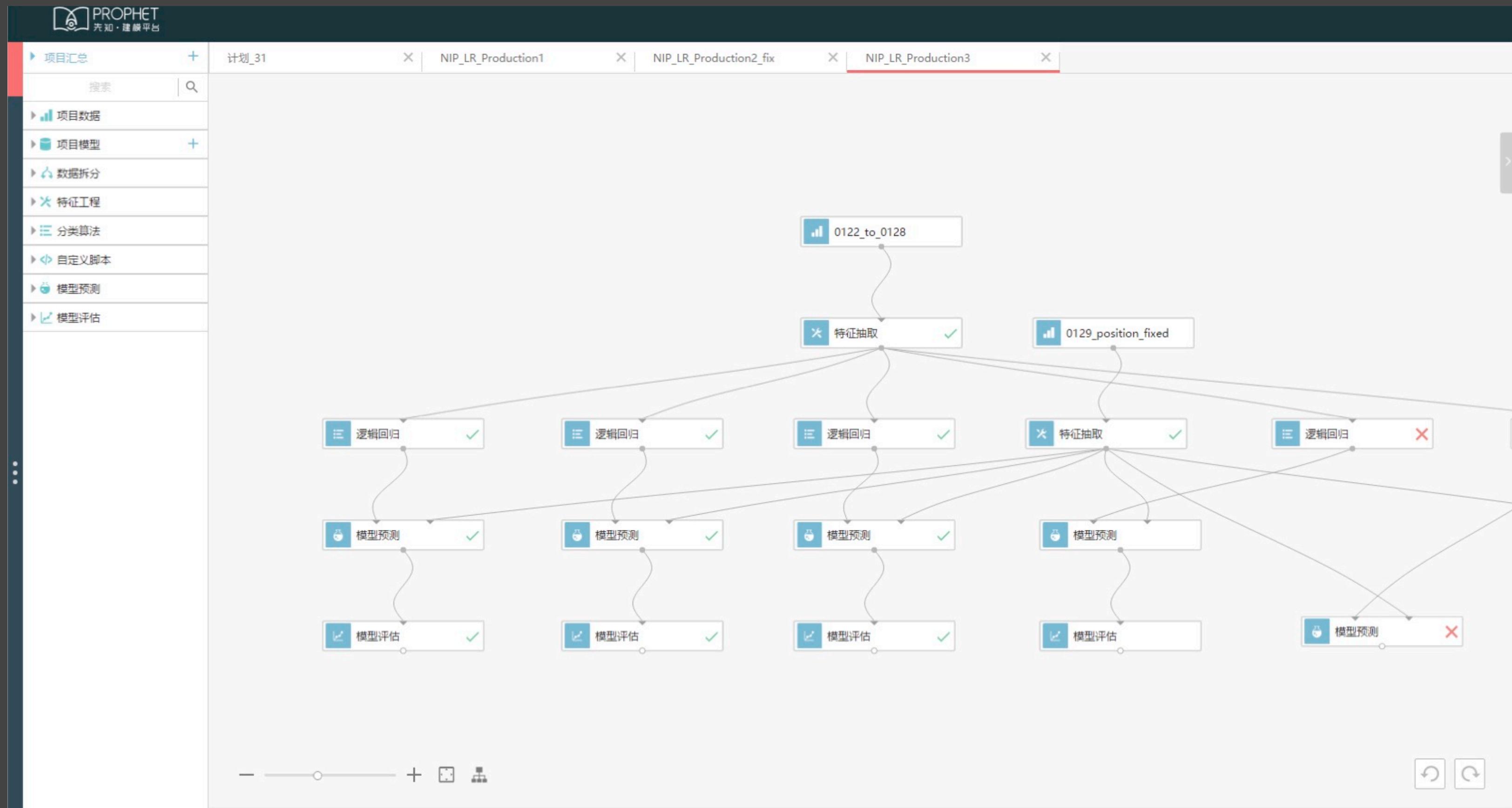
```

Step 3: Submit to Kubernetes

云机器学习平台 - 工作流

- ♣ 定义数据引入 -> 特征抽取 -> 模型训练 -> 模型评估工作流
- ♣ 拖拽式 (Drag-and-drop) 操作方式
- ♣ 自定义功能算子

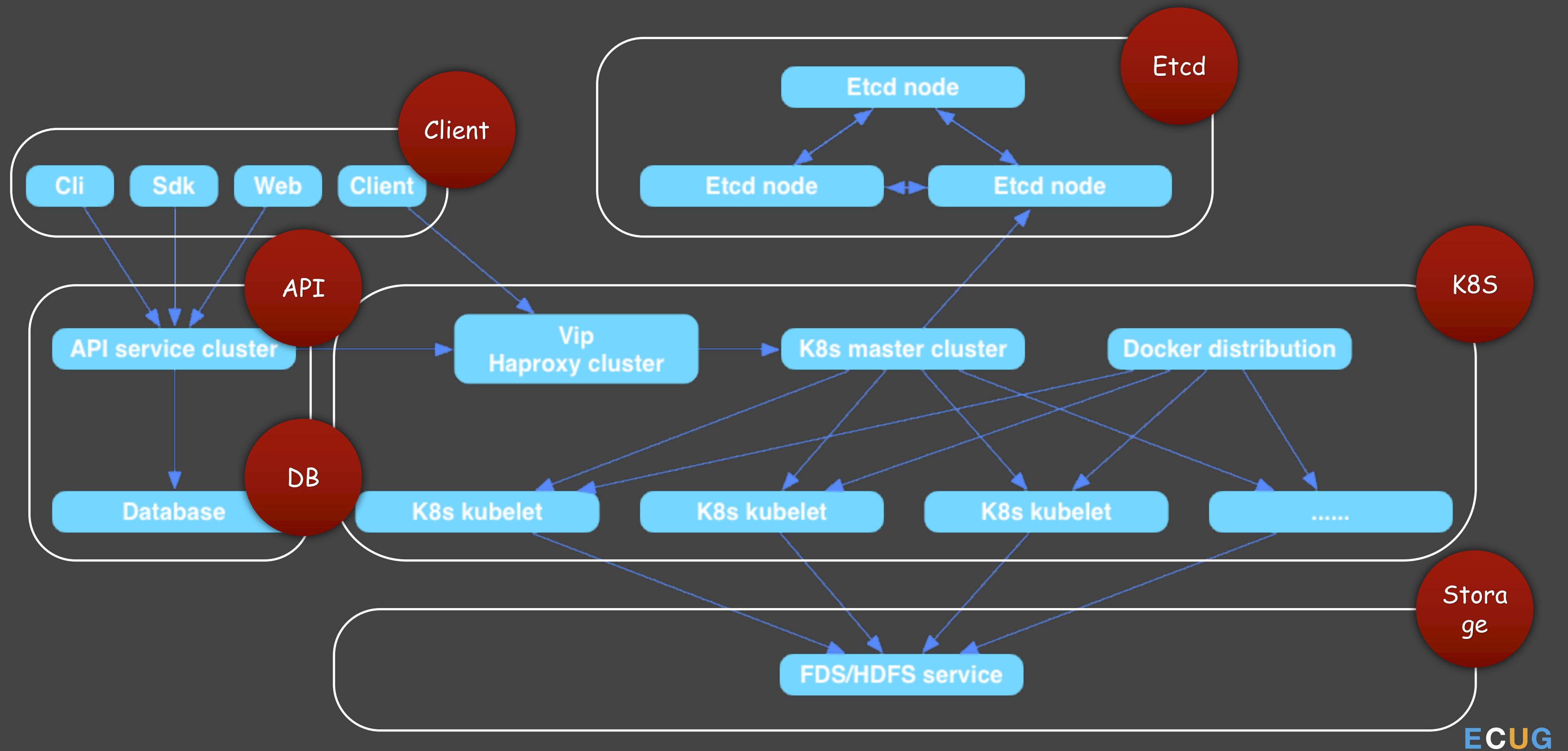
