Postgres and the Artificial Intelligence Landscape

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This presentation explains how to do machine learning inside the Postgres database.

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Outline

- 1. What is artificial intelligence?
- 2. Machine learning and deep learning
- 3. Demonstration using Postgres
- 4. Hardware/software efficiency
- 5. Tasks
- 6. Why use a database?

1. What is Artificial Intelligence?

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

What is Artificial about Artificial Intelligence?

If only the physical world exists, then human intelligence only differs from machine intelligence because it has not naturally developed. It hence differs only in how it is created. Human free will becomes an illusion.

History of Artificial Intelligence (AI)

- Pre-computer philosophy
- Robotics
- Turing test
- Expert systems
- AI winter
- Machine/deep learning
- Generative AI: output text, images, videos
- Like fusion energy, it is always ten years away

Artificial Intelligence Naming

"In a certain sense I think that artificial intelligence is a bad name for what it is we're doing here. As soon as you utter the words 'artificial intelligence' to an intelligent human being, they start making associations about their own intelligence, about what's easy and hard for them, and they superimpose those expectations onto these software systems."

Kevin Scott, chief technology officer of Microsoft.

Artificial Intelligence Reality

"AI algorithms are just math. And one of math's functions is to simplify the world so our brains can tackle its otherwise dizzying complexity. The software we call AI is just another way to arrive at complicated mathematical functions that help us do that."

Kevin Scott, chief technology officer of Microsoft.

2. Machine Learning and Deep Learning



The most comprehensive video I have seen about machine learning is at https://www. youtube.com/watch?v=r00gt-q956I.

Machine Learning



https://www.verypossible.com/insights/machine-learning-algorithms-what-is-a-neural-network

Unsupervised Machine Learning



https://medium.com/@machadogj/ml-basics-supervised-unsupervised-and-reinforcement-learning-b18108487c5a and a supervised-and a supervised-and a supervised and a supervised a supervised

Supervised Machine Learning





Reinforcement Machine Learning



Deep Learning



Result

https://www.zendesk.com/blog/machine-learning-and-deep-learning/

3. 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 https://momjian.us/main/writings/pgsql/landscape.sql.

Generate Tensor

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

The tensor has 11 neurons.

Create and Populate Input Layer with Training Data

Input Layer

SELECT * FROM <pre>training_set</pre> LIMIT 10;				
value	training_output	tensor		
+	++			
8360692	t	${t,t,t,t,t,t,t,t,f,f,f,f}$		
58297366	f	$\{t,t,t,t,t,t,t,t,t,f,f,f\}$		
569005	t	${t,t,t,t,t,t,f,f,f,f,f}$		
236	f	${t,t,t,f,f,f,f,f,f,f,f,f}$		
43	f	${t,t,f,f,f,f,f,f,f,f,f,f,f}$		
35	f	${t,t,f,f,f,f,f,f,f,f,f,f,f}$		
610510	t	$\{t,t,t,t,t,t,f,f,f,f,f\}$		
638484259	f	$\{t,t,t,t,t,t,t,t,t,t,f,f\}$		
5983	f	$\{t,t,t,t,f,f,f,f,f,f,f\}$		
34	f	${t,t,f,f,f,f,f,f,f,f,f,f,f}$		

Generate Weights for Tensor

```
-- Runs the supplied query and generates weights
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

```
foreach my $rn (0 .. $nrows - 1) {
           mv  $row = rv > rows [$rn]:
            mv $tensor = $row->{(sort kevs %$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] eg $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 = ();
        mv $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 mv $neuron (0 .. $#weight) {
                $weight[$neuron] = ($weight[$neuron] / $sum) *
                                   ($desired input / $nrows);
        return encode typed literal(\@weight, 'real[]');
$$ LANGUAGE plper1 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 $$
        my $tensor = shift;
        mv $weight = shift:
        my $desired output = shift;
        my \ @result = ();
       elog(ERROR, 'tensor and weight lengths differ')
                if ($#$tensor != $#$weight);
        foreach my $i (0 .. $#$tensor) {
                push(@result.
                        ($tensor->[$i] eq $desired output) ?
                        weight -> [$i] : 0):
        return encode_typed literal(\@result, 'real[]');
$$ LANGUAGE plper1 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 plper1 STRICT;
```

Create Soft_Max

