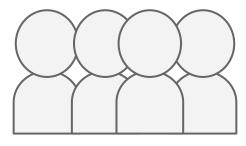
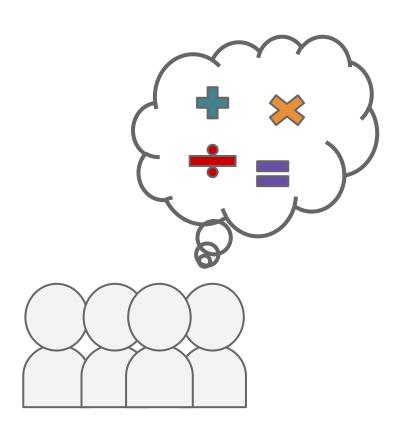
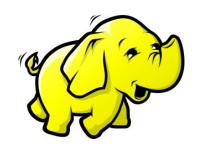
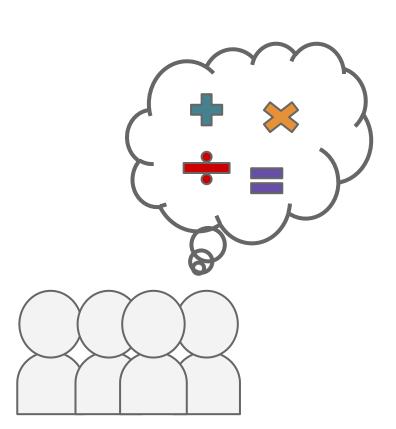
Simplifying Big Data with Apache Crunch

