

- PostgreSQL And Artificial
- Intelligence
- Slides:

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- https://github.com/ibrarahmad/PostgreSQLTalks/tree/mai
- n/Conferences/



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#### Software Career

Software industries since 1998.

#### PostgreSQL Career

- Working on PostgreSQL Since 2006.
- EnterpriseDB (Senior Software Architect) 10 Years
- Percona (Principal Engineer) 2018 Present

#### **Open-source**

- PostgreSQL
- Google Chrome
- Google Chromium Project.

#### PostgreSQL Books

- PostgreSQL Developer's Guide
- PostgreSQL 9.6 High Performance

## PostgreSQL And Artificial Intelegence





#### Artificial Intelligence

#### What is Artificial Intelligence?

• Machines that mimic "cognitive" functions that humans associate with the human mind, such as "learning" and "problem solving"\*.



#### Artificial Intelligence

• "Artificial Intelligence is no match for natural stupidity"

Albert Einstein

• "A machine with strong A.I. is able to think and act just like a human. It is able to learn from experiences"

Albert Einstein

• "I think the development of full artificial intelligence could spell the end of the human race"

Stephen Hawking

• "The one who becomes the leader in this sphere will be the ruler of the world"

**Vladimir Putin** 

#### Types of Artificial Intelligence

ANI - Artificial Narrow Intelligence

1

3

2

) AGI - Artificial General Intelligence

) ASI - Artificial Super-Intelligence

- 1. ANI has a narrow range of ability to perform specific tasks.
- 2. AGI can perform different tasks that humans can do.
- 3. ASI is more capable than a human, learning from past experience and from new data to do a variety of tasks.



#### Sub-Fields of Artificial Intelegence





#### Types of Machine Learning

) Supervised Learning (Regression, Classification)

) Un-Supervised Learning (Association, Clustering)

3 Reinforcement Learning (Robotics)

1

2



#### **Prococess Of Machine Learning**



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#### Deep Learning

#### **Neural Networks**

- Neural Networks, with more layers and modules
- It has a model which can take any input/output type and size



#### Deep Learning



#### Machine Learning Vs Deep Learning





#### Demonstration



#### **Demonstration Using Postgres**

- Does an Integer Have Non-Leading Zeros?
- 31903 is true
- 82392 is false

## **Install PL/Perl**

- CREATE EXTENSION IF NOT EXISTS plperl;
- All queries in this presentation can be downloaded from <u>https://momjian.us/main/writings/ pgsql/Al.sql</u>.

#### **Generate Tensor**

- CREATE OR REPLACE FUNCTION generate\_tensor(value INTEGER) RETURNS BOOLEAN[] AS \$\$
- **my** \$value = shift;
- my @tensor = (
- # this many digits or more?
- (map { length(\$value) >= \$\_ } 1..10),
- # divisible by zero? \$value % 10 == 0, );
- # map to t/f
- grep { \$ \_ = (\$ \_ ? 't' : f') } @tensor; return encode\_typed\_literal(\@tensor, 'boolean[]'); \$\$
   LANGUAGE plper1 STRICT;

#### Create and Populate Input Layer

- CREATE TABLE training\_set(value INTEGER, training\_output BOOLEAN, tensor BOOLEAN[]);
- WITH randint (value) AS ( SELECT (random() \* (10 ^ (random() \* 8 + 1)::integer))::integer
- FROM generate\_series(1, 10000) ) INSERT INTO training\_set
- SELECT value, value::text LIKE '%0%', generate\_tensor(value) FROM randint;

#### Input Layer

• SELECT \* FROM training\_set LIMIT 10;

Value.	training_output	tensor
	-+	+
28762748	f	<pre>  {t,t,t,t,t,t,t,t,f,f,f}</pre>
44550313	t	<pre>  {t,t,t,t,t,t,t,t,f,f,f}</pre>
72	f	<pre>  {t,t,f,f,f,f,f,f,f,f,f,f,f</pre>
4891026	t	<pre>  {t,t,t,t,t,t,t,f,f,f,f}</pre>
3413	f	<pre>  {t,t,t,t,f,f,f,f,f,f,f,f}</pre>
62	f	<pre>  {t,t,f,f,f,f,f,f,f,f,f,f,f</pre>
86517976	f	<pre>  {t,t,t,t,t,t,t,t,f,f,f}</pre>
967	f	<pre>  {t,t,t,f,f,f,f,f,f,f,f,f,f</pre>
636667644	f	<pre>  {t,t,t,t,t,t,t,t,f,f}</pre>
36419	f	<pre>  {t,t,t,t,t,f,f,f,f,f,f,f}</pre>

