Enterprise Control Language (ECL) Workshop



The Four Corners of ECL

Foundations in Big Data





Welcome! – Workshop Overview

- ✓ Setting up your IDEs
- ✓ Connecting to your cluster
- ✓ Basic Queries
- ✓ Data Definitions
- ✓ Advanced Relational Queries
- ✓ Data Evaluation CrossTab Reports
- ✓ Data Transformation
- ✓ Getting Queries ready for ROXIE
- ✓ Machine Learning Analytics





What is ECL and common ECL terms:

- ✓ ECL is a Query and ETL (Extract, Transform, & Load) Language
 - Scalable, Extensible, Declarative, Non-Procedural
 - Also supports Rapid Data Delivery Engine Queries and Data Refinery ETL operations (ROXIE and THOR)
- ✓ ECL code is made up of ECL *Definitions and Actions*
 - ECL *Definitions* define *what* things are, not *how* to do them
 - Their order of appearance in source code does not define execution order (ECL is non-procedural)
- ✓ ECL Actions cause compilation and execution
- ✓ ECL *Functions* (definitions with parameters) are used in ECL expressions
- ✓ ECL files are stored in folders in the ECL Repository



HPCC (High Performance Computing Clusters)

LN's HPCC environment contains *clusters* which you define and use according to your needs.

There are two types of clusters in the HPCC:

- 1. Data Refinery (THOR) Used to process every one of billions of records in order to create billions of "improved" records.
 The ECL Agent (HTHOR) is also used to process simple jobs that would be an inefficient use of the THOR cluster.
- 2. Rapid Data Delivery Engine (ROXIE) Used to search quickly for a particular record or set of records.



Data on the HPCC

THOR/ECL Agent (HTHOR):

- Data loading is controlled through the Distributed File Utility (DFU) Server.
- Data typically arrives on the Landing Zone (for example, by FTP).
- File movement is initiated by DFU.
- Data is copied from the Landing Zone and is distributed (sprayed) into the Data Refinery (THOR).
- Data can now be further processed (<u>Extract Transform and Load process</u>).
- Data retrieval process places the file back on the landing zone (despraying).
- A single physical file is distributed into multiple physical files across the nodes of a cluster. The aggregate of the physical files creates one logical file that is addressed by the programmers of ECL code.

ROXIE

- Data and queries are compiled in ECL IDE and deployed (published) in ECL Watch.
- Data moves in parallel from THOR nodes to the receiving ROXIE nodes. Parallel bandwidth utilization improves the speed of putting new data into play.



HPCC Client Tools

ECL IDE



A full-featured GUI for ECL development providing access to the ECL repository and many of the ECL Watch capabilities.

Uses various ESP services via SOAP.

Provides the easiest way to create:

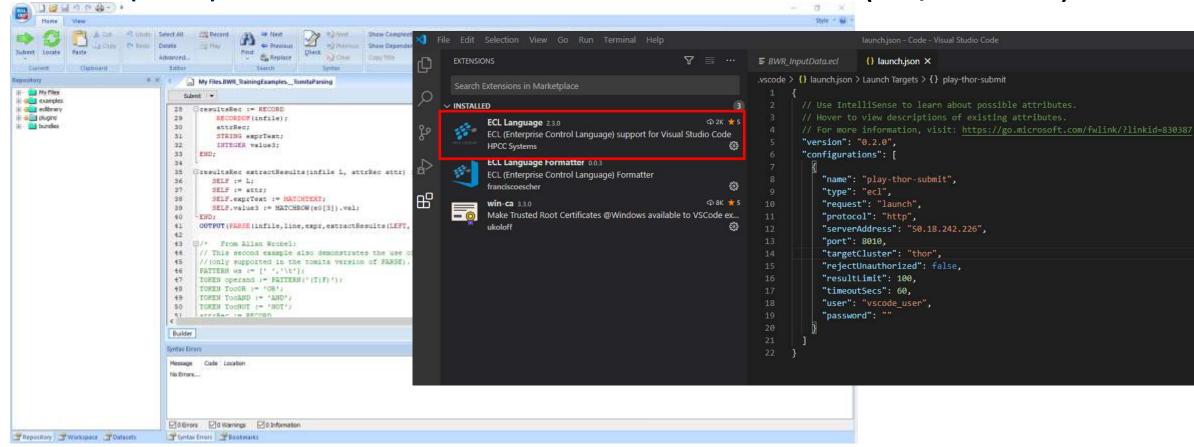
- Queries into your data.
- ECL Definitions to build your queries which:
 - Are created by coding an expression that defines how some calculation or record set derivation is to be done.
 - Once defined, can be used in succeeding ECL definitions.



Integrated Development Environments

ECL IDE (Win)

Visual Studio Code (Ux/MacOS)



And CLI too! ECL.EXE



HPCC Client Tools

ECL Watch

A web-based query execution, monitoring and file management interface. It can be accessed via ECL IDE or a web browser.

ECL Watch allows you to:

- See information about active workunits.
- Monitor cluster activity.
- Browse through previously submitted WUs:
 - See a visual representation of the data flow within the WU.
 - Complete with statistics which are updated as the job progresses.
- Search through files and see information including:
 - Record counts and layouts.
 - Sample records.
 - The status of all system servers whether they are in clusters or not.
- View log files.
- Start and stop processes.





Workshop Clusters

3 Node THOR

2 Node ROXIE

http://training.us-hpccsystems-dev.azure.lnrsg.io:8010/

4 Node THOR

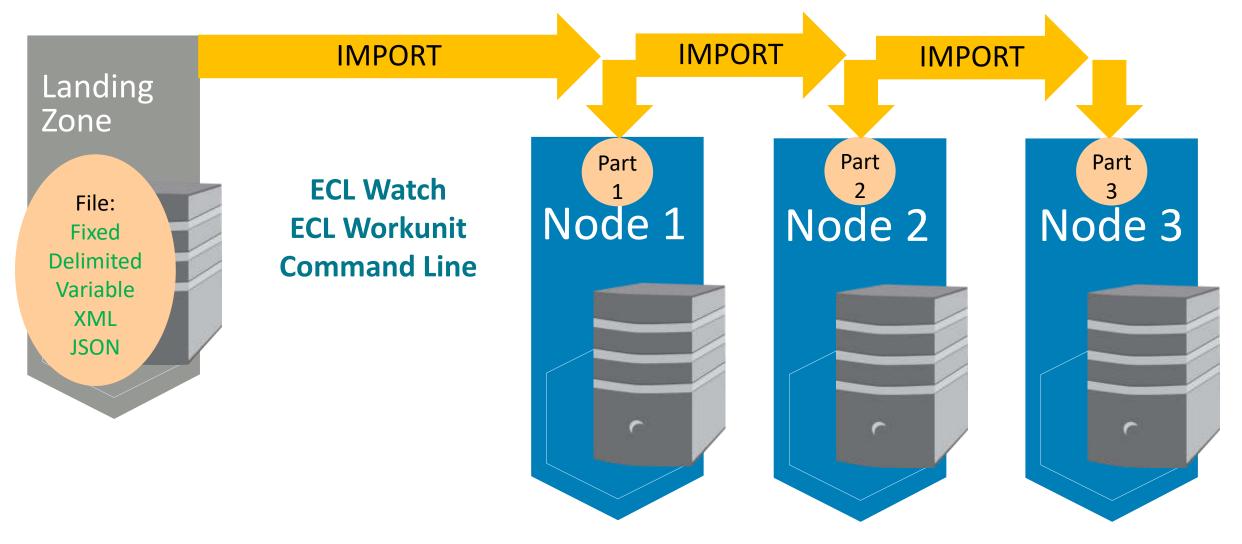
One way ROXIE

https://play.hpccsystems.com:18010



SPRAY/IMPORT Operation

HPCC Systems Cluster

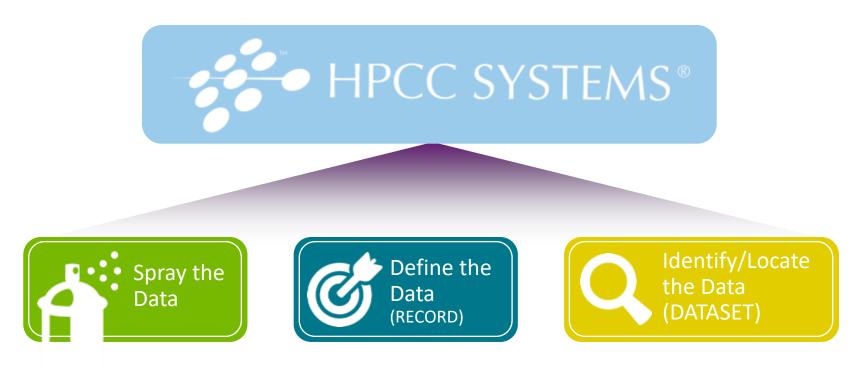


Referenced in ECL as a single logical file...



Three ECL Data Rules

Before you begin to work on any data in the HPCC cluster, you must always do three things:





RECORD Structure

```
DefinitionName := RECORD
 fields;
END;
```

- DefinitionName The name of the RECORD structure.
- fields The data type and name of each data field.

A **RECORD** structure defines the field layout of a DATASET, recordset, INDEX, or TABLE. The keywords RECORD and END may be replaced with the open and close curly braces ({}) and the semi-colon field delimiters may be replaced with commas.

```
Example:
{field1, field2, fieldn}
```



RECORD Structure Example:

```
EXPORT Layout_Company := RECORD
UNSIGNED sic_code;
STRING1
           source;
STRING120 company_name;
STRING10
           prim_range;
STRING2
           predir;
STRING28
           prim_name;
           addr_suffix;
STRING4
STRING2
           postdir;
STRING5
           unit_desig;
STRING8
           sec_range;
STRING25
           city;
STRING2
           state;
STRING5
           zip;
STRING4
           zip4;
           phone;
STRING10
END;
```



DATASET

```
name := DATASET( file, recorddef, THOR [options]);
name := DATASET( file, recorddef, CSV [ ( options ) ] );
name := DATASET( file, recorddef, XML( path,[options] ) );
name := DATASET( file, recorddef, JSON( path,[options] ) );
name := DATASET( file, recorddef, PIPE( command ) );

√ name – The definition name by which the file is subsequently referenced.

   ✓ file – A string constant containing the logical filename.
   ✓ recorddef – The RECORD structure of the dataset.
   ✓ options – options specific to the dataset type.
   ✓ path - A string constant containing the full XPATH to the tag that delimits the
     records in the file
   ✓ command – third-party program that creates the dataset.
```

DATASET introduces a new data file into the system with the specified recorddef layout.



