# State of structural typing support in Scala 3.3.0

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#### Abstract

Scala's type system is primarily based on nominal typing. Scala 3 introduces a special type, Selectable, which provides an infrastructure for structural typing. Karlsson and Haller proposed improvements to Selectable to support extensible records. In this paper, we review several Scala 3 projects that involve structural typing. We find that their implementation or usability would benefit from the extensible records proposal from Karlsson and Haller. We investigate the remaining common challenges when working with structural types. In particular, we identify that a dedicated syntax for both record types and record literals would be the most beneficial addition to the language.

Keywords: structural typing

## 1 Introduction

Nominal typing is very practical for most cases but, for some use cases, such as data manipulation involving intermediate transformations, it is not convenient. Indeed, defining a data type for every intermediate result, even if it is used only once, is cumbersome. On the other hand, structural typing is well suited for such cases as it allows developers to manipulate the structure of data on-the-fly without having to declare the type of every intermediate transformation.

A common workaround in Scala is to use a tuple. For instance, a method performing a Euclidean division would typically be defined with a return type of (Int, Int), modeling a pair containing both the quotient and the remainder. This solution has the benefit of being concise. On the other hand, it is not obvious to the users of that method which element of the pair is the quotient and which one is the remainder. It is easy to confuse them. A solution to that drawback would be to define a bespoke data type for the result of the Euclidean division:

case class EuclideanDivisionResult(quotient: Int, remainder: Int)

Thus, it would be obvious to the users which element is the quotient and which one is the remainder. However, in practice, Scala developers do not do follow this approach because it is too verbose compared to returning a tuple.

Structural typing would bring the best of both worlds: implementers would not need to define a data type such as EuclideanDivisionResult, and yet users would get a value with proper fields quotient and remainder preventing the risk of confusion.

The goal of this paper is to identify the issues that prevent Scala developers from using structural typing, and to propose solutions to them.

In the next section, I describe the current support of structural typing in Scala 3 based on the marker type Selectable, and I explain why it does not provide a satisfactory solution to the above example. Additionally, I present the language changes proposed by Karlsson and Haller to support extensible records[1], and I show how it improves the situation.

In their paper, Karlsson and Haller evaluated the performance of their records implementation. They showed that it is competitive with cached reflection for structural field access. However, they did not validate that their design provides a satisfactory solution to the use cases that involve structural types. In section 3, I review several existing libraries that involve structural types (currently based on Selectable), and I assess how much their implementation and usage would be simplified by the extensible records proposal.

In section 4, I summarize the pros and cons of the extensible records proposal, and I investigate the remaining common challenges regarding both the implementation and usage of the reviewed libraries. My analysis suggests that the most valuable addition to the extensible records proposal would be a dedicated syntax for record types and record literals.

The last section concludes.

In summary, the contributions of this work are the following:

- I evaluated the impact of the extensible records proposal on the implementation and usage of several real world libraries,
- I rebased the original implementation of the extensible records on the development branch of the Scala 3.3.x series,
- I proposed a new idea to improve further the support of structural typing in Scala 3: a dedicated syntax for record types and record literals.

## 2 Support of Structural Typing in Scala 3

In this section, I describe the current support of structural typing in Scala 3, and I present the extensible records proposal from Karlsson and Haller.

## 2.1 Status Quo: the Marker Type Selectable

Scala 3 comes with a special type, Selectable, which provides a minimal infrastructure for structural typing.

As an example, the result of the Euclidean division can be modeled by the type Record { val quotient: Int; val remainder: Int }, provided that type Record is defined as a subtype of Selectable that declares a method named selectDynamic:

```
trait Record extends Selectable:
    def selectDynamic(label: String): Any
```

Assuming we would have a value result of type Record { val quotient: Int; val remainder: Int }, the compiler would type the expression result.quotient as