## Generate Weights for Tensor (Cont...)

CREATE OR REPLACE FUNCTION generate\_weight
(query TEXT, desired\_output BOOLEAN)
RETURNS REAL[]

#### **AS** \$\$

my \$rv = spi\_exec\_query(shift);
my \$status = \$rv->{status};
my \$nrows = \$rv->{processed};
my \$desired\_output = shift;
my @success\_neurons = ();
my @desired\_neurons = ();
my \$desired\_input = 0;

## Generate Weights for Tensor (Cont...)

```
foreach my $rn (0 .. $nrows - 1) {
       my $row = $rv->{rows}[$rn];
       my $tensor = $row->{ (sort keys %$row) [0] }; my $training output = $row-> {
               (sort keys %$row) [1] };
               # only process training rows that match our desired output
               foreach my $neuron (0 .. $#$tensor) {
                      $success neurons[$neuron] //= 0;
                       $desired neurons[$neuron] //= 0;
                      # Neuron value matches desired output value;
                      # does the value match the desired output?
                      if ($tensor->[$neuron] eq $desired output) {
                              # Prediction success/failures that match our desired output.
                              $success neurons[$neuron]++
                              if ($training output eq $desired output);
                                      $desired neurons[$neuron]++;
               $desired input++
               if ($training output eq $desired output);
```

#### Generate Weights for Tensor

```
my @weight = ();
```

```
my $sum = 0;
```

```
# compute percentage of tests that matched requested outcome
foreach my $neuron (0 .. $#success neurons {
        $weight[$neuron] = $desired neurons[$neuron] != 0 ? $success neurons[$neuron] /
        $desired neurons[$neuron] : 0;
        $sum += $weight[$neuron];}
# balance weights so they total the observed probability;
# this prevents an overly-predictive output value from skewing
# the results.
foreach my $neuron (0 .. $#weight) {
        $weight[$neuron] = ($weight[$neuron] / $sum) * ($desired input / $nrows);
```

```
return encode_typed_literal(\@weight, 'real[]');
```

\$\$ LANGUAGE plperl STRICT;

#### Create Tensor\_Mask

-- Return weights where our neuron value matches the desired output CREATE OR REPLACE FUNCTION tensor\_mask(tensor BOOLEAN[], weight REAL[], desired\_output BOOLEAN)

```
RETURNS REAL [] AS \$\$
```

```
my $tensor = shift;
```

```
my $weight = shift;
```

```
my $desired output = shift;
```

```
my @result = ();
```

```
elog(ERROR, 'tensor and weight lengths differ')
```

```
if ($#$tensor != $#$weight);
```

```
foreach my $i (0 .. $#$tensor) {
```

push(@result, (\$tensor->[\$i] eq \$desired\_output) ? \$weight->[\$i] : 0);

```
}
return encode_typed_literal(\@result, 'real[]');
```

```
$$ LANGUAGE plperl STRICT;
```

#### Create Sum\_Weight

**CREATE OR REPLACE FUNCTION** sum\_weight(weight **REAL**[]) **RETURNS REAL AS** \$\$

```
my $weight = shift;
```

```
my $sum = 0; # sum weights
```

```
foreach my $i (0 .. $#$weight) {
```

```
$sum += $weight->[$i];
```

```
return encode_typed_literal($sum, 'real');
```

```
$$ LANGUAGE plperl STRICT;
```

#### Create Soft\_Max

--Normalize the values so the probabilities total one CREATE OR REPLACE FUNCTION softmax(vall REAL, val2 REAL) RETURNS REAL[] AS \$\$

my \$val1 = shift;

```
my $val2 = shift;
```

- **my** \$sum = \$val1 + \$val2;
- # What percentage is each of the total?
- my @result = ( \$val1 / \$sum, \$val2 / \$sum, );

return encode\_typed\_literal(\@result, 'real[]');
\$\$ LANGUAGE plper1 STRICT;

#### Store Weights

CREATE TABLE tensor\_weight\_true
AS SELECT generate\_weight('SELECT tensor AS x1,
training\_output AS x2 FROM training\_set', true) AS
weight;

CREATE TABLE tensor\_weight\_false AS SELECT
generate\_weight('SELECT tensor AS x1, training\_output
AS x2 FROM training\_set', false) AS weight;