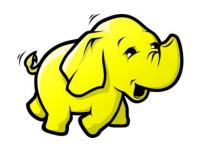
Micah Whitacre @mkwhit

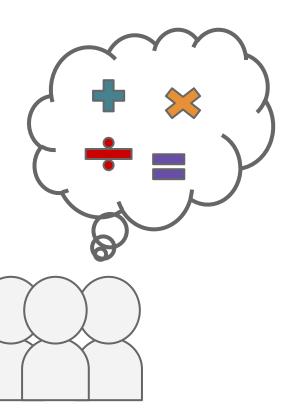




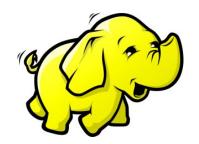


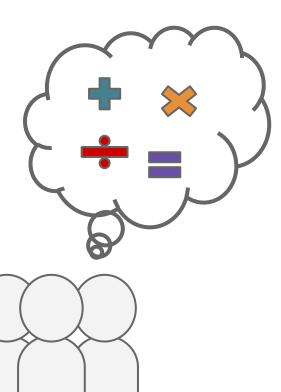






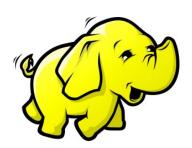




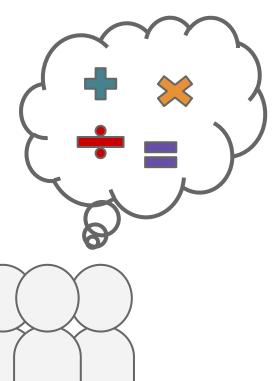






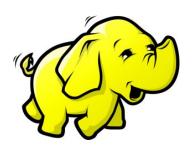


HBASE



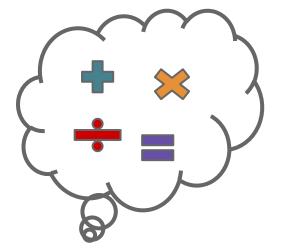




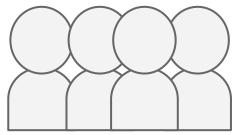




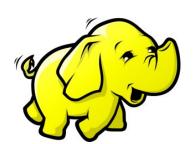








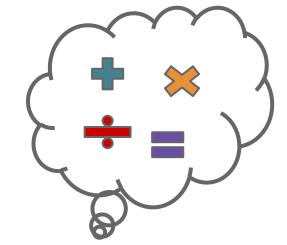




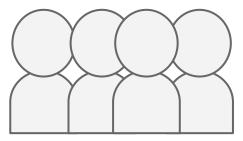




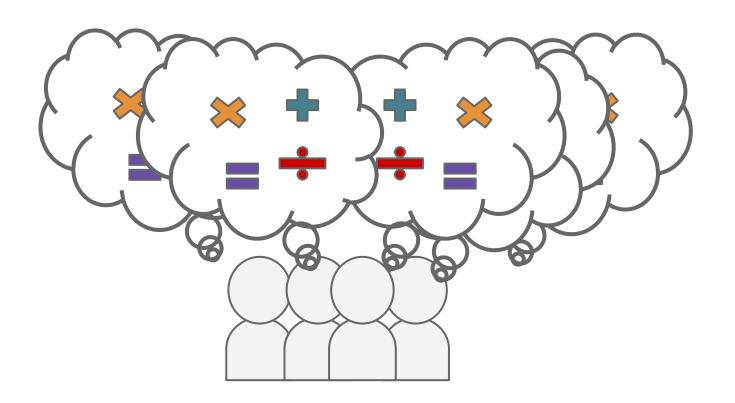














Semantic Chart Search

Medical Alerting System

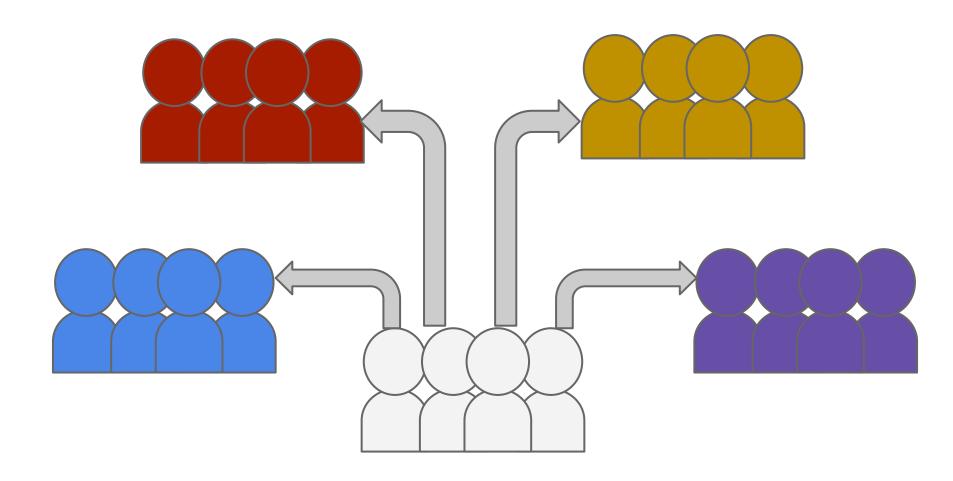
Cloud Based EMR

Population Health Management

Problem moves from scaling architecture...

Problem moves from not only scaling architecture...