RECORD and DATASET example

```
Layout_Company := RECORD
UNSIGNED
              sic_code;
STRING120
              company name;
STRING10
              prim range;
STRING2
              predir;
STRING28
              prim name;
              addr_suffix;
STRING4
              postdir;
STRING2
STRING5
              unit desig;
STRING8
              sec_range;
STRING25
              city;
STRING2
              state;
STRING5
              zip;
STRING4
              zip4;
END;
```

EXPORT File_Company_List := **DATASET**('~CLASS::Company_List', Layout_Company, THOR);



Datasets, Recordsets, and Filters

Physical **Datasets**:

can be *memory* or *disk* based

Recordsets:

subsets of physical datasets or other recordsets based on *filtering* conditions

Filters:

A Boolean (true or false) expression specified in parentheses following a dataset or recordset name.

```
Lifestyle_File := Polk.File_Polk_DS(DOB > '191604' AND
                                 DOB <= '196204' AND
                                 DOB != "
```



Recordset Filters

- Any Boolean expression within parentheses following a Dataset or Recordset name is a filter expression, used to define the specific subset of records to use
- Multiple filter conditions may be specified by separating each filter expression with a comma (,) or explicitly using the AND operator

```
T Names := People(Lastname >= 'T', Lastname < 'U');
```

RateGE7Trds := Trades(trd rate >= 7);

ValidTrades := Trades(NOT IsMortgage AND NOT IsClosed);



Actions and Definitions

```
Actions instigate Workunits, Definitions don't
 OUTPUT(People); // will instigate a Workunit (WU)
 People;
                     // implicit action, will instigate a Workunit
Actions can be ECL definitions
 A := OUTPUT(People); // will NOT instigate a Workunit
 A;
Functions that return scalar values can act as Actions
 B := COUNT(People); // definition will not instigate a WU
 COUNT(People); // WILL instigate a Workunit
  B;
```



ECL Definition Types - Boolean

Boolean -

A Boolean definition is a logical expression resulting in a TRUE/FALSE result

```
IsGoodClass := TRUE;
```

IsFloridian := People.state = 'FL';

IsSeniorCitizen := People.Age >= 65;



ECL Definition Types - Value

Value -

A Value definition is an arithmetic or string expression with a single-valued result

```
ValueTrue := 1;
FloridianCount := COUNT(People(IsFloridian));
SeniorAvgAge := AVE(People(IsSeniorCitizen), People.Age);
```



ECL Definition Types - Set

Set -

A Set definition is a set of explicitly declared constant values or expressions within square brackets. All elements must be the same type.

```
SetTrueFalseValues := [0, 1];

SetSoutheastStates := ['FL','GA','AL','SC'];

SetStatusCodes := ['1','X','9','W'];

SetInts := [1,2+3,45,def1,7*3,def2];
```



ECL Definition Types - Recordset

Recordset -

A Recordset definition is a filtered dataset or recordset

FloridaPeople := People(IsFloridian);

SeniorFloridaPeople := FloridaPeople(IsSeniorCitizen);



Expressions: IN Operator

IN Operator

value IN value_set

- value The value definition or constant to search for in the value_set.
- value_set The set to search.

The **IN** operator is a shorthand for a collection of OR conditions. It searches the value_set to find a match for the value and returns a Boolean TRUE or FALSE.

```
SetSoutheastStates := ['FL','GA','AL','SC'];
BOOLEAN IsSoutheastState(STRING2 state) := state IN SetSoutheastStates;
```



Expressions: BETWEEN Operator

BETWEEN Operator

seekval BETWEEN loval AND hival

- seekval The value to find.
- *loval* The low value in the inclusive range.
- hival The high value in the inclusive range.

The **BETWEEN** operator is shorthand for an inclusive range check using standard comparison operators (*SeekVal* >= *LoVal* AND *SeekVal* <= *HiVal*). It may be combined with NOT to reverse the logic.

```
IsOne2Ten(val) := val BETWEEN 1 AND 10;

// is the passed val in the inclusive range of 1 to 10?
```



More Structures – MODULE

MODULE Structure

```
modulename [ ( parameterlist ) ] := MODULE
  definitions;
END;
```

- modulename The ECL definition name of the module.
- parameterlist The parameters available to all definitions.
- definitions The ECL definitions that comprise the module. These definitions
 may receive parameters, and may include actions (such as OUTPUT).

The **MODULE** structure allows you to pass parameters to a set of related ECL definitions. This is similar to the FUNCTION structure except that there is no RETURN and the *definitions* may be EXPORT or SHARED, making them visible outside the **MODULE** structure.

More Structures – MODULE example

```
EXPORT filterDataset(STRING search, BOOLEAN onlyOldies) := MODULE
 f := namesTable;
  SHARED g := IF (onlyOldies, f(age >= 65), f);
  EXPORT included := g(surname = search);
  EXPORT excluded := g(surname <> search);
END;
//New file -----
filtered := filterDataset('Halliday', true);
OUTPUT(filtered.included,,NAMED('Included'));
OUTPUT(filtered.excluded,,NAMED('Excluded'));
```



Basic Actions

OUTPUT

[name :=] **OUTPUT**(recordset [,format] [,file [,OVERWRITE]]**)**

- name Optional definition name for this action
- recordset The set of records to process
- format The format of the output records: a previously defined RECORD structure, or an "on-the-fly" record layout enclosed in { } braces.
- file Optional name of file to write the records to. If omitted, formatted data stream returns to the command line or ECL IDE program.
- OVERWRITE Allows file to be overwritten if it exists

The **OUTPUT** action writes the *recordset* to the specified *file* in the specified *format*.



OUTPUT Examples:

```
OUTPUT(File_Accounts.File); //Equivalent to: File_Accounts.File;
OUTPUT(Persons,{FirstName,LastName}, NAMED('Names Only'));
OUTPUT(MyRecordset,,'~CLASS::BMF::NewData', OVERWRITE);
//THOR is default format, but you can also output to:
OUTPUT(MyRecordset,,'~CLASS::BMF::NewData', CSV);
OUTPUT(MyRecordset,,'~CLASS::BMF::NewData', XML);
OUTPUT(MyRecordset,,'~CLASS::BMF::NewData', JSON);
```



Aggregate Functions - COUNT

COUNT(recordset)

COUNT(valuelist)

- recordset The set or set of records to process.
- valuelist A comma-delimited list of expressions to count. This may also be a SET of values.

The **COUNT** function returns the number of records in the specified *recordset*.

```
COUNT(Person(per_state IN ['FL','NY']));
TradeCount := COUNT(Trades);
```



Aggregate Functions - MAX

MAX(recordset , value)
MAX(valuelist)

- recordset The set or set of records to process.
- value The field or expression to find the maximum value of.
- valuelist A comma-delimited list of expressions to find the maximum value of.
 This may also be a SET of values

The **MAX** function returns the maximum value of the specified *field* from the specified *recordset*. Returns 0 if the *recordset* is empty.

MaxBal := MAX(Trades, Trades.trd_bal);



Aggregate Functions - MIN

MIN(recordset, value)
MIN(valuelist)

- recordset The set or set of records to process.
- value The field or expression to find the minimum value of.
- valuelist A comma-delimited list of expressions to find the minimum value of.
 This may also be a SET of values

The **MIN** function returns the minimum value of the specified *field* from the specified *recordset*. Returns 0 if the *recordset* is empty.

MinBal := MIN(Trades, Trades.trd_bal);



Aggregate Functions - SUM

```
SUM(recordset , value )
SUM(valuelist)
```

- recordset The set or set of records to process.
- value The expression of field in the recordset to sum.
- valuelist A comma-delimited list of expressions to find the sum of. This may also be a SET of values

The **SUM** function returns the additive sum of the value contained in the specified *field* for each record of the *recordset*. Returns 0 if the *recordset* is empty.

```
SumBal := SUM(Trades, Trades.trd bal);
```



Aggregate Functions - AVE

```
AVE(recordset, value)
AVE(valuelist)
```

- recordset The set or set of records to process.
- value The field or expression to find the average value of.
- valuelist A comma-delimited list of expressions to find the average value of.
 This may also be a SET of values

The **AVE** function returns the average value (arithmetic mean) of the specified *field* from the specified *recordset*. Returns 0 if the *recordset* is empty.

```
AvgBal := AVE(Trades, Trades.trd_bal);
```



Builder Window Runnable Files (BWR)

- Stored in Repository Ready to Run!
- Rules:
 - -File must have at least one action
 - –No EXPORT or SHARED
 - –References to other definitions must be fully qualified:

```
IMPORT $;
Persons := $.File_Persons.File;
Accounts := $.File_Accounts.File;
COUNT(Persons);
COUNT(Accounts);
OUTPUT(Persons,{ID,LastName,FirstName});
```



Lab Demos

Ex1_BWR_BasicPersonsFilters
Ex2_BWR_BasicAccountsFilters

- Basic Filters
- Use of COUNT
- Use of String Indexing
- Use of Logical Operators





Conditional Functions – IF

IF(expression, truevalue, falsevalue**)**

- expression A conditional expression.
- truevalue If the expression is true, this is the value or recordset to return, or an action to perform.
- falsevalue If the expression is false, this is the value or recordset to return, or an action to perform.

The **IF** function evaluates the *expression* (which must be a conditional expression with a Boolean result). If the condition evaluates to true, the *truevalue* is evaluated and returned, otherwise the *falsevalue* is evaluated and returned. Both the *truevalue* and *falsevalue* parameters must be of the same type.

TradeDate := IF(trades.status IN ['A','D','0','1'], Stringlib.GetDateYYYYMMDD, ");



Conditional Functions - MAP

MAP(condition=>val,[condition=>val,... ,] [elseval])

- ✓ condition A conditional expression.
- ✓=> The "results in" operator. Valid only in CASE, MAP and CHOOSESETS.
- √ val the value to return if the condition is true. This may be a single value, recordset, or action.
- ✓ elseval The value to return if all conditions are false. This may be a single value, recordset (optional), or action (optional).

The **MAP** function performs evaluation of the list of *conditions* in sequence. The <u>first</u> *condition* that evaluates as TRUE returns its corresponding *val* as the result of the function. If none of them match, the *elseval* is returned. MAP is an "IF... ELSIF ... ELSE" type of structure. All *val* and *elseval* expressions must be of the same type.



Recordset Functions

EXISTS(recordset)

√ recordset – The set or set of records to process.

The **EXISTS** function returns true if the number of records in the specified *recordset* is > 0. This is commonly used to detect whether a filter has filtered out all the records.

```
PropertyPeople := People(EXISTS(Property));
NoPropertyPeople := People(NOT EXISTS(Property));
```



Conditional Functions – MAP Example

```
RecCode := MAP(
          NOT EXISTS(Trades)
                                              => 99,
          NOT EXISTS(Trades(isValid))
                                             => 98,
          NOT EXISTS(Trades(isValidDate)) => 96,
          -9);
   /* If there are no trades, RecCode gets 99,
      else if there are no valid trades, RecCode gets 98,
     else if there are no valid dates, RecCode gets 96
     else, RecCode = -9
   */
```



Conditional Functions – CASE

CASE(expression, caseval => val,[...] elseval)

- expression An expression that results in a single value.
- caseval A value to compare against the result of the expression.
- => The "results in" operator.
- *val* The value to return a single value, recordset, or action.
- elseval The value to return if the result of the expression does not match any of the caseval values. This may be a single value, recordset, or action.

