Hacking the Query Planner

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Why this talk?

One thing I could really use is an overview of what order things get done in the planner. What are the major phases of processing and what's the function of each one? — R. Haas

Agenda

• What's the problem we need to solve?

- Structure of the planner
- Some key data structures
- Cost estimation
- Future work

Overall backend structure

- Parser
 - Determines the semantic meaning of a query string
- Rewriter
 - Performs view and rule expansion
- Planner
 - Designs an execution plan for the query
- Executor
 - Runs the plan

The planner's problem

- Find a good query plan
- Don't spend too much time (or memory) finding it
- Support the extensible aspects of Postgres
 - eg, custom data types, operators, functions
 - this means hard-wired knowledge about data types and operators should be avoided as much as possible

What's a plan, exactly?

- A plan is a tree of plan nodes
- Each plan node represents a specific type of processing to do, with all details executor needs
- At execution, a node yields a stream of tuples
- Relation scan nodes get their tuples from a table
- Most other node types read tuple stream(s) from child plan nodes, and process them somehow to create their result stream

Types of plan nodes

- Relation scan nodes
 - Sequential, plain index, bitmap index
- Join nodes
 - Nestloop, nestloop with inner indexscan, hash, merge
- Special plan nodes
 - Sort, aggregate, set operations (UNION etc)

Attributes of a plan node

- Data source
 - Relation to scan, if a table scan
 - Input plan node, if a "processing" node
 - Two input plan nodes, if a join node
- Target list (expressions to compute and return)
 - Think SELECT list
- Selection conditions ("qualifiers" or "quals")
 - Think WHERE conditions

Estimates for a plan node

- Output row count
 - Need this to estimate sizes of upper joins
- Average row width
 - Need both rowcount and width to estimate workspace for sorts, hashes, etc that must store node's output
- Total cost
 - Usually, the thing we want to minimize
- Startup cost
 - For LIMIT queries, we interpolate between startup and total costs, since not all rows will be fetched

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But first, some jargon

- **Var** = variable = table column reference
- **Rel** = relation = real or virtual table
- **Base rel** = primitive FROM item (actual table, or separately-planned subquery, or set-returning function)
- **Join rel** = join relation (the result of joining a specific set of base relations)
- **Qual** = qualifier = WHERE clause or filter condition
- **Join qual** = qualifier that uses Vars from more than one base relation

Phases of planning

- Preprocessing
 - simplify the query if possible; collect information
- Scan/join planning
 - decide how to implement FROM/WHERE
- Query special feature handling
 - deal with plan steps that aren't scans or joins
- Postprocessing
 - convert results into form the executor wants

Early preprocessing

- Simplify scalar expressions
- Expand simple SQL functions in-line
- Simplify join tree

Simplify scalar expressions

What we know how to do is mainly constant-folding:

 $2 + 2 \Rightarrow 4$

CASE WHEN 2+2 = 4 THEN x+1ELSE 1/0 END

 \Rightarrow x+1

... not "ERROR: division by zero", please

Why bother simplifying?

- Do computations only once, not once per row
- View expansion and SQL function inlining can expose constant-folding opportunities not visible in the original query, so query author wasn't necessarily stupid
- Simplifying takes a lot of load off the estimation functions, which by and large can't cope with anything more complex than Variable = Constant

Constant-folding is simple

- All we need for constant-folding is the ability to hand an expression tree to the executor to execute; the planner need know nothing of the operation's detailed semantics
- People are sometimes surprised that we don't simplify cases like reducing x + 0 ⇒ x
- Problem is that would require a **lot** of datatypespecific and operator-specific knowledge, plus infrastructure for extensions to add such knowledge
- ... but maybe someday ...

Expand simple SQL functions

CREATE FUNCTION incr(int) RETURNS int AS 'SELECT \$1 + 1' LANGUAGE SQL;

SELECT incr(col) FROM tab;

 \Rightarrow

SELECT col + 1 FROM tab;

Simplify join tree

- Flatten ("pull up") sub-selects if possible
 - else, we'll recurse to generate sub-plans
- Flatten UNION ALL, expand inheritance trees
- Reduce join strength (outer join \Rightarrow inner join)
- Convert IN, EXISTS sub-selects to semi-joins
- Identify anti-joins

Flattening a simple view

```
CREATE VIEW v AS
SELECT a, b+c AS d FROM t WHERE x > 0;
SELECT v.a, v.d FROM v WHERE v.a = 42;
Rewriter produces:
SELECT v.a, v.d FROM
(SELECT a, b+c AS d FROM t WHERE x > 0) v
WHERE v.a = 42;
```

Sub-select flattening produces:

SELECT t.a, t.b + t.c FROM t
WHERE t.x > 0 AND t.a = 42;

I lied about the ordering ...