To how to scale the knowledge



Battling the 3 V's

Battling the 3 V's

60+ different data formats

Battling the 3 V's

60+ different data formats

Battling the 3 V's

Constant streams for near real time

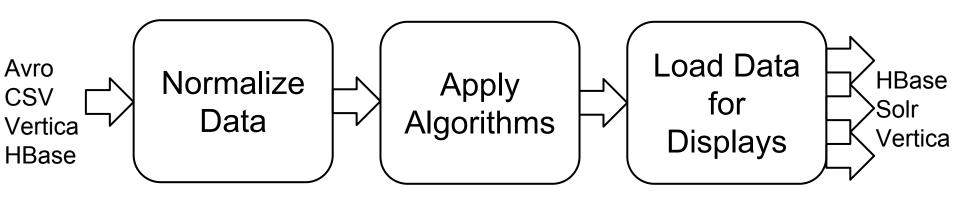
60+ different data formats

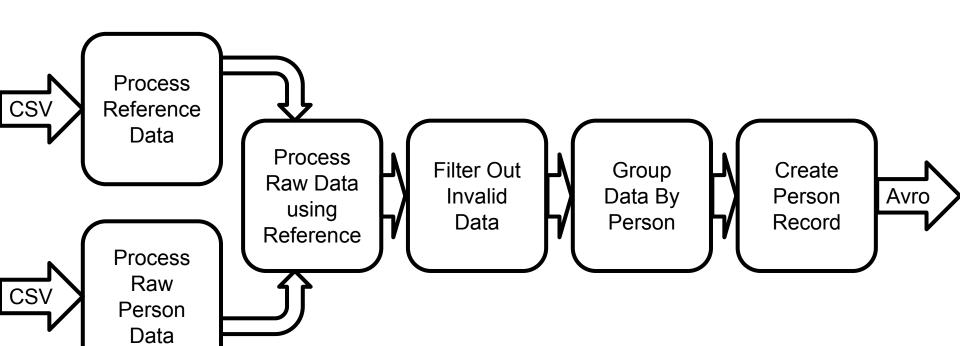
Battling the 3 V's

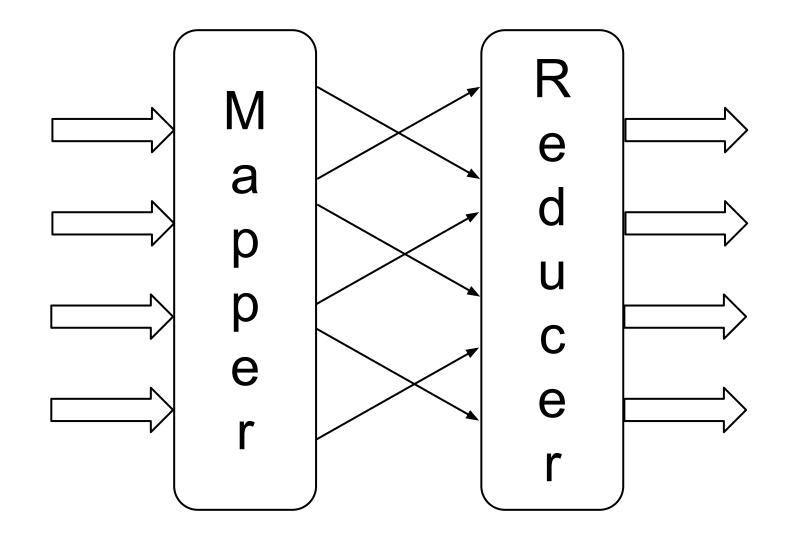
Constant streams for near real time

2+ TB of streaming data daily

Population Health







Integration done through persistence

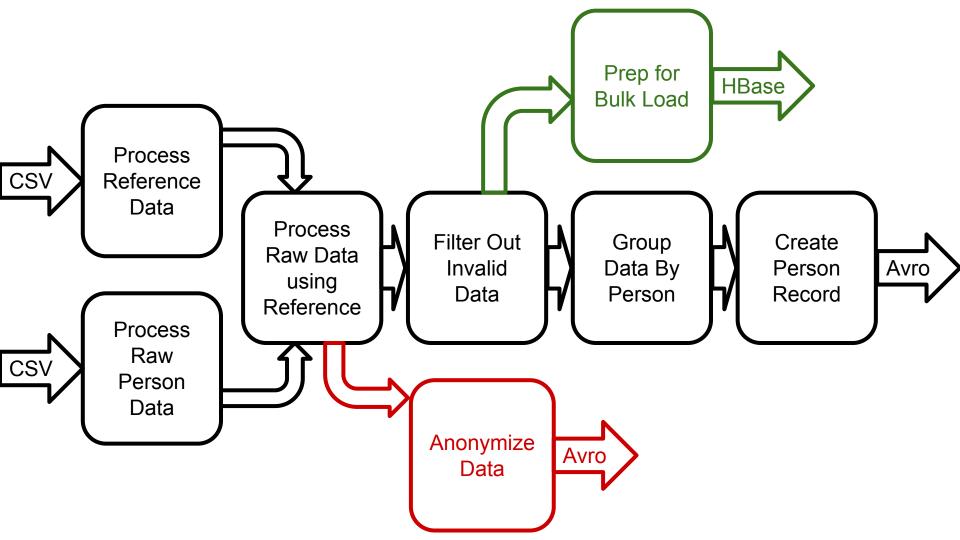
Integration done through persistence

Custom impls of common patterns

Integration done through persistence

Custom impls of common patterns

Evolving Requirements



Easy integration between teams

Focus on processing steps

Shallow learning curve

Ability to tune for performance

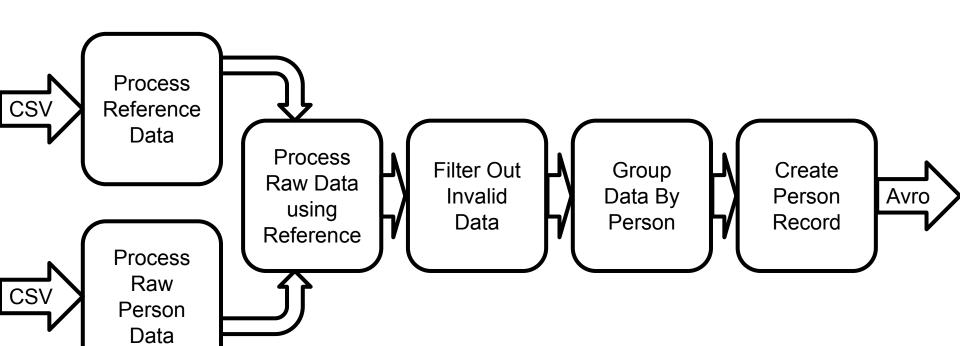
Apache Crunch

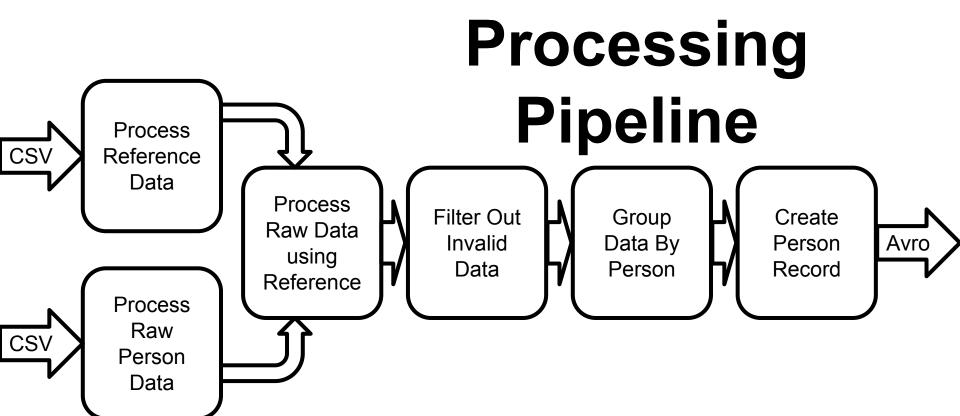
Compose processing into pipelines

Open Source FlumeJava impl

Transformation through fns (not job)

Utilizes POJOs (hides serialization)





Pipeline

Programmatic description of DAG Supports lazy execution Implementations indicate runtime MapReduce, Spark, Memory

```
Pipeline pipeline =
   new MRPipeline(Driver.class, conf);
Pipeline pipeline =
  MemPipeline.getIntance();
Pipeline pipeline =
   new SparkPipeline(sparkContext, "app");
```

Source

Reads various inputs

At least one required per pipeline

Creates initial collections for processing

Custom implementations

Source

Sequence Files

Avro

Parquet

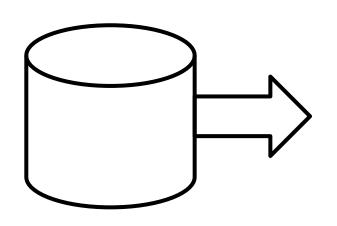
HBase

JDBC

HFiles

Text

CSV



Strings
AvroRecords
Results
POJOs
Protobufs
Thrift
Writables

pipeline.read(

From.textFile(path));

pipeline.read(new TextFileSource(path,ptype));

```
PType<String> ptype = ...;
pipeline.read(
  new TextFileSource(path,ptype));
```