云机器学习平台 - 工作流



云机器学习平台 - 算子

- ◆ 最简抽象接口 : `abstract void execute()`
- ◆ ~10 数据处理算子, 10+ 机器学习训练算子, 3 AutoML算子
- ◆ Hadoop/Spark/Kubernetes多集群支持
- ◆ More on <https://www.4paradigm.com/support/help>

云机器学习平台 - 高可用



云机器学习平台 - 多租户

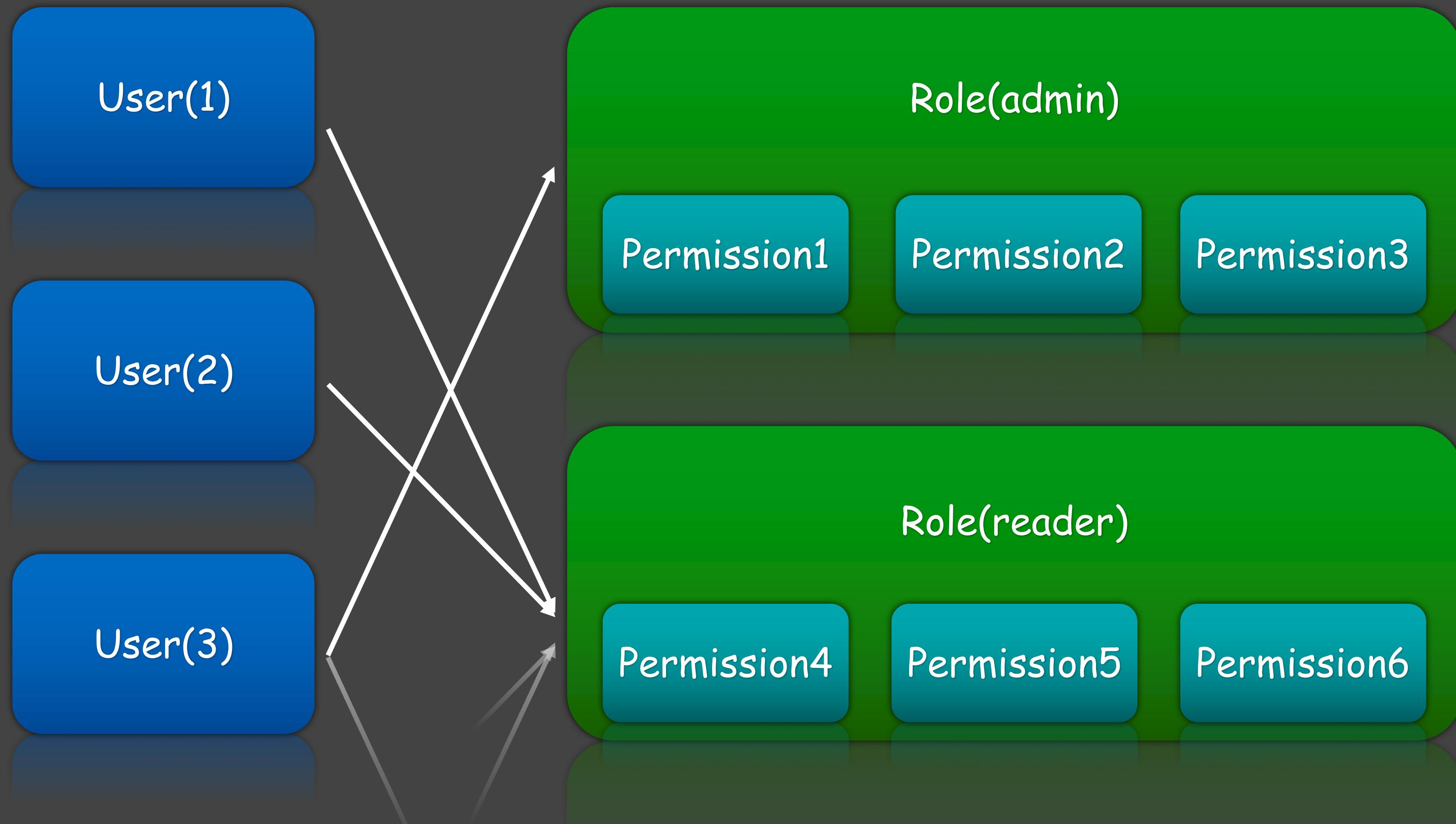
❖ Authentication

- Username / Password
- Access key / Secret key
- LDAP

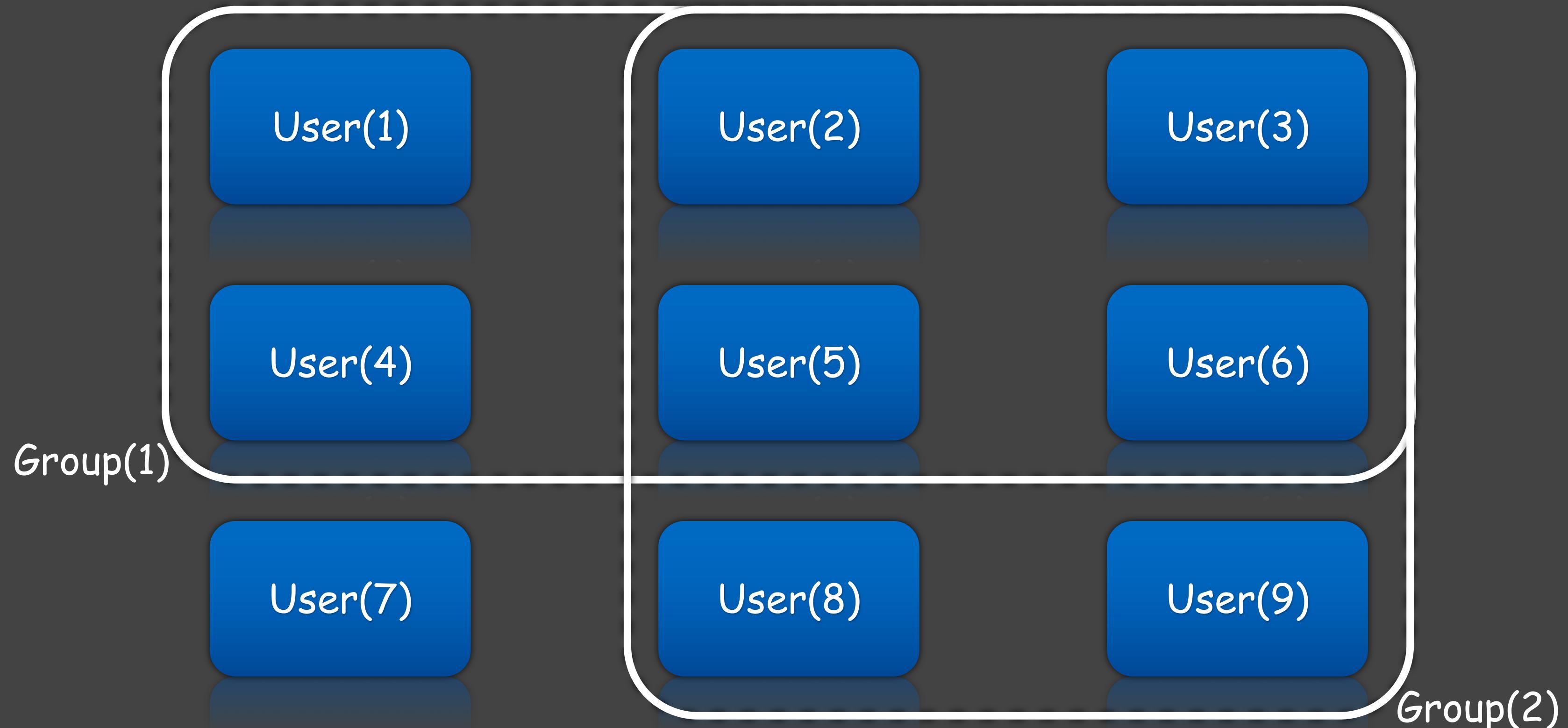
❖ Authorization

- Role-based access control

云机器学习平台 - 多租户



云机器学习平台 - 多租户



Conclusion

- ◆ 人工智能与机器学习介绍
- ◆ 机器学习算法原理与实现
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Thanks

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www.ecug.org