#### **Stored Weights**

SELECT \* FROM tensor\_weight\_true;
 weight

{0.020473005,0.021917565,0.024002228,0.026247077,0.0284 82921, \

0.030471962,0.032726202,0.034238704,0.036621932,0,0.064<br/>1184}

SELECT \* FROM tensor\_weight\_false;
 weight

{0,0.0820682,0.07662672,0.074060954,0.07129263,0.068018 064, \

0.06497674,0.061864104,0.059269458,0.058057636,0.064465 51}

#### Test 100

```
WITH test_set (checkval) AS ( SELECT 100 )
```

```
SELECT softmax(
    sum_weight(
        tensor_mask(
        generate_tensor(checkval),
        tensor_weight_true.weight,
    true)),
    sum_weight(
        tensor_mask(
        generate_tensor(checkval),
        tensor_weight_false.weight,false))
)
FROM test set, tensor weight true, tensor weight false;
```

softmax

```
{0.22193865,0.77806133}
```

## Test 101

```
WITH test set (checkval) AS
   SELECT 101
SELECT softmax (
   sum weight (
        tensor mask (
             generate tensor(checkval),
             tensor weight true.weight, true)),
   sum weight (
        tensor mask (
             generate tensor(checkval),
             tensor weight false.weight, false))
FROM test set, tensor weight true, tensor weight false;
                                                          softmax
\{0.11283657, 0.88716346\}
```

#### Test 487234987

```
WITH test set (checkval) AS
   SELECT 487234987
SELECT softmax(
   sum weight (
        tensor mask(
             generate tensor(checkval),
             tensor weight true.weight, true)),
   sum weight (
        tensor mask (
             generate tensor(checkval),
             tensor weight false.weight, false))
FROM test set, tensor weight true, tensor weight false;
                                                          softmax
\{0.68860435, 0.31139567\}
```

## Test One Thousand Values

# WITH test\_set (checkval) AS ( SELECT (random() \* (10 ^ (random() \* 8 + 1)::integer))::integer FROM generate\_series(1, 1000)

),

#### Second Table Expression

```
ai (checkval, output_layer) AS ( SELECT
checkval, softmax(
    sum_weight(tensor_mask(generate_t
    ensor(checkval),tensor_weight_true.wei
ght,true)),
    sum_weight(tensor_mask(generate_tensor
(checkval), tensor_weight_false.weight,
false)) ) FROM test_set,
tensor_weight_true, tensor_weight_false ),
```

#### Third Table Expression

```
analysis (checkval, cmp_bool, output_layer,
accuracy) AS (
   SELECT checkval, checkval::text LIKE '%0%',
   output_layer,
    CASE checkval::text LIKE '%0%'
    -- higher/lower than random chance WHEN
    true THEN output layer[1] - 0.5 ELSE
    output_layer[2] = 0.5
    END
FROM ai
```

#### **Final Table Expression**

SELECT \* FROM analysis UNION ALL SELECT
NULL, NULL, NULL, AVG(accuracy) FROM
analysis UNION ALL SELECT NULL, NULL,
NULL, SUM(CASE WHEN accuracy > 0 THEN 1
END)::REAL/COUNT(\*) FROM analysis;

Checkval   cmp_bool	output_layer	Accuracy
6   f	<pre>+</pre>	0.47080177068710327
61859931   f	{0.5459184,0.4540816}	-0.045918405055999756
53138008   t	{0.5459184,0.4540816}	0.045918405055999756
727   f	{0.11283657,0.88716346}	0.3871634602546692
33397006   t	{0.5459184,0.4540816}	0.045918405055999756
38380069   t	{0.5459184,0.4540816}	0.045918405055999756
8915576   f	{0.4306789,0.5693211}	0.06932109594345093
446   f	{0.11283657,0.88716346}	0.3871634602546692
(null)   (null)	(null)	0.15426481181383134
(null)   (null)	(null)	0.722

# Why to Use Database in Al?



#### Why Use Database?

- Machine learning requires a lot of data
- Most of your data is in your database
- Why not do machine learning where your data is, in a database?

#### Advantages of doing Machine Learning in a Database?

- Use the previous activity as training data
- Have seamless access to all your current data
- Take immediate action on AI results, e.g., commit transaction only if likely non-fraudulent
- Al can benefit from database transactions, concurrency, backup
- Other benefits include complex data types, full-text search, GIS, indexing
- Postgres can do GPU-based computations inside the database.

## General Artificial Intelligence Uses by Databases?

- User applications
  - Performance adjustments
  - Optimizer plans
  - Index creation/destruction
  - Database settings
  - Resource usage
- Alerting
  - Malicious activity
  - Resource exhaustion

## Questions