- Actually, these preprocessing steps are done in a very specific order so that opportunities exposed by one step can be exploited later
- Some of them have to be intermixed in a single recursive pass over the query tree
- Getting all the optimizations to happen without duplicate processing is a bit tricky

Later preprocessing

- Determine where WHERE clauses ("quals") should be evaluated
 - In general, we want to use each qual at the lowest possible join level
- Identify all referenced table columns (Vars), and find out how far up in the join tree their values are needed
- Build equivalence classes for provably-equal expressions
- Gather information about join ordering restrictions
- Remove useless joins (needs results of above steps)

Scan/join planning

- Basically deals with the FROM and WHERE parts of the query
- Knows about ORDER BY too
 - mainly so that it can design merge-join plans
 - but also to avoid final sort if possible
- Cost estimate driven

Scan/join planning

- Identify feasible scan methods for base relations, estimate their costs and result sizes
- Search the join-order space, using dynamic programming or heuristic "GEQO" method, to identify feasible plans for join relations
- Honor outer-join ordering constraints
- Produce one or more "Path" data structures

Paths versus Plans

- A **Path** is a simplified representation of a potential plan tree
- We build many Paths during planning, but convert only the finally selected Path to a fullfledged Plan that the executor could handle
- Saves time, memory space when considering a large number of competing alternative plans

Path generation/comparison

- Generate a Path data structure for each feasible method of performing a scan or join
- Compare cost to previously-generated Paths for the same base relation or join relation
- Immediately discard inferior Paths
 - Keep only those that are cheapest (in either total or startup cost) for a given output sort ordering of their relation
 - Get rid of useless sort orderings, too

Join searching

- Multi-way joins have to be built up from pairwise joins, because that's all the executor knows how to do
- For any given pairwise join step, we can identify the best input Paths and join methods via straightforward cost comparisons, resulting in a list of Paths much as for a base relation
- Finding the best ordering of pairwise joins is the hard part

Join searching

- We usually have many choices of join order for a multi-way join query, and some orders will be cheaper than others
- If the query contains only plain inner joins, we can join the base relations in any order
- Outer joins can be re-ordered in some but not all cases; we handle that by checking whether each proposed join step is legal

Standard join search method

- Generate paths for each base relation
- Generate paths for each possible two-way join
- Generate paths for each possible three-way join
- Generate paths for each possible four-way join
- Continue until all base relations are joined into a single join relation; then use that relation's best path
- This "dynamic programming" approach was invented many years ago for IBM's System R

Inner indexscans are special

- The dynamic programming method supposes that any join is formed from independent Paths for the two input relations
- Doesn't work for nestloop with inner indexscan, because using a join clause as an index condition requires the outer variable(s) to be available from the particular outer relation being joined to
- We keep separate lists of "inner indexscan" Paths for each base relation that has any indexable join clauses, organized according to required outer rels

Join searching is expensive

- An **n**-way join problem can potentially be implemented in **n!** (**n** factorial) different join orders
- Considering all possibilities gets out of hand **real fast**, and is not feasible for queries with more than around ten base relations
- We use a few heuristics, like not considering clause-less joins
- With too many relations, fall back to "GEQO" (genetic query optimizer) search, which is even more heuristic and tends to fail to find desirable plans

Heuristics used in join search

- Don't join relations that are not connected by any join clause, unless forced to by join-order restrictions
 - Implied equalities count as join clauses, so this rule seldom leads us astray
- Break down large join problems into sub-problems according to the syntactic JOIN/sub-select structure
 - Actually, it's done by not merging sub-problems to make a big problem in the first place (see "collapse limits")
 - This frequently sucks; would be useful to look for smarter ways of subdividing large join trees

Genetic query optimizer

- Treats join order searching as a Traveling Salesman Problem, i.e., minimize the length of a "tour" visiting all "cities" (base relations)
- Does a partial search of the tour space using heuristics found useful for TSP
- Problem: join costs don't behave very much like inter-city distances; they interact. This makes the TSP heuristics not so effective
- This area desperately needs improvement

Query special feature handling

- Deal with GROUP BY, DISTINCT, aggregate functions, window functions
- Deal with UNION/INTERSECT/EXCEPT
- Apply final sort if needed for ORDER BY
- This code is very ad-hoc, not very pretty, not terribly bright either
- Maybe someday we will rewrite into generateand-compare-paths style

Postprocessing

- Convert to representation used by executor
- Expand Paths to Plans
- Example task: renumber Var nodes to meet executor's requirements (Vars in join nodes must be labeled "OUTER" or "INNER", not with original base relation's number)
- Mostly boring, except when it breaks

I lied again ...