```
# Normalize the values so the probabilities total one
CREATE OR REPLACE FUNCTION softmax(val1 REAL, val2 REAL)
RETURNS REAL[] AS $$
        my $val1 = shift;
        my $val2 = shift;
        my  $sum = exp($val1) + exp($val2);
        # What percentage is each of the total?
        mv @result = (
                exp($val1) / $sum.
                exp($va12) / $sum,
        );
        return encode typed literal(\@result, 'real[]');
$$ LANGUAGE plper1 STRICT:
```

Uses the exponential function (e^x)

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



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
 \{0.42048895, 0.57951105\}
```

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
```

{0.3893167,0.61068326}

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
```

 $\{0.5340263, 0.46597365\}$

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

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 * FRO	DM analysi	s	
UNION ALL			
SELECT NULL	, NULL, NU	JLL, AVG(accuracy)	
FROM analys	is		
UNION ALL			
SELECT NULL	, NULL, NU	JLL, SUM(CASE WHEN accura	<pre>cy > 0 THEN 1 END)::REAL/COUNT(*)</pre>
FROM analys	is;		
checkval	cmp_bool	output_layer	accuracy
	+	-+	+
35173962	f	{0.5102168,0.48978326	} -0.010216742753982544
27023	t	{0.4379818,0.5620182}	-0.06201818585395813
9669661	f	{0.48645663,0.5135433	7} 0.013543367385864258
6348988	f	{0.48645663,0.5135433	7} 0.013543367385864258
7736	f	{0.41372445,0.5862756	} 0.08627557754516602
910139	t	{0.4622842,0.5377158}	-0.037715792655944824
8642	f	{0.41372445,0.5862756	} 0.08627557754516602
98	f	{0.3652915,0.6347085}	0.13470852375030518
6	f	{0.34178793,0.6582120	7} 0.1582120656967163
962	f	{0.3893167,0.61068326	} 0.11068326234817505
(null)	(null)	(null)	0.049643657088279725
(null)	(null)	(null)	0.722

4. Hardware/Software Efficiency: Software

- Client-side
 - Matlab
 - Scikit, https://kb.objectrocket.com/postgresql/ machine-learning-with-python-and-postgres-1114
 - Tensorflow
 - PyTorch
 - Weka
- Server-side
 - PL/Python with the above libraries
 - MADlib, https://www.youtube.com/watch?v=uLW5By66Lf0
 - Scikit, https://www.cybertec-postgresql.com/en/ machine-learning-in-postgresql-part-1-kmeans-clustering/
 - pgvector, https://github.com/pgvector/pgvector

Gpu



- Tensors can have millions of neurons
- Deep learning can use thousands of tensor layers
- Every neuron passes its data to every neuron in the next layer
- This requires a lot of repetitive calculations
- GPUs are designed to efficiently perform simultaneous repetitive computations
- PG-Strom adds GPU acceleration to Postgres

5. Tasks

- Chess (Deep Blue)
- Jeopardy (Watson)
- Voice recognition (Siri)
- Search (Google)
- Recommendations (Netflix)
- Image detection
- Weather forecasting (NOAA)

Fraud Detection Example

Choose attributes:

- Charge amount
- Magnetic swipe, chip, pin, online charge
- Vendor distance from chargee billing address
- Distance from last chargee charge
- Vendor country
- Previous charges to this vendor for chargee
- Previous fraudulent charges by vendor

Fraud Detection Steps

- 1. Choose attributes
- 2. Create machine learning neurons for each attribute
- 3. Create training data, with the required attributes of each transactions and its outcome, i.e., valid or fraudulent
- 4. Feed the training data into the machine to set the weights of each neuron, based on how much the neuron's attribute predicts the validity or fraudulence of transactions
- 5. Start feeding real data into the machine and get results
- 6. Feed correct and incorrect results back into the neurons to improve the accuracy of the weights, and to adjust for changes in the environment

6. Why Use a 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 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
- AI 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 (https://momjian.us/ main/blogs/pgblog/2020.html#June_29_2020)

General Artificial Intelligence Uses by Databases?

• User applications not already covered

- human language queries and results (large language model)
- retrieval-augmented generation (RAG)
- Performance adjustments
 - optimizer plans
 - index creation/destruction
 - database settings
 - resource usage
- Alerting
 - malicious activity
 - resource exhaustion

Conclusion





https://www.flickr.com/photos/corneveaux/