The **CASE** function evaluates the *expression* and returns the value whose *caseval* matches the expression result, or the *elseval* if no *caseval* matches. All *val* and *elseval* expressions must be of the same type.

NumString123 := CASE(num,1 => 'one',2 => 'two',3 => 'three', 'error');



Conditional Functions – CHOOSE

CHOOSE(expression, value,[..., value,] elseval)

- expression An arithmetic expression that results in a positive integer and determines which value parameter to return.
- value the values to return. There may be as many value parameters as necessary to specify all the expected values of the expression.
- elseval The value to return when the expression returns an out-of-range value. The last parameter is always the elseval.

The **CHOOSE** function evaluates the *expression* and returns the *value* parameter whose ordinal position in the list corresponds to the result of the *expression*, or the *elseval* if none matches. All the *value* and *elseval* expressions must be of the same type.

NumString123 := CHOOSE(num, 'one', 'two', 'three', 'error');



Enterprise Control Language Workshop



Advanced Relational Queries!





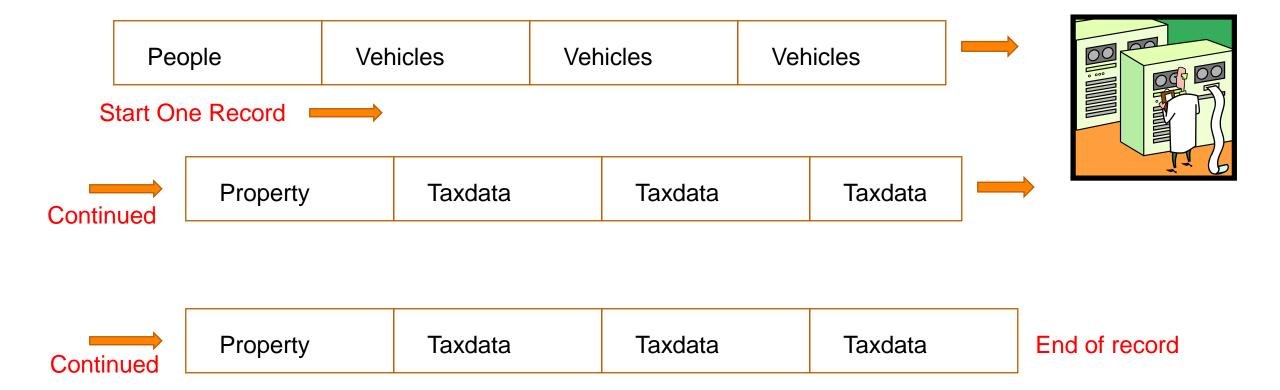
The Relational Data:

Our Workshop Database: (a 3-level hierarchical relational database)

- **≻**People
 - > Vehicle
 - > Property
 - > Taxdata



Denormalizing Related Data:





The Data "Donut" (Normalized Data)





Normalized RECORD and DATASET:

```
EXPORT File_People := MODULE
EXPORT Layout := RECORD
 UNSIGNED8 id;
 STRING15
           firstname;
 STRING25
            lastname;
 STRING15
           middlename;
           namesuffix;
 STRING2
 STRING8
           filedate;
 STRING1
           gender;
 STRING8
            birthdate;
  END;
EXPORT File := DATASET('~Workshop::People',Layout,THOR);
END;
```



Nested Child Dataset RECORD:

```
EXPORT PropTax := RECORD
 $.File_Property.Layout;
 UNSIGNED1 ChildTaxCount;
 DATASET($.File_Taxdata.Layout) TaxRecs{MAXCOUNT(20)};
END;
EXPORT PeopleAll := RECORD
 $.File People.Layout;
 UNSIGNED1 ChildVcount;
 UNSIGNED1 ChildPcount;
 DATASET($.File_Vehicle.Layout) VehicleRecs{MAXCOUNT(20)};
 DATASET(PropTax) PropRecs{MAXCOUNT(20)};
 END;
EXPORT People := DATASET('~Workshop::OUT::PeopleAll',$.Layouts.PeopleAll,FLAT);
```

Working with Child Datasets

- ✓ DATASET fields in RECORD Structures
- **✓**TRANSFORM Function
- ✓ DENORMALIZE Function



ECL TRANSFORM Structure:

```
resulttype funcname( parameterlist ) := TRANSFORM SELF.outfield := transformation; END;
```

- ✓ resulttype The name of a RECORD structure attribute specifying the output format of the function.
- ✓ funcname The name of the function the TRANSFORM structure defines.
- ✓ parameterlist The value types and labels of the parameters that will be passed to the TRANSFORM function.
- ✓ SELF Indicates the resulttype structure.
- ✓ outfield The name of a field in the resulttype structure.
- ✓ transformation An expression specifying how to produce the value assigned to the outfield.



Creating a Relational Dataset

DENORMALIZE(parentoutput,childrecset,condition,transform)

- ✓ parentoutput The set of parent records already formatted as the result of the combination.
- ✓ *childrecset* The set of child records to process.
- ✓ condition An expression that specifies how to match records between the parent and child records.
- ✓ transform The TRANSFORM function to call.

The **DENORMALIZE** function forms flat file records from a parent and any number of children.

The *transform* function must take at least 2 parameters: a LEFT record of the same format as the resulting combined parent and child records, and a RIGHT record of the same format as the *childrecset*. An optional integer COUNTER parameter can be included which indicates the current iteration of child record.



DENORMALIZE Workshop Example:

✓ DENORMALIZE the Data:

Ex3_BWR_DenormalizePeople





Querying Relational Data:

Implicit Dataset Relationality

(Nested child datasets):

- ✓ Parent record fields are always in memory when operating at the level of the Child
- ✓ You may only reference the related set of Child records when operating at the level of the Parent
 - **≻**People
 - **≻** Vehicle
 - **≻**Property
 - >Taxdata



Relational Queries: Property Focused

- 1. Calculate the total number of properties that have, or have *ever* had, 3 or more bedrooms. (Ex4_BWR_PropTaxBeds3)
- 2. Calculate the number of Property records with 3 or more bathrooms, ever. This will be used to calculate number of the specified properties for each person. For the purpose of this definition, the number of "bathrooms" is defined as:

The number of full baths, plus, the rounded number of half baths divided by two. (Ex5_BWR_PropBath_3)

- **≻**People
 - **≻** Vehicle
 - **≻**Property
 - **≻**Taxdata



Queries into action!

✓ Enhancing your Data

Ex6_BWR_ProjectSmallStreet







End of Day 1!







ETL with ECL

The Extract, Transform and Load Process





Extract defined:

Extract involves:

Importing and cleaning raw data from multiple sources.

- In general, the goal of the extraction phase is to convert the data into a usable format which is appropriate for transformation processing.
- An intrinsic part of the extraction involves the parsing of extracted data, resulting in a check if the data meets an expected pattern or structure. If not, the data may be rejected entirely or in part.



Transform defined:

Transform involves:

Combining and collating information from multiple sources.



Mapping and translating source field data into common record layouts, depending on their purpose.

Some examples include:

- Combining multiple records into one (denormalize), or splitting single records into multiple related child records (normalize).
- Translating codes into descriptions and/or vice versa.
- Splitting conjunctive names.
- Reformatting and validating date fields.
- Identifying and appending internal IDs to people, businesses, and other entities, to link them together across datasets.

Transform defined:

Transform cont'd:

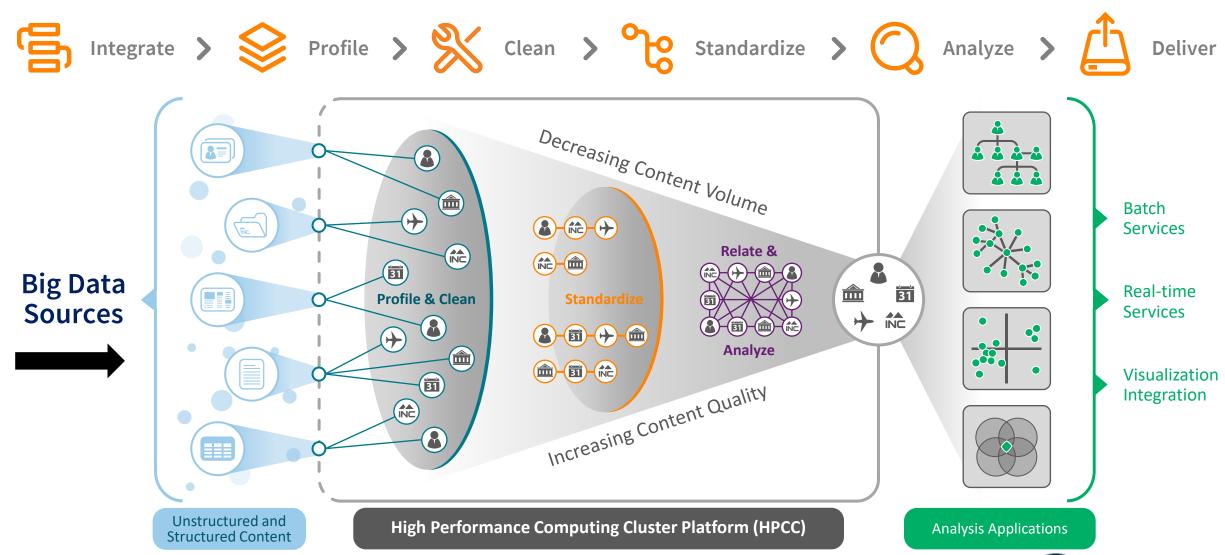


Apply duplication and combination rules to each source dataset and the common build datasets, as required:

- Depending upon type of source, datasets are processed. All new records will simply be added, existing records may have to be updated, or existing records may have to be replaced.
- Duplicates may have to be identified and either excluded or combined with existing data, possibly expanding the valid date range of the data.
- Link records to each other, if applicable, providing time lines of activity or status.

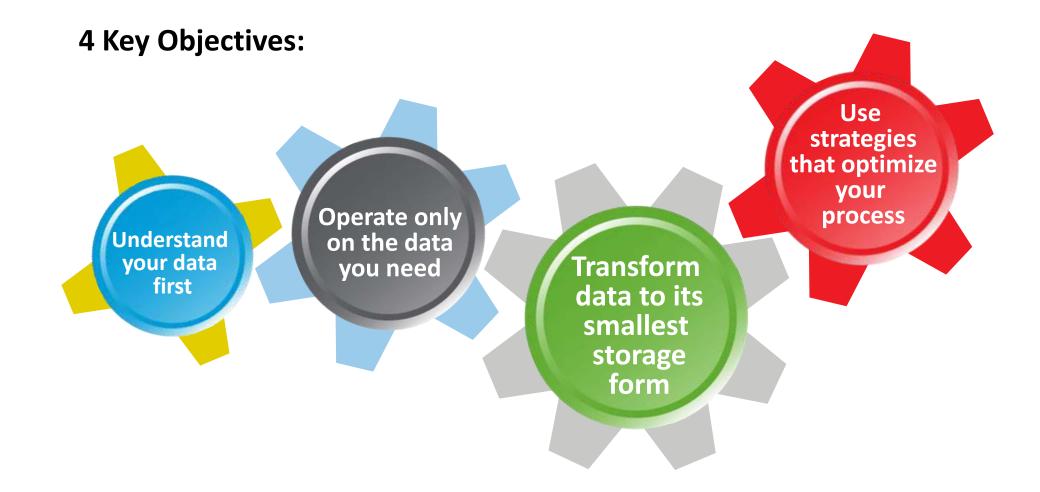


HPCC Systems ETL





ETL in ECL Objectives





Understand your data first

Use Crosstab reports ("grouped by" TABLE) to:

- Get Field population counts
- Determine the Cardinality of key field values (the COUNT per unique value) to discover possible skews
- Get MAX & MIN numeric values to identify actual field value ranges, any out-ofexpected-range values, and other possible issues with the data

Use this information to:

- Develop strategies for even distribution across all nodes
- Develop strategies for data compression and architecture



Operate only on the data that you need:

Assign unique record IDs to all input data

Use a MACRO, PROJECT(dataset, COUNTER), or initial FILEPOS to create them

Clean and standardize the data as appropriate

Standardize names, addresses, dates, times, etc.

Use Vertical projections ("vertical slice" TABLE)

Allows you to work only with the fields you actually need

Use Horizontal projections (record filters)

- Allows you to select only the records relevant to the problem.
- Exclude records with NULL values in key fields, and otherwise filter out bad or irrelevant data unless it can be cleaned.



Transform data to its smallest storage form:

Numeric representation

- UNSIGNED is best, use INTEGER only if signed values are needed
- Select the smallest size appropriate for the range of values

Use Hash values and test for uniqueness

- The COUNT of the DEDUP of the fields to be hashed should equal the COUNT of the hash values
- Use appropriate hash size: 32-bit (HASH, HASH32), 64-bit (HASH64), and 128-bit (HASHMD5)

Use Lookup tables where possible

"Standard" string values can be reduced to representative integers and retrieved at output



Use strategies that optimize your process:

Use ECL Watch as an execution Profiler

- Check the record counts and distribution/skew from step to step.
- Check timing of subgraphs to identify the "hot spots" where execution efficiency can potentially be improved

Keep data de-duped throughout the process, especially after JOINs

- This minimizes any ballooning effect from step to step that will increase processing time and may cause uneven distribution
- Use either DEDUP, ALL [a hash dedup], or SORT then DEDUP

Keep distributions even throughout the process

 If skew develops on one step because of the nature of the data, re-DISTRIBUTE to ensure efficiency on subsequent steps

Use GROUP where possible

- Complex sequences of operations (SORT, DEDUP, ITERATE, ROLLUP, etc.) are more efficient when all data is in memory
- Use GROUP with ALL option or SORT data prior to grouping



More Strategies:

Use LOCAL operations most of the time.

LOCAL can be used on most operations, including GROUP and TABLE

Use LOOKUP or ALL option where possible

 The LOOKUP and ALL options on a JOIN loads all the lookup file records onto each node, so it is implicitly a LOCAL operation

PERSIST intermediate steps/record sets

• This will facilitate both debugging a new process and minimizing rerun times as parts of a process are developed, and provide some recovery capability from system failures

Use OUTPUT(,NAMED) for key intermediate values

These values are stored in the workunit and can be analyzed to identify potential problems



Recordset Functions - SORT

SORT(recordset, value)

- recordset The set of records to process.
- value An expression or key field in the recordset on which to sort. A leading minus sign (-)
 indicates a descending order sort. You may have multiple value parameters to indicate sorts
 within sorts.

The **SORT** function sorts the *recordset* according to the *values* specified. Any number of *value* parameters may be supplied, with the leftmost being the most significant sort criteria. A leading minus sign (-) on any *value* parameter indicates a descending sort for that one parameter.

```
Business_Contacts_Sort := SORT(Business_Contacts_Dist, company_name, company_title, lname, fname, mname, name_suffix, zip, prim_name, prim_range);

HighestBals := SORT(ValidBalTrades, -trades.trd_bal);

UCC_Sort := SORT(UCC_Dist, file_state, orig_filing_num);
```



Lab Demo – SORT Example

Ex7_BWR_SORT_Example





Removing Duplicate Records - DEDUP

DEDUP(recset [,condition [,**ALL**][**BEST** (sort-list)],[,**KEEP** n] [,keep]]**)**

- recset The set of records to process.
- condition The expression that defines "duplicate" records.
- ALL –Matches all records to each other using the condition, not just adjacent records.
- **BEST**-Provides additional control over which records are retained from a set of "duplicate" records. The first in the <sort-list> order of records are retained. BEST cannot be used with a KEEP parameter greater than 1.
- **KEEP** *n* –Specifies keeping *n* number of duplicates. The default is 1.
- keep –LEFT (the default) keeps the first and RIGHT keeps the last.

The **DEDUP** function removes duplicate records from the *recordset*. The *condition* defines what constitutes "duplicate" records.

```
Company_Dedup := DEDUP(Company_Init(zip<>0), company_name,
                         zip, prim_name, prim_range, sec_range, ALL);
```



LEFT and RIGHT Keywords

LEFT.field / RIGHT.field

The **LEFT** and **RIGHT** keywords qualify which record a *field* is from in those operations that use pairs of records, such as DEDUP, JOIN, etc. This helps prevent any code ambiguity for the compiler.



Lab Demo – DEDUP Example

Ex8_BWR_DEDUP_Example





RECORD Field Definitions:

```
datatype identifier [ := defaultvalue] [ { modifier } ];
identifier := defaultvalue;
defaultvalue;
sourcefield;
recstruct [ identifier ];
sourcedataset;
childdataset identifier [ { modifier } ];
```

- datatype The value type of the data field.
- identifier The name of the field.
- defaultvalue Optional. An expression defining the source of the data.
- modifier Optional. One of the keywords listed in the Field Modifiers.
- sourcefield The name of a previously defined data field, which implicitly provides the datatype, identifier, and defaultvalue for the new field—all inherited from the existing field.
- recstruct The name of a previously defined RECORD structure.
- sourcedataset The name of a previously defined DATASET or derived recordset definition. See the **Field**Inheritance section in the LRM.
- childdataset A child DATASET declaration.



Example RECORD structures:

```
R1 := RECORD
  Persons.LastName;
 //field name is LastName - size is as declared in the persons dataset
  STRING25 LocalID := Persons.FirstName;
 //LocalID field gets its data from Persons.FirstName
  INTEGER8 COUNT(Trades); //this field is unnamed in the output file
  BOOLEAN IsBogey := FALSE; //IsBogey defaults to false
END;
R2 := RECORD
  Trades;
  //include all fields in the Trades dataset at already-defined names, types and sizes
  UNSIGNED8 fpos {virtual(fileposition)}; //the relative byte position within the file
END;
```



TABLE Function:

TABLE(recordset, format [.expression [.FEW|MANY] [. UNSORTED]] [.LOCAL][.KEYED])

- recordset The set of records to process.
- format The RECORD structure of the output records.
- expression The "group by" clause for crosstab reports. Multiple comma-delimited expressions create one logical "group by" clause.
- FEW Indicates that the *expression* will result in fewer than 10,000 distinct groups.
- MANY Indicates that the expression will result in many distinct groups.
- UNSORTED Indicates you don't care about the order of the groups.
- LOCAL Specifies independent node operation.
- KEYED Specifies activity is part of an index read operation

The **TABLE** function is similar to OUTPUT, but instead of writing records to a file, it outputs those records into a new memory table (a new dataset in the supercomputer). The new table inherits any implicit relationality the *recordset* has unless an *expression* is present. The new table is temporary, and exists only while the query is running.



TABLE example (vertical slice):

```
//"vertical slice" TABLE:
Layout Name State := RECORD
  Persons.LastName;
  Persons.FirstName;
  Persons.State;
  END;
Per_Name_State := TABLE(Persons, Layout Name State);
```



String functions: LENGTH

LENGTH(expression)

expression – A string expression.

The **LENGTH** function returns the length of the string resulting from the expression.

INTEGER1 StrCnt := LENGTH('ABC' + 'XYZ'); // Result is 6



String Function: TRIM

TRIM(string*value* [, *flag*])

- *stringvalue* The string from which to remove spaces.
- flag Optional. Specifies which spaces to remove. RIGHT removes trailing spaces (this is the default). LEFT removes leading spaces. RIGHT, LEFT removes both leading and trailing spaces. ALL removes all spaces.

The **TRIM** function returns the *stringvalue* with all trailing and/or leading spaces removed.

STRING20 StrVal := ' ABC'; // Contains 3 leading, 14 trailing spaces

VARSTRING StrVal1 := **TRIM**(StrVal, ALL); // Contains 3 characters



CROSSTAB Example:

Ex9_BWR_Crosstab_Example

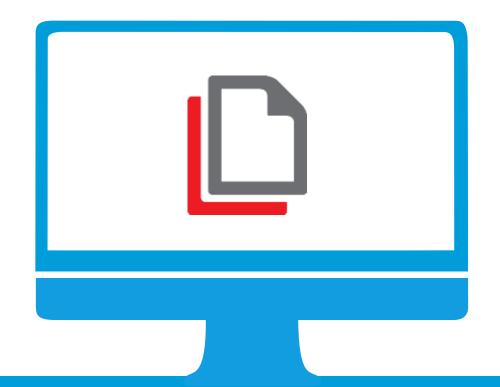




Lab Exercises:

Crosstab Reports Examples

Ex10_BWR_CT_Gender Ex11_BWR_CT_HighCredit





GROUP Keyword:

The **GROUP** keyword replaces the *recordset* parameter of any aggregate function used in the record structure of a TABLE definition where a group by *expression* is present.

This is only used to generate a crosstab report (set of statistics) on a recordset. There is also a GROUP function which provides similar functionality.

```
// Create a crosstab report for each sex in each state
R := RECORD
Persons.State;
Persons.Gender;
COUNT(GROUP);
COUNT(GROUP,Persons.Gender = 'M') //Scoping count
END;
CTOut := TABLE(Persons,R, State, Gender);
```



INLINE DATASET

name := DATASET(recordset, recorddef)

- name The definition name by which the file is subsequently referenced.
- recordset A set of data records contained within square brackets (indicating a set definition). Within the square brackets, each record is delimited by curly braces ({}) and separated by commas. The fields within each record are comma delimited.
- recorddef The RECORD structure of the dataset.

DATASET introduces a new table into the system with the specified *recorddef* layout. This form allows you to treat an inline set of data as a data file.

```
NamesRec := RECORD
     STRING20 first_name;
     STRING20 last_name;
END;
Names := DATASET([{'John','Jones'}, {'Jane','Smith'}], NamesRec);
```



Lab Demo

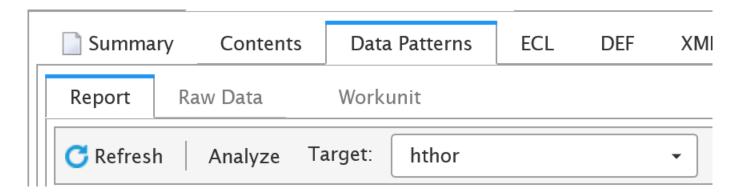
Ex12_BWR_PopulationCount





Built-In Data Profiling: Data Patterns

An extensive Data Profiling report is now built-in and available in the ECL Watch for all logical files. This report can be accessed via the Data Patterns tab:



There are three ways to use Data Patterns:

- ECL Watch (via the Data Patterns tab)
- Bundle (found on the HPCC Git Hub): https://github.com/hpcc-systems/DataPatterns.git
- Standard Library Reference (STD.DataPatterns)

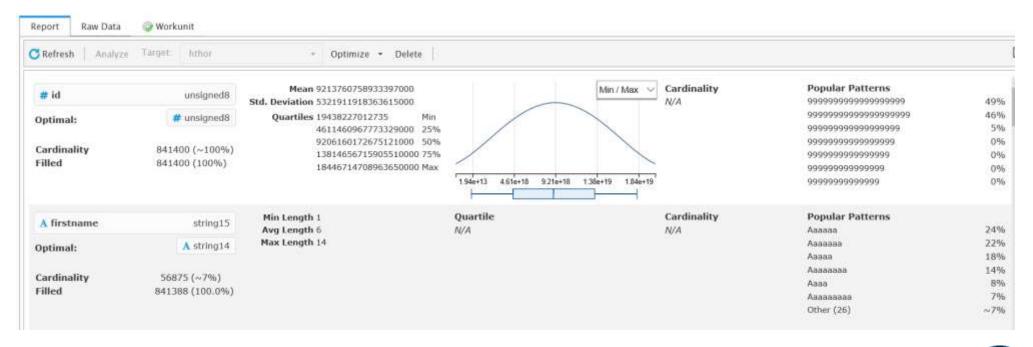


Built-In Data Profiling: Data Patterns

```
filePath := '~aaa::bmf::reptest::accounts';
ds := DATASET(filePath, RECORDOF(filePath, LOOKUP), csv);
profileResults := DataPatterns.Profile(ds);
OUTPUT(profileResults, ALL, NAMED('profileResults'));
```

Requirements (for ECL Watch Analysis);

- File must have an ECL RECORD
- 2. Size of the file should not be too large (Use SAMPLE for larger files.





IMPORT STD.DataPatterns;

Built-In Visualization Tools

HPCC Systems provides built-in Visualization of your output data in a variety of charts and graphs. You can visualize your data in three ways:

- Using the Chart Tool in the ECL Playground.
- Accessing the Visualize tab in all ECL workunits
- Using the Resources tab in conjunction with the ECL Visualizer bundle.

```
Installing:
```

```
ecl bundle install https://github.com/hpcc-systems/Visualizer.git
```



Visualization Function Summary

Global or Helpers (Any MODULE):

Meta

Creates a special output record set that contains the meta information for use in the target visualization. Outputs visualization meta information.

Grid

Used with the **Meta** function to render data into the appropriate data grid or table. Mappings can be used to limit and/or rename the columns.

Two Dimensional Ordinal (TwoD MODULE):

Bubble - a series of circles whose size is proportional to the field's value

Pie - a single circle divided into proportional slices

Summary - values are displayed in designated intervals

RadialBar - basically a bar chart plotted on polar coordinates instead of a Cartesian plane

WordCloud - a visual representation of text data, whose size is proportional to its value.



Visualization Function Summary

Two Dimensional Linear (**TwoDLinear** MODULE):

Scatter - uses Cartesian (X/Y) coordinates to display dot or point values for two numeric fields in a dataset **HexBin** - useful to represent 2 numerical fields when you have a lot of data points. Instead of overlapping, the plotting window is split into several hexbins, and the number of points per hexbin is counted. The color denotes this number of points.

Contour - A graphical technique that uses contour lines. A contour line of a function of two variables is a curve along which the function has a constant value, so that the curve joins points of equal value.

Multi-Dimensional (MultiD MODULE):

Area - Label is plotted on X-axis, and values for each label are plotted on the Y-Axis. Common fields are joined by line and area below is shaded.

Bar - Label is plotted on Y-axis, and values for each charted by bar on the X-Axis.

Column - Similar to Bar, but Label is plotted on the X-Axis, and associated values charted by bar on the Y-Axis.

Line - Similar to Area, but no shaded area is colored (lines only)

Radar - Similar to the Area plotting, but uses a center point in a circle for the X and Y axis zero coordinate. Think of a radar scope for this graph type.

Scatter - Points are marked only on this graph type

Step - Uses vertical and horizontal lines to connect data points, resembling a step ladder type of display.



Visualization Function Summary

Relational (Relational MODULE):

Network - assembles the vertices and edges defined into an entity relation chart

Geospatial (Choropleth MODULE):

USStates - A US States Map that maps a two letter State code with its associated value.

USCounties - A US County Map that maps a numeric FIPS (Federal Information Processing Standard) county code with its associated value.

Euro - a base map graph that allows you to select any country in the European continent. A two letter international code is required for the region.

EurolE - Example function in the bundle that uses the **Euro** function to show the country map of Ireland.

EuroGB - Example function in the bundle that uses the **Euro** function to show the country map of Great Britain.

Each of these categories have an associated ___Test function that can be used to view all graphs/charts/maps in that category.

```
//Example:
IMPORT Visualizer;
Visualizer.Choropleth. test;
```



Visualization Quick Start

Using the Visualization Bundle is essentially a three-step process.

- 1. Get your data ready. Too many points can make a graph unreadable, too little points diminishes the analysis benefit.
- 2. Import the Visualizer folder. If you use the "git" method, your folder will be easily located.
- 3. OUTPUT your data using the NAMED attribute.
- 4. Call your Visualizer chart using the NAMED attribute as your chart source

Example:



Lab Demo Examples

- Ex13_BWR_DataPatterns.ecl
- Ex14_BWR_Visualization.ecl





Data Transformation Tools

- PROJECT Function
- ITERATE Function
- PERSIST Workflow Service



PROJECT Function

PROJECT(recordset, transform)

- recordset The set of records to process.
- transform The TRANSFORM function to call for each record in the recordset.

The **PROJECT** function processes through all the records in the *recordset* performing the *transform* function on each record in turn.

The *transform* function must take at least one parameter: a **LEFT** record of the same format as the recordset. It may take an optional second parameter: an integer COUNTER specifying the number of times the transform has been called. The resulting record format can be different from the input.



PROJECT Example:

```
YP_SlimRec := RECORD
  INTEGER
                seq;
  STRING125
                business_name;
  STRING100
                orig_street;
                orig_city;
  STRING75
  STRING2
                orig state;
  STRING9
                orig_zip;
                orig_phone10;
  STRING10
END;
YP_SlimRec SlimYp(YellowPagesRec Le,INTEGER cnt) := TRANSFORM
  SELF.seq := cnt;
  SELF := Le;
END;
YP_Slim := PROJECT(File_YellowPages, SlimYP(LEFT, COUNTER));
```

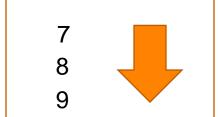


LOCAL Sequencing

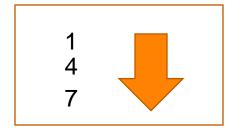


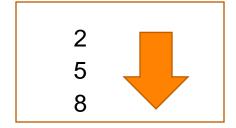


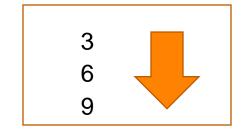




PROJECT with LOCAL









Standard Library Reference

The Standard Library Reference includes utility functions for string handling, date manipulation and others.

To implement, simply IMPORT the base library folder: STD

IMPORT STD;

Includes functions for:

Logical Files Superfiles External Files

File Movement String Handling Metaphone Support Cryptography Support

Date/Time Handling Cluster Handling Job Handling File Monitoring

Logging Auditing Utilities Debugging

Email Workunit Services BLAS Support Math Support

Data Patterns



File Browsing

Lab Demo: Record Sequencing

Ex15_BWR_LocalRecordSequencing.ecl





ITERATE Function

ITERATE(recordset, transform [, LOCAL])

- recordset The set of records to process.
- transform The TRANSFORM function to call.
- •LOCAL Optional. Specifies the operation is performed on each HPCC node independently, without requiring interaction with all other nodes to acquire data; the operation maintains the distribution of any previous DISTRIBUTE.

The **ITERATE** function processes sequentially through all records in a *recordset* one pair of records at a time, performing the *transform* function on each pair in turn. The LEFT record passed to the *transform* is always the result record from the previous iteration, making the first left record always all blanks and/or zeroes.

The *transform* function must take at least 2 parameters: a **LEFT** record and a **RIGHT** record that must both be of the same format as the resulting *recordset*. An optional third parameter may be specified: an integer COUNTER specifying the number of times the *transform* has been called.



ITERATE Example:

```
Testfile Match Group := GROUP(Testfile Match Sort, id, LOCAL);
Layout_Match AssignSequence(Layout_Match L, Layout_Match R) := TRANSFORM
  SELF.sequence := L.sequence + 1;
  SELF.match_indicator := 'M';
  SELF := R;
END;
Testfile Match Seq := ITERATE(Testfile Match Group, AssignSequence(LEFT, RIGHT));
```



ITERATE Demo:

Ex16_ITERATE_Books_Example





PERSIST Workflow Service:

PERSIST

name := exp : PERSIST(file [,cluster] [,CLUSTER(target)] [,EXPIRE(days)]);

- name The definition name.
- exp The definition expression.
- file A string constant defining the logical filename in which the results of the definition's expression are stored.
- cluster A string constant specifying the name of the cluster on which to re-build the name if/when necessary.
- CLUSTER Specifies writing the file to another cluster.
- target The cluster to write the file to.

The **PERSIST** service stores the result of the *expression* in the specified *file* so it remains permanently available for use. This is particularly useful for recordset definitions that perform large, expensive data manipulation sequences.

Large processes can be segmented using PERSIST to both facilitate automation and minimize re-run time. PERSISTed definitions are re-computed only when the code or data used to compute them have changed.



DATA Standardization

- MODULE Structure
- Transforming data using TABLEs
- Explicit CASTING
- Measuring size of RECORDs and Fields



SIZEOF Function:

SIZEOF(data)

• data – The name of a dataset, RECORD structure, a fully-qualified field name, or a constant string expression.

The **SIZEOF** function returns the total number of bytes defined for storage of the specified *data* structure or field.

```
MyRec := RECORD
INTEGER5
                 F2;
 QSTRING12
                 F3;
VARSTRING12
                 F4;
END;
MyData := DATASET([\{1,33333333333, 'A',v'A'\}],MyRec);
SIZEOF(MyRec);
                         //result is 27
                         //result is 27
SIZEOF(MyData);
SIZEOF(MyData.F2); //result is 5
SIZEOF(MyData.F3);
                        //result is 9 -12 chars stored in 9 bytes
SIZEOF(MyData.F4);
                         //result is 13 -12 chars plus null terminator
SIZEOF('abc' + '123');
                         //result is 6
```



Data Standardization Issues:

Data Standardization: transform using a TABLE!

- ➤ Wrap TABLE in MODULE, define Layout and File
- Do for both Persons (STD_Persons) and Accounts (STD_Accounts)
- Default values in Layout can be used as transformation expressions: STRING15 FirstName := std.Str.ToUpperCase(\$.UID_Persons.FirstName);
- Dates
 STRING8 to UNSIGNED4
- Names
 Convert all to Upper Case
- Zip STRING5 to UNSIGNED3
- ➤ What about the Middle Name?
- What about MaritalStatus and Dependent Count?



Lab Demo

STD_Persons.ecl

Data Standardization

Using "vertical slice" TABLEs





Creating Lookup Tables:

- **ROLLUP Function**
- **MODULE Structure**
- OUTPUT to file and DATASET



ROLLUP Function:

ROLLUP(recordset, condition, transform)

- recordset The set of records to process.
- condition An expression that specifies which records are considered to be duplicates.
- transform The TRANSFORM function to call for each pair of duplicate records in the recordset.

The **ROLLUP** function is similar to a DEDUP (with the KEEP 1 LEFT or RIGHT options) with the addition of a call to the transform function, allowing you to salvage data from the duplicate records before discarding them.

The transform function must take at least two parameters: a LEFT record and RIGHT record, which must be of the same formats. The format of the resulting record set must be the same as the inputs.



ROLLUP Example

```
TempCustRec RollupRevenue(TempCustRec Le, TempCustRec Ri) := TRANSFORM
 SELF.revenue_total := Le.revenue_total + Ri.revenue_total;
 SELF := Le;
END;
SortedCustRecs := SORT(CustomerRecs, bid, SubsidiaryCode, BUCode);
TotalRevenue :=
  ROLLUP(SortedCustRecs,
      LEFT. bid
                         = RIGHT. bid AND
      LEFT.SubSidiaryCode = RIGHT.SubSidiaryCode AND
      LEFT.BUCode = RIGHT.BUCode,
      RollupRevenue(LEFT,RIGHT));
```



Lab Demo

EX17_ROLLUP_CSZ.ecl File_LookupCSZ.ecl





JOIN Function

JOIN(*leftset*, *rightset*, *condition*[, *transform*] [,*type*] [,*flag*])

- *leftset* The LEFT set of records to process. This should be the larger of the two files.
- rightset The RIGHT set of records to process.
- condition The expression that specifies how to match records between the leftset and rightset.
- transform –The TRANSFORM function to call.
- type The type of join to perform (default is an inner join).
- flag Any option to specify exactly how the JOIN operation executes.

The JOIN function processes through the records in the *leftset* and *rightset*, evaluating the *condition* to find matching records. The *transform* function executes on each pair of matching records.

The transform function must take at least 2 parameters: a LEFT record formatted like the leftset, and a RIGHT record formatted like the rightset. These may be different formats, and the resulting record set can be a different format.

JOIN Types

INNER – All matching records from both record sets.

LEFT OUTER – At least one record for every record in the *leftset*.

RIGHT OUTER – At least one record for every record in the *rightset*.

FULL OUTER – At least one record for every record in both the *leftset* and *rightset*.

LEFT ONLY – One record for every record in the *leftset* for which there is no matching record in the *rightset*.

RIGHT ONLY – One record for every record in the *rightset* for which there is no matching record in the *leftset*.

FULL ONLY – One record for every record in the *leftset* and *rightset* for which there is no matching record in the opposite record set.



JOIN Flags

[MANY] LOOKUP – Copies *rightset* to each node for a Many-0/1 (without MANY) or Many-0/Many lookup.

GROUPED – Copies *rightset* to each node for a Many-0/Many lookup, preserving grouping (Roxie-specific).

ALL – Copies *rightset* to each node for a Many-0/Many lookup – no equality portion of *condition* required.

SMART – Specifies to use an in-memory lookup when possible, but use a distributed join if the right dataset is large.



JOIN Example:

```
Layout_Comp_Temp EnhancePhones(Layout_Comp_Temp Le, Layout_Bus_Header Ri) := TRANSFORM
  SELF.phone_score := Ri.phone_score;
  SELF.phone := IF( (INTEGER)Ri.phone <> 0, Ri.phone, Le.phone);
  SELF := Le;
END;
Company_Enh := JOIN(Company_Seq_Dist,
                     Phone_Business_Dist,
                     LEFT.zip = RIGHT.zip AND
                     LEFT.address = RIGHT. address AND
                     STD.CompSimilar100(LEFT.co_name,RIGHT.co_name) <= 35,</pre>
                     EnhancePhones(LEFT, RIGHT),
                     LEFT OUTER);
```



Lab Demo:

EX18_BWR_File_Persons_Slim.ecl







End of Day 2!







ROXIE ECL Queries

Standard and Payload Indexes





Introduction:

- ✓ Simple Indexing
 - ✓ Defining and Building
 - ✓ Using FETCH
- ✓ Payload Indexing
 - ✓ Defining and Building
 - ✓ Full and half-keyed JOINs



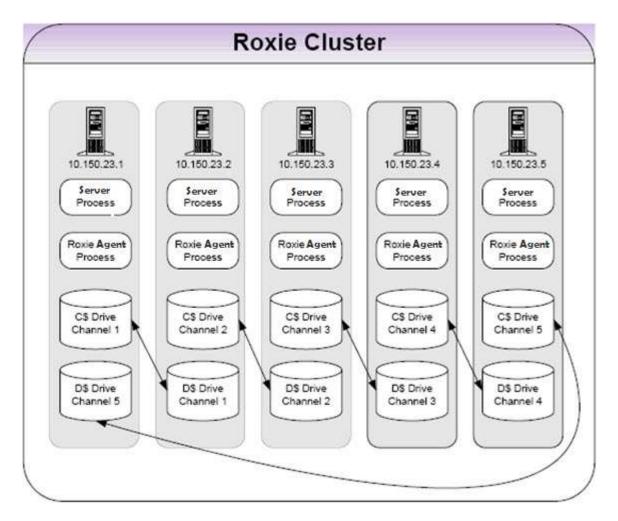
What is ROXIE?

ROXIE is also known as the HPCC "Rapid Data Delivery Engine".

It is a number of machines connected together that function as a single entity running:

Server (Farmer) processes Agent (Slave) processes

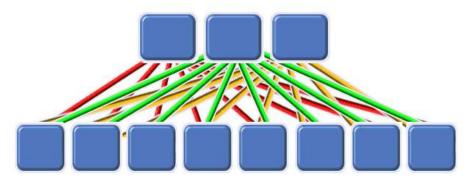
Some special configurations have Server Only or Agent Only nodes.



This example shows a 5-node ROXIE



What is ROXIE? (cont.)



ROXIE is a massively parallel query processing supercomputer with:

- Structured Query Engine:
 - Uses disk-based indexed datasets.
 - Uses pre-compiled repeatable queries.
- Scalable retrieval engine:
 - High throughput.
 - Millions of users receiving sub-second responses.
- Built-in redundancy:
 - Copies of data are stored on two or more nodes.
 - Any peer can deliver a task for a channel.
 - High availability ROXIE cluster continues to perform even if one or more nodes fail.



How ROXIE works

Each machine runs a Server process and an Agent process:

- The Server process:
 - Accepts queries and returns results.
 - Combines results from multiple agents.
- The Agent process:
 - Stores the data.
 - Delivers selected data to the Server on demand.

The Server process waits for a request.

When a request is received:

- The Server determines which Agent nodes have the appropriate data using the metakey.
- The Server sends a portion of the query to those Agents and waits for results.
- When all results are received, the Server:
 - Collates the results.
 - Performs any additional processing.
 - Returns the result set to the requestor.



ECL and ROXIE

INDEX, BUILD and FETCH



INDEX - Standard

INDEX([base,] def, file [options])

- ✓ base The recordset for which the index file was created.
- ✓ def The RECORD structure of the indexfile. Contains key and file position information for referencing into the baserecordset. Field names and types must match baserecordset field names.
- ✓ file A string constant containing the logical filename of the index file created by BUILD.
- ✓ options SORTED, PRELOAD, COMPRESSED, DISTRIBUTED.

INDEX defines an index file.



INDEX (cont.) - PAYLOAD

INDEX([base,] def, [payload,] file [options])

- ✓ base The recordset for which the index file was created.
- ✓ def The RECORD structure of the fields in the *file* that contains key and file position information for referencing into the base.
- √ payload The RECORD structure that contains additional fields not used as key fields.
- ✓ file A string constant containing the logical filename of the index file created by BUILD.
- ✓ options SORTED, PRELOAD, COMPRESSED, DISTRIBUTED.

This **INDEX** adds a payload:



Payload and non-payload indexes

Payload indexes contain additional fields as well as the searchable ones.

Non-payload indexes only contain the searchable fields.

The index maps a searchable field into one or more results fields. In the simplest cases, all fields are searchable and the result is an implied " Is it there or not? ".

In another case, all fields may be searchable apart from a file offset into a raw data file where the result is to be found. For performance reasons it is often preferable to copy that data into the non-searchable fields of the index so they can be retrieved using a *single* fetch rather than *two*.

Payload indexes provide better performance but at the expense of disk space. The result is duplicated in every index rather than shared between them.



KEYED and WILD

KEYED (expression) WILD (field)

The **KEYED** and **WILD** keywords are **valid only in INDEX filters** (which also qualifies as part of the *joincondition* for a "half-keyed" JOIN). They indicate to the compiler which of the leading index fields are used as filters (KEYED) or wild carded (WILD) so that the compiler can warn you if you've gotten it wrong. Trailing fields not used in the filter are ignored (always treated as wildcards).



BUILD – Standard (Access) INDEX

BUILD(baserecordset, [indexdef], indexfile)

- ✓ baserecordset The base recordset for which the index file will be created or a recordset derived from the base recordset with the key fields and file position.
- ✓ indexdef The RECORD structure of the indexfile. Contains key and file position information for referencing the recordset. Field names and types must match baserecordset field names.
- ✓ indexfile A string constant containing the logical filename of the index file.

The **BUILD** (or BUILDINDEX) action creates an index file for use with a *baserecordset*.

BUILD(ssn_data,{did,RecPos},'key::header.ssn.did');



BUILD – Payload INDEX

BUILD(baserecset, keys, payload, indexfile [, options])

- ✓ baserecordset The base recordset for which the index file will be created or a recordset derived from the base recordset with the key fields and file position.
- ✓ keys The RECORD structure of fields in the indexfile that contains key and file position information for referencing into the baserecset.
- ✓ payload The RECORD structure of the indexfile that contains additional fields not used as keys.
- ✓ indexfile A string constant containing the logical filename of the index file.

The **BUILD** (or BUILDINDEX) Form 2 action creates an index file containing extra payload fields in addition to the keys.

BUILD(Vehicles, {st, city}, {Iname}, 'vkey::st.city');



BUILD (from INDEX definition)

BUILD(indexdef [, options]**)**

✓ indexdef – The name of the INDEX attribute to build.

The **BUILD** (or BUILDINDEX) Form 3 action creates an index file by using a previously defined INDEX definition.

nameKey := **INDEX**(mainTable,{surname,forename,filepos},'name.idx'); **BUILD**(nameKey); //gets all info from the INDEX definition



FETCH(baserecset, indexfile, positionfield [, transform] **)**

- ✓ baserecset The base set of records from which the indexfile was created.
- ✓ indexfile An index on the baserecset, usually filtered.
- ✓ positionfield The field in the indexfile (RIGHT.fieldname) containing the record pointer into the *baserecset*.
- ✓ transform The TRANSFORM function to call.

The **FETCH** function evaluates the *indexfile* to determine the key(s) to use to access the indexfile, which in turn provides the positionfield values of the corresponding record(s) in the baserecset. All matching records from the baserecset are returned.

The transform function must take at least one parameter: a LEFT record of the same format as the baserecset.



FETCH Example:

```
basefile := file_header_plus;
indexfile := key_header_ssn;
TYPEOF(basefile) FetchHeader(basefile Le) := TRANSFORM
SELF := Le;
END;
EXPORT Fetch_Header_SSN(STRING9 rssn) := FETCH(basefile,
                                                   indexfile(ssn=rssn),
                                                   RIGHT.fpos,
                                                   FetchHeader(LEFT));
```



FETCH Demo Example:

Ex19_BWR_FETCH_Example.ecl

✓ Change the DataFile and KeyFile definitions to begin with YOUR INITIALS.





FUNCTION Structure

```
[resulttype] funcname( parameterlist ) := FUNCTION
  code;
  RETURN returnvalue;
END;
```

- ✓ resulttype The return value type of the function. If omitted, the type is implicit from the returnvalue expression.
- √ funcname The ECL attribute name of the function.
- ✓ parameterlist The parameters to pass to the code available to all attributes defined in the FUNCTION's code.
- ✓ code The attributes that define the FUNCTION's process.
- ✓ RETURN Specifies the function's returnvalue expression.
- ✓ returnvalue The value, expression, recordset, row (record), or action to return.

The **FUNCTION** structure allows you to pass parameters to a set of related attribute definitions. This makes it possible to pass parameters to an attribute that is defined in terms of other non-exported attributes without the need to parameterize all of those as well.

Without FUNCTION Example:

```
isTradeRateEQSince(STRING1 rate,age) := Trades.trd_rate = rate AND
                                        ValidDate(trades.trd drpt) AND
                                        AgeOf(trades.trd_drpt) <= age;
isPHRRateEQSince(STRING1 rate, age) := EXISTS(Prev_rate(phr_rate = rate,
                                              ValidDate(phr_date),
                                              AgeOf(phr date) <= age,
                                              phr_grid_flag = TRUE));
isAnyRateEQSince(STRING1 rate,age) := isTradeRateEQSince(rate,age)
                                      OR
                                      isPHRRateEQSince(rate,age);
```



With FUNCTION Example

EXPORT is AnyRate EQSince (STRING1 rate, age) := **FUNCTION**

RETURN isTradeRateEQSince OR isPHRRateEQSince; **END**;



STORED

[name :=] expression : **STORED**(storedname);

- √ name The definition name.
- ✓ expression The definition of the name.
- ✓ storedname A string constant defining the logical name in the workunit in which the results of the attribute expression are stored. If omitted, the result is stored using the definition name.

The **STORED** service stores the result of the expression in the workunit so that it remains available for use throughout the workunit. If the *name* is omitted, then the stored value can only be accessed after completion of the workunit. If a *name* is provided, the value of the definition will be pulled from storage, and if it has not yet been set, will be computed.

Adding the STORED workflow service to any definition makes that definition transparent to anyone accessing the ROXIE query.



STORED Example:

```
STRING30 fname_value := ": STORED('FirstName');

STRING30 lname_value := ": STORED('LastName');

STRING2 state_value := ": STORED('State');

STRING1 sex_value := ": STORED('Sex');

STRING ToSearch := ": STORED('SearchText');
```



ROXIE Demos:

EX20_BWR_TestRoxieFunction PeopleFileSearchService





ECL and ROXIE

HALF and FULL KEYED JOINS
The heart of a ROXIE query!



JOIN Function

JOIN(*leftset*, *rightset*, *condition*[, *transform*] [,*type*] [,*flag*])

- *leftset* The LEFT set of records to process. This should be the larger of the two files.
- rightset The RIGHT set of records to process.
- condition The expression that specifies how to match records between the leftset and rightset.
- transform –The TRANSFORM function to call.
- type The type of join to perform (default is an inner join).
- flag Any option to specify exactly how the JOIN operation executes.

The JOIN function processes through the records in the *leftset* and *rightset*, evaluating the *condition* to find matching records. The *transform* function executes on each pair of matching records.

The transform function must take at least 2 parameters: a LEFT record formatted like the leftset, and a RIGHT record formatted like the rightset. These may be different formats, and the resulting record set can be a different format.

ROXIE JOIN Flags

KEYED(*idx***)** – Uses *idx* into *rightset* to do the JOIN. The *idx* must contain the virtual(fileposition) rec pointer.

ATMOST(exp, n) – Eliminates matches where exp is true and > n result records.

LIMIT(n [,SKIP]) – Eliminates matches with > n result records. Fails the job unless SKIP is present.



FULL KEYED JOIN:

- A "full-keyed" JOIN uses the KEYED(index) option
- The joincondition must be based on key fields in the index.
- The join is actually done between the leftrecset and the index into the rightrecset
- The *index* needs the dataset's record pointer (virtual(fileposition)) field to properly fetch records from the *rightrecset*.
- The typical KEYED join passes only the rightrecset to the TRANSFORM.



JOIN Example – FULL KEYED:

```
Layout Joined Xform(LeftDS Le, RightDS Ri) := TRANSFORM
 SELF := Ri;
 SELF := Le;
END;
FullKey := JOIN(LeftDS, RightDS,
               LEFT.keyfield = RIGHT.keyfield,
               Xform(LEFT, RIGHT),
               KEYED(RightDS Idx)); //KEYED(Index) usage
```



HALF KEYED JOIN:

- The rightrecset is an INDEX
- Usually, the INDEX in a "half-keyed" JOIN contains "payload" fields, which frequently eliminates the need to read the base dataset.
- The "payload" INDEX does not need to have the dataset's record pointer (virtual(fileposition)) field declared.
- The joincondition may use the KEYED and WILD keywords that are available for use in INDEX filters, only.



JOIN Example – HALF KEYED:

```
HalfKey := JOIN(LeftDS, RightDS_PayloadIDX,

KEYED(LEFT.keyfield = RIGHT.keyfield),

TRANSFORM(SELF := LEFT,

SELF := RIGHT));

//KEYED(Join Condition) usage
```



Why FULL and HALF?

•In SQL if you do a join across two files you end up with a VIEW which has columns from both files — the *indexing* piece is done in two stages — one stage looks up the data from file 1 in an index — the second looks up the index result in the second file to produce the result. This is a FULL-keyed join

•In the HALF-keyed join we skip the second step — and return results based upon file one and the index *only*.



ROXIE Full/Half Keyed Demos:

EX21_BWR_FullKeyed
EX22_BWR_HalfKeyedJoin
HalfKeyedSearchService







End of Day 2!







Introduction to Machine Learning

Concepts and Terminology





What is Machine Learning?

The broad definition of Machine Learning (from Wikipedia) "the scientific study of algorithms and statistical models that computer systems use to perform a specific task without using explicit instructions, relying on patterns and inference instead. It is seen as a subset of artificial intelligence.

The "learning" part of Machine Learning (or ML) has several categories, here is what we currently showcase in ECL:

- Supervised The most common type of ML. This method involves the training of the system where the
 recordsets along with the target output pattern is provided to the system for performing a task.
- Unsupervised This method does not involve the target output which means no training is provided to the system. The system has to learn by its own through determining and adapting according to the structural characteristics in the input patterns.
- **Deep** Moves into the area of the ML Neural Networks methods. Deep Learning implies multiple layers greater than two (2). Deep also implies techniques use with complex data, like video or audio analysis.



Machine Learning Basics

Let's examine Supervised learning:

Given a set of data samples:

```
Record1: Field1, Field2, Field3, ..., FieldM
Record2: Field1, Field2, Field3, ..., FieldM
RecordN: Field1, Field2, Field3, ..., FieldM
```

And a set of target values,

```
Record1: TargetValue
Record2: TargetValue
RecordN: TargetValue
```

"Independent" Variables

Note: The fields in the independent data are also known as "features" of the data

"Dependent" Variables

Learn how to predict target values for new samples.

- The set of Independent and Dependent data is known as the "Training Set"
- The encapsulated learning is known as the "Model"
- Each model represents a "Hypothesis" regarding the relationship of the Independent to Dependent variables.
- The hallmark of machine learning is the ability to "Generalize"



Machine Learning Example

Given the following data about trees in a forest:

Height	Diameter	Altitude	Rainfall	Age	
50	8	5000	12	80	
56	9	4400	10	75	
72	12	6500	18	60	
47	10	5200	14	53	

- Learn a Model that approximates Age (the Dependent Variable) from Height, Diameter, Altitude, and Rainfall (the Independent Variables).
- It is hard to see from the data how to construct such a model
- Machine Learning will automatically construct a model that (usually) minimizes "prediction error".
- In the process, depending on the algorithm, it may also provide insight into the relationships between the independent and dependent variables (inference).
- Note: We normally want to map easy to measure ("observed") features to hard-to-measure ("unobserved") features.



Quantitative and Qualitative Models

- Supervised Machine Learning supports two major types of model:
 - Quantitative e.g., Determine the numerical age of a tree
 - Qualitative e.g., Determine the species of the tree, Determine if the tree is healthy or not.
- Learning a Quantitative model is called "Regression" (a term with archaic origins don't try to make sense of it).
- Learning a Qualitative model is called "Classification".



Machine Learning Algorithms

- There are many different ML algorithms that all try to do the same things (Classification or Regression)
- Each ML algorithm has limitations on the types of model it can produce. The space of all possible models for an algorithm is called its "Hypothesis Set".
 - "Linear models" assume that the dependent variable is a function of the independent variables multiplied by a coefficient (e.g., f(field1 * coef1 + field2 * coef2 + ... + fieldM * coefM + constant))
 - "Tree models" assume that the dependent variable can be determined by asking a hierarchical set of questions about the independent data. (e.g., is height >= 50 feet? If so, is rainfall >= 11 inches?...then age is 65).
 - Some other models (e.g., Neural Nets) are difficult to explain or visualize and are therefore not considered "Explanatory".



Using Machine Learning

- Machine Learning is fun and easy
 - It is simple to invoke one of our ML algorithms, learn a model and make some predictions.
 - If you have some data, give it a try.
 - It will likely provide good insights and may identify some significant possibilities for adding value.



- Machine Learning is also dangerous (not just because of robot death rays)
 - There are many ways to produce inaccurate and misleading predictions.
 - Bad questions: yesterday's temperature -> today's stock close
 - Bad data Missing values, incorrect values -> bad predictions
 - Wrong assumptions regarding the algorithm chosen
 - Insufficient Training Data
 - Overfitting
 - Many others ...
 - Always consult a qualified Data Scientist before applying your models to a production activity.



HPCC Platform Machine Learning Approaches

- Embedded Language support for Python and R allow industry ML platforms:
 - Various R packages
 - Sckit-learn (Python)
 - Google TensorFlow (Python)
- Production Bundles
 - Fully supported, Validation tested, HPCC optimized, and Performance profiled.
 - Bundles are generally independent of platform release, and easy to install
 - The Bundle installer will alert if there is a platform version dependency for a particular bundle



ML Core and Supervised Bundles

- ML_Core Machine Learning Core (https://github.com/hpcc-systems/ML_Core)
 - Provides the base data types and common data analysis methods.
 - Also includes methods to convert between your record-oriented data and the matrix form used by the ML algorithms.
 - This is a dependency for all of the ML bundles.
- PBblas Parallel Block Basic Linear Algebra Subsystem (https://github.com/hpcc-systems/PBblas)
 - Provides parallelized large-scale matrix operations tailored to the HPCC Platform.
 - This is a dependency for many of the ML bundles.
- <u>LinearRegression</u> Ordinary least squares linear regression (https://github.com/hpcc-systems/LinearRegression)
 - Allows multiple dependent variables (Multi-variate)
 - Provides a range of analytic methods for interpreting the results
 - Can be useful for testing relationship hypotheses (
- <u>LogisticRegression</u> Logistic Regression Bundle (https://github.com/hpcc-systems/LogisticRegression)
 - Despite "regression" in its name, provides a linear model-based classification mechanism
 - Binomial (yes/no) classification
 - Multinomial (n-way) classification using Softmax.
 - Allows multiple dependent variables (Multi-variate)
 - Includes analytic methods.



Supervised ML Bundles (cont'd)

- <u>SupportVectorMachines</u> SVM Bundle (https://github.com/hpcc-systems/SupportVectorMachines)
 - Classification or Regression.
 - Uses the popular LibSVM library under the hood
 - Appropriate for moderate sized datasets or solving many moderate sized problems in parallel.
 - Includes automatic parallelized grid-search to find the best regularization parameters.
- GLM General Linear Model (https://github.com/hpcc-systems/GLM)
 - Provides various linear methods that can be used when the assumptions of Linear or Logistic Regression don't apply to your data.
 - Includes: binomial, quasibinomial, Poisson, quasipoisson, Gaussian, inverse Gaussian and Gamma GLM families.
 - Can be readily adapted to other error distribution assumptions.
 - Accepts user-defined observation weights/frequencies/priors via a weights argument.
 - Includes analytics.



Supervised ML Bundles (cont'd)

- <u>LearningTrees</u> —Random Forests (https://github.com/hpcc-systems/LearningTrees)
 - Classification or Regression
 - One of the best "out-of-box" ML methods as it makes few assumptions about the nature of the data, and is resistant to overfitting. Capable of dealing with large numbers of Independent Variables.
 - Creates a "forest" of diverse Decision Trees and averages the responses of the different Trees.
 - Provides "Feature Importance" metrics.
 - Later versions are planned to support single Decision Trees (C4.5) and Gradient Boosting Trees.
- Generalized Neural Network (GNN) (https://github.com/hpcc-systems/GNN)
 - Combines the parallel processing power of HPCC Systems with the powerful Neural Network capabilities of Keras and Tensorflow
 - Each node is attached to an independent Keras / Tensorflow environment, which can contain various hardware acceleration capabilities such as Graphical Processing Units (GPUs) or Tensor Processing Units (TPUs).
 - Provides a distributed environment that can parallelize all phases of Keras / Tensorflow usage.
 - Provides a Tensor module, allowing users to efficiently encode, decode, and manipulate Tensors within ECL. These Tensor data sets are used as the primary data interface to the GNN functions.



Unsupervised ML Bundles

- K-Means (https://github.com/hpcc-systems/KMeans)
 - Used to automatically cluster unlabeled data
 - It is an implementation of K-Means algorithm, an unsupervised machine learning algorithm to automatically cluster Big Data into different groups.
 - Adopts a hybrid parallelism to enable K-Means calculations on massive datasets and also with a large count of hyper-parameters. The hybrid parallelism can be realized by utilizing the flexible data distribution control and Myriad Interface of HPCC Systems.
- DBSCAN Scalable Parallel DBSCAN Clustering (https://github.com/hpcc-systems/DBSCAN)
 - DBSCAN is Density-Based Spatial Clustering of Applications with Noise
 - An unsupervised machine learning algorithm to automatically cluster the data into subclasses or groups.
 - Provides a range of analytic methods for interpreting the results
- <u>TextVectors</u> ML for Textual Data (https://github.com/hpcc-systems/TextVectors)
 - Allows for the mathematical treatment of textual information.
 - Words, phrases, sentences, and paragraphs can be organized as points.
 - Closeness in space implies closeness in meaning
 - Vectorization allows text to be analyzed numerically and used with other ML techniques.





Machine Learning Tutorial

Learning Trees(a.k.a. Random Forest)





Why Learning Trees?

- Widely considered to be one of the easiest algorithms to use. Why?
 - It makes very little assumption about the data's distribution or its relationships
 - It can handle large numbers of records and large numbers of fields
 - It can handle non-linear and discontinuous relations
 - It almost always works well using the default parameters and without any tuning
- It scales well on HPCC Systems clusters of almost any size
- Its prediction accuracy is competitive with the best state-of-the-art algorithms



Learning Trees Overview

- Random Forests are based on Decision Trees.
- In Decision Trees, you ask a hierarchical series of questions about the data (think a flowchart) until you have asked enough questions to split your data into small groups (leaves) that have a similar result value. Now you can ask the same set of questions of a new data point and return the answer that is representative of the leaf into which is falls.
- Random Forest builds a number of separate and diverse decision trees (a decision forest) and average the results across the forest. The use of many diverse decision trees reduces overfitting since each tree is fit to a different subset of the data, and it will therefore incorporate different "noise" into its model. The aggregation across multiple trees (for the most part) cancels out the noise.



Tutorial Contents

This tutorial example demonstrates the following:

- Installing ML
- Preparing Your Data
- Training the Model
- Assessing the Model
- Making Predictions



Installing ML and LearningTrees

Installing the ML Core Libraries and the LearningTrees Bundles

- 1. Be sure HPCC Systems Clienttools is installed on your system. Also, you will need **Git for Windows** if you do not already have it installed.
- 2. Install **HPCC Systems ML_Core** From your *clienttools/bin* folder, run in the Command Window: *ecl bundle install https://github.com/hpcc-systems/ML_Core.git*
- 3. Install the **HPCC Systems LearningTrees** bundle. Run: ecl bundle install https://github.com/hpcc-systems/LearningTrees.git

NOTE: For PC users, the ecl bundle install must be run as an Administrator. Right click on the command icon and select "Run as administrator" when you start your command window.

Easy! Now let's get your data ready to use.



Preparing Your Data

The data should contain all numeric values, and the first field of your record is an UNSIGNED unique identifier.

1. The first step is to segregate your data into a *training* set and a *testing* set, using a random sample.

Download and import MLTutorial.MOD

Open MLTutorial.Prep01.ECL



Preparing Your Data

 Convert your data to the form used by the ML bundles. ML requires the data in a celloriented matrix layout known as the NumericField.

```
IMPORT ML_Core;
ML Core.ToField(myTrainData, myTrainDataNF);
```

Open MLTutorial.Convert02.ECL

Record Formatted Data - N Records of M fields each

Record 1: Record Id, Field 1, Field 2, ..., Field M

Record 2: Record Id, Field 1, Field 2, ..., Field M

Record N: Record Id, Field 1, Field 2, ..., Field M

Cell-oriented Data - N * M records each holding one original field

Record 1: Record ID, Field Number, Value

Record 2: Record ID, Field Number, Value

...

Record N*M: Record ID, Field Number, Value

At this point, myTrainDataNF will contain your converted training data. Note that the MACRO does not return the converted data but creates an new in-line definition that contains it.



Preparing Your Data (More)

3. Separate the Independent fields from the Dependent fields:

Note that using the PROJECT to set the number field to 1 is not strictly necessary, but is a good practice. This indicates that it is the first field of the dependent data. Since there is only one Dependent field, we number it accordingly.

4. Do the same thing for the test data:

We are now ready to train the model and assess our results!



Training the Model

- 1. Define our learner (Regression/Classification Forest)
- 2. Train and Retrieve the Model
 - Regression (MLTutorial.Train03)
 - ✓ RegressionForest
 - ✓ GetModel/Predict
 - ✓ Accuracy Assessment
 - Classification (MLTutorial.Train04)
 - ✓ Discretize (ByRounding)
 - ✓ ClassificationForest
 - ✓ GetModel/Classify
 - ✓ Accuracy Assessment



Assessing the Model

Assessing the Model is fairly easy:

- 1. Create a new set of Independent field samples
- 2. Use the Predict FUNCTION for Regression:

```
predictedValues := myLearnerR.Predict(myModelR, myNewIndData);
```

3. Use the Classify FUNCTION for Classification:

```
predictedClasses := myLearnerC.Classify(myModelC, myNewIndData);
```

Note: For best results, your train/test data should be representative of the population whose values you will try to predict.



Conclusion

- You now know everything you need to know to build and test ML models against your data and to use those models to predict qualitative or quantitative values (i.e. classification or regression).
- If you want to use a different ML bundle, you'll find that all the bundles operate in a very similar fashion, with minor variation.
- We've utilized only the most basic aspects of the ML bundles.

And finally:

ML's conceptual simplicity is somewhat misleading. Each algorithm has its own quirks, (assumptions and constraints) that need to be taken into account in order to maximize the predictive accuracy.



End of Workshop!

Thank You!