- Actually, Path-to-Plan conversion happens after scan/join planning, and before query special feature handling
- Other postprocessing does happen at the end
- This is because the query special feature code doesn't use Paths to represent what it's thinking about; it works on actual Plan trees
- If we were to switch over to doing special features with Paths, presumably this would change

A map of backend/optimizer/

Subdirectory	Contents
geqo	GEQO join searching
path	Path generation and cost estimation
plan	Main planning driver code
prep	Preprocessing
util	Miscellaneous

... and don't forget

backend/utils/adt/selfuncs.c

operator-specific selectivity functions

geqo/geqo_copy.c geqo/geqo_cx.c geqo/geqo_erx.c geqo/geqo_eval.c geqo/geqo_main.c geqo/geqo_misc.c geqo/geqo_mutation.c geqo/geqo_ox1.c geqo/geqo_ox2.c geqo/geqo_pmx.c geqo/geqo_pool.c geqo/geqo_px.c geqo/geqo_random.c geqo/geqo_recombination.c geqo/geqo_selection.c 1200 lines

boring support code unused method for generating a mutated tour active method for generating a mutated tour evaluate cost of tour glue code debug printout code randomly mutate a tour (by swapping cities) unused method for generating a mutated tour unused method for generating a mutated tour unused method for generating a mutated tour boring support code (manage "pool" of tours) unused method for generating a mutated tour boring support code boring support code randomly select two "parent" tours from pool

path/allpaths.c path/clausesel.c path/costsize.c path/equivclass.c path/indxpath.c path/joinpath.c path/joinrels.c path/orindxpath.c path/pathkeys.c path/tidpath.c 8000 lines

core scan/join search code (mostly about base rels) clause selectivity (driver routines mostly) estimate path costs and relation sizes support code for managing equivalence classes core path generation for indexscan paths core path generation for joins core scan/join search code (mostly about join rels) path generation for "OR" indexscans support code for managing PathKey data structures core path generation for TID-scan paths (WHERE ctid = constant) plan/analyzejoins.c plan/createplan.c plan/initsplan.c plan/planagg.c plan/planmain.c plan/planner.c plan/setrefs.c plan/subselect.c 8500 lines late-stage join preprocessing build Plan tree from selected Path tree scan/join preprocessing (driven by planmain.c) special hack for planning min/max aggregates driver for scan/join planning driver for all "extra" query features Plan tree postprocessing handle sub-selects (that aren't in FROM)

prep/prepjointree.c	early-stage join preprocessing
prep/prepqual.c	WHERE clause (qual) preprocessing
prep/preptlist.c	target list preprocessing (mostly just for INSERT/ UPDATE/DELETE)
prep/prepunion.c	plan set operations (UNION/INTERSECT/EXCEPT, but not simple UNION ALL); also has some support code for inheritance cases ("appendrels")
3000 lines	

util/clauses.c	assorted code for manipulating expression trees, includes constant folding and SQL function inlining
util/joininfo.c	support code for managing lists of join clauses
util/pathnode.c	code for creating various sorts of Path nodes, and for comparing the costs of different Path nodes \Rightarrow add_path() can be seen as the very heart of the planner
util/placeholder.c	code for managing PlaceHolderVars
util/plancat.c	code for extracting basic info about tables and indexes from the system catalogs (sets up RelOptInfo and IndexOptInfo)
util/predtest.c	code for proving that a WHERE clause implies or contradicts another one; used for constraint exclusion and seeing if partial indexes can be used
util/relnode.c	support code for managing RelOptInfo nodes
util/restrictinfo.c	support code for managing RestrictInfo nodes
util/tlist.c	support code for managing target lists
util/var.c	support code for managing Vars
7000 lines	

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PathKeys

- PathKeys are a List structure representing the sort ordering of the output tuples of a Path; for example ORDER BY x, y is represented by a list of a PathKey for x and a PathKey for y
- They can also represent a desired ordering
- "Canonical" pathkeys are used to make pathkey comparison cheap (we can use pointer equality)
- For more info, read src/backend/optimizer/README

EquivalenceClasses

- An EquivalenceClass represents a set of values that are known equal as a consequence of clauses like WHERE x = y AND y = z
- By transitivity, we can deduce that any two members of an EquivalenceClass are equal
- EquivalenceClasses also represent the value that a PathKey orders by (since if x = y, then ORDER BY x must be the same as ORDER BY y)
- Again, see src/backend/optimizer/README