PType

Hides serialization

Exposes data in native Java forms

Supports composing complex types

Avro, Thrift, and Protocol Buffers

Multiple Serialization Types Serialization Type = PTypeFamily Avro & Writable available Can't mix families in single type Can easily convert between families

```
PType<Integer> intTypes =
  Writables.ints();
PType<String> stringType =
 Avros.strings();
PType<Person> personType =
 Avros.records(Person.class);
```

```
PType<Pair<String, Person>> pairType =
Avros.pairs(stringType, personType);
```

```
PTableType<String, Person> tableType =
Avros.tableOf(stringType,personType);
```

```
PType<String> ptype = ...;
PCollection<String> strings = pipeline.read(
   new TextFileSource(path, ptype));
```

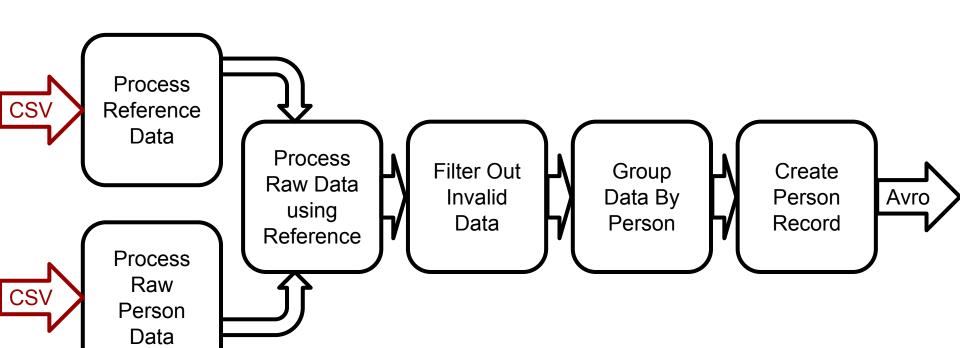
PCollection

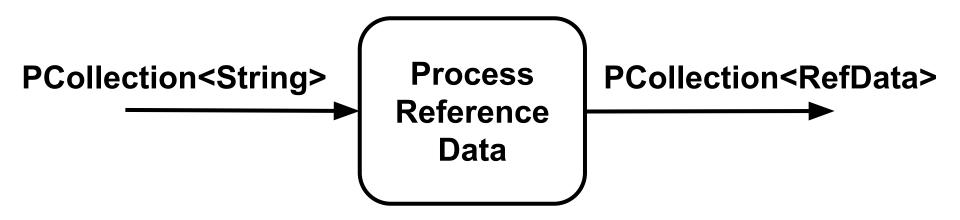
Immutable

Unsorted

Not created only read or transformed

Represents potential data





DoFn

Simple API to implement

Transforms PCollection between forms

Location for custom logic

Processes one element at a time

For each item emits 0:M items

MapFn - emits 1:1

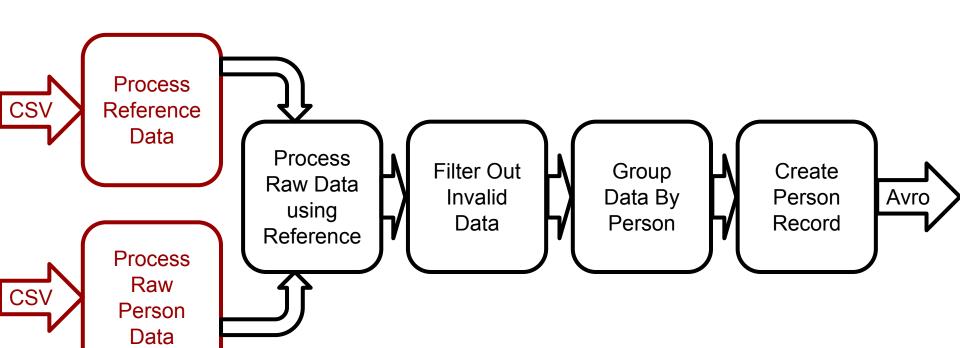
FilterFn - returns boolean

DoFn API

```
Type of Data Out
 Type of Data In
public void process
(String s,
   Emitter<RefData> emitter) {
   RefData data = ...;
   emitter.emit(data);
```

```
PCollection<String> refStrings
PCollection<RefData> refs =
   refStrings.parallelDo(fn,
Avros.records(RefData.class));
```

```
PCollection<String> dataStrs...
PCollection<RefData> refs =
   dataStrs.parallelDo(diffFn,
   Avros.records(Data.class));
```



Hmm now I need to join... But they don't have a common key? We need a PTable

PTable<K, V>

Immutable & Unsorted

Multimap of Keys and Values

Variation PCollection<Pair<K, V>>

Joins, Cogroups, Group By Key

```
class ExampleDoFn extends
DoFn<String, RefData>{
    ...
```

```
class ExampleDoFn extends
DoFn<String,
 Pair<String, RefData>>{
```

```
PCollection<String> refStrings
PTable<String, RefData> refs =
 refStrings.parallelDo(fn,
Avros.tableOf(Avros.strings(),
Avros.records(RefData.class)));
```

PTable<String, RefData> refs...;
PTable<String, Data> data...;

data.join(refs);

(inner join)

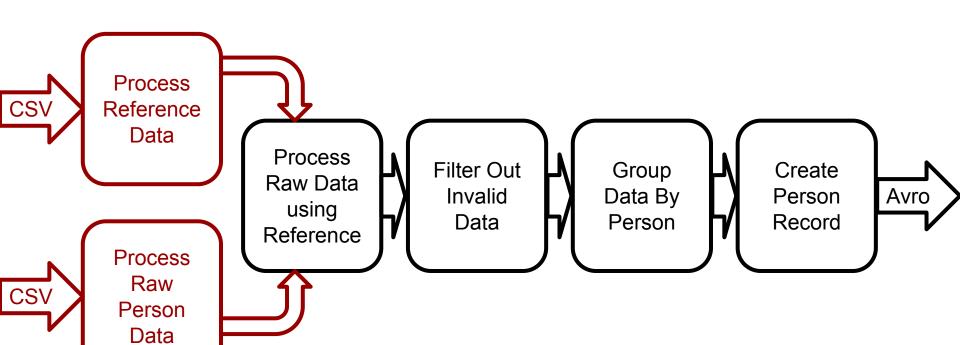
```
PTable<String,
  Pair<Data, RefData>>
  joinedData = data.join(refs);
```

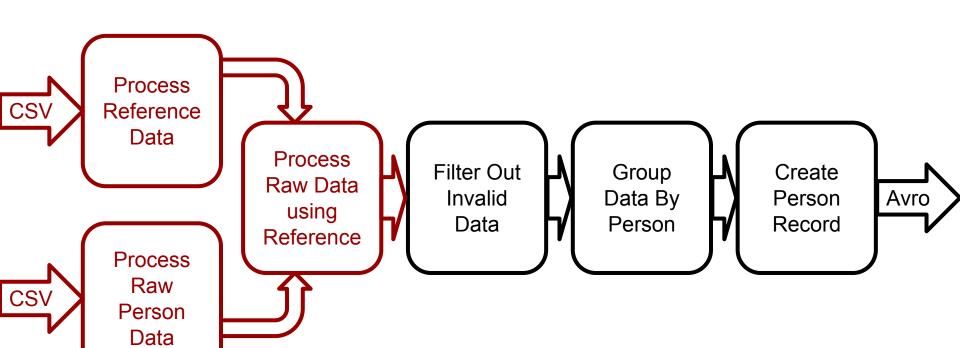
Joins

right, left, inner, outer

Eliminates custom impls

Mapside, BloomFilter, Sharded



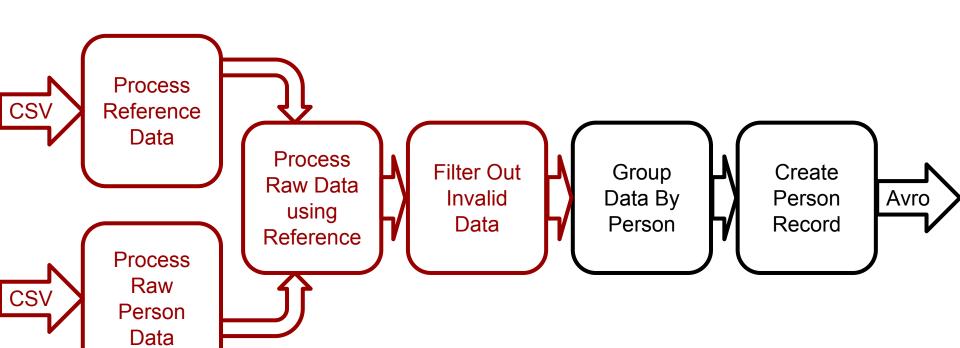


FilterFn API

```
public boolean accept
(... value){
```

return value > 3;

```
PCollection<Model> values = ...;
PCollection<Model> filtered =
 values.filter(new
   MyFilterFn());
```



Keyed By PersonId

↓

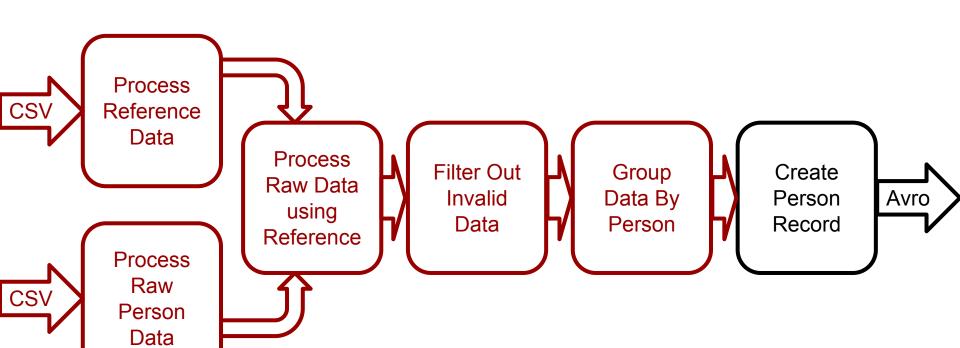
PTable<String, Model> models = ...;

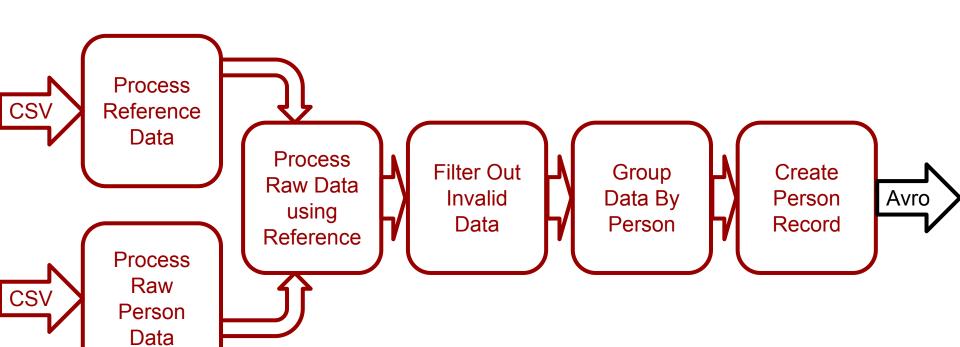
```
PTable<String, Model> models = ...;
PGroupedTable<String, Model>
    groupedModels =
    models.groupByKey();
```

PGroupedTable<K, V>

Immutable & Sorted

PCollection<Pair<K, Iterable<V>>>





PCollection<Person> persons = ...;

```
PCollection<Person> persons = ...;
pipeline.write(persons,
   To.avroFile(path));
```

```
PCollection<Person> persons = ...;
pipeline.write(persons,
   new AvroFileTarget(path));
```

Target

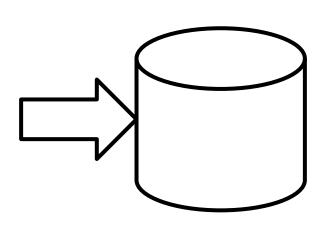
Persists PCollection

At least one required per pipeline

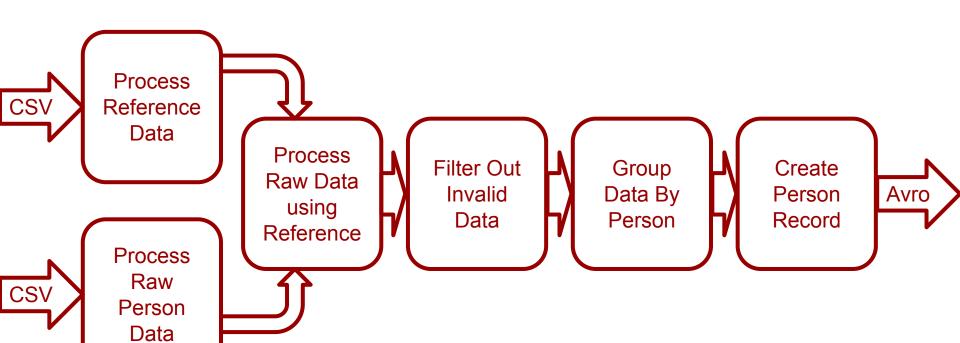
Custom implementations

Target

Strings
AvroRecords
Results
POJOs
Protobufs
Thrift
Writables

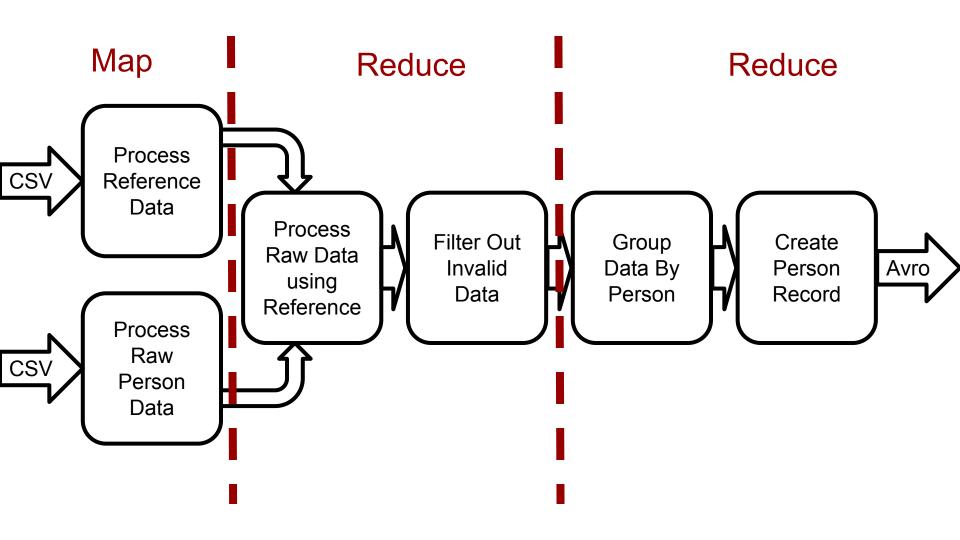


Sequence Files Avro Parquet HBase JDBC HFiles Text CSV



Execution

```
Pipeline pipeline = ...;
pipeline.write(...);
PipelineResult result
 pipeline.done();
```



Tuning

Tweak pipeline for performance

GroupingOptions/ParallelDoOptions

Scale factors

Functionality first Focus on the transformations **Smaller learning curve Less fragility**

Iterate with confidence

Integration through PCollections

Extend pipeline for new features

Links

http://crunch.apache.org/

http://dl.acm.org/citation.cfm?id=1806596.1806638

http://www.quora.com/Apache-Hadoop/What-are-the-differences-between-Crunch-and-Cascading