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Cost estimation

- Everything I've talked about so far is just mechanism for generating alternatives to consider
- Cost estimation is what really drives the planner's behavior
- If the planner can't generate the plan you want, you need to fix the mechanism
- If it generates and rejects the plan you want, you need to fix the cost estimation

Cost estimation: basic theory

- Cost of a plan node is assumed to increase linearly from startup cost to total cost as more tuples are returned
- With a LIMIT we will stop short of paying the total cost, if we fetch fewer than the total number of tuples



of tuples returned

Cost estimation: ugly reality

- Sometimes the real world is not so linear
- If we're selecting a subset of scanned tuples, we will skip over some tuples
- Non-uniform distribution of desired tuples results in non-linear runtime
- This can result in seriously bad estimates for small LIMITS



of tuples returned

Plan cost models

- For each plan node type, costsize.c has a function to estimate its cost in terms of the primitive cost variables (page reads, operator evaluations, etc), given estimates about the numbers of rows, total data size, etc
- These models are a bit simplistic, but for the most part when the planner falls down, it's not a modeling failure but a statistical failure
- "Garbage in, garbage out" applies here!

Result sizes for relation scans

- Number of rows returned by any given table scan is estimated as the raw relation size multiplied by the "selectivity" of relevant WHERE conditions
- Raw relation size is taken to be tuple density found by last ANALYZE (that is, # live tuples / # blocks) times relation's current size in blocks
- Not perfect, but seems to work pretty well
- Width is just the sum of the per-column average widths estimated by ANALYZE for all the variables needed in the query

Result sizes for joins

- Number of rows returned is the Cartesian product size (product of estimated rowcounts of input relations) multiplied by the selectivity of relevant join conditions
- ... with some special twiddling for outer joins, for instance a LEFT JOIN cannot produce fewer rows than its left input has
- Width is sum of the per-column average widths for variables that are needed above the join

Selectivity is the hard part

- Applicable WHERE/JOIN ON conditions are broken into "clauses" that are combined with AND/OR/NOT
- Estimate selectivity of each clause separately, then combine results
- We have per-operator and per-special-clause selectivity estimation functions (these are mostly in utils/adt/selfuncs.c)
- Combination of per-clause estimates is done in clausesel.c
- AND/OR/NOT combinations are easy, **if** the clauses are independent ... but often they are not, and we get a bad result
- clausesel.c has some smarts for range restrictions, that is
 X >= C1 AND X <= C2

Operator selectivity functions

- "Restriction" estimators are used for clauses containing Vars of just one relation
- These generally don't try to handle anything more complex than "Var op Constant" (but we have a liberal definition of "Constant")
- "Join" estimators are used for clauses containing Vars from more than one relation
- These generally don't try to handle anything more complex than "Var op Var"
- Lots of unfinished work here

Aggressive constant-folding

- When trying to reduce a clause sub-expression to a constant for selectivity estimation, we will substitute any available values for parameter symbols, and will evaluate stable as well as immutable functions
- For example,

ts_col >= now() - interval '1 day'

will be estimated using current time minus one day as the comparison constant

Cost estimation API issues

- The "per operator" selectivity functions mostly are not truly specific to a single operator; rather we use functions like eqsel and scalarltsel, which embody knowledge about a class of operators
- I have a feeling that this is not the best API, mainly because it doesn't seem extensible to cover estimation of interrelated clauses. But changing it would break a lot of extension modules ...
- Also, nobody's ever fixed the Berkeley-era omission of selectivity estimators for functions that are accessed directly rather than via operators

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Parameterized scans

- A nestloop inner indexscan is basically a scan parameterized by values from the current row of the outer relation
- Sometimes it'd be useful to parameterize a larger chunk of the plan than a single base-relation scan (another way to say that is we'd like an indexscan to be able to use a parameter from more than one join level up)
- We need this in situations involving join order restrictions
- Problem #1: avoid explosion in number of paths to consider
- Problem #2: size estimates for inner scans are not independent of what the outer relation is

Foreign data wrappers

- As of 9.1, wrappers are on their own to produce plans and cost estimates for scans of foreign tables
- This obviously isn't good for the long run
- Particularly bad: no support for inner indexscan on a foreign table
- Need to figure out what sorts of functionality FDWs need, then refactor existing planner code to provide that in a reasonably clean fashion

Conclusion

- We really need more people working on the planner, the selectivity functions, etc
- I hope this talk has demystified the planner a bit, and given you some idea of where things can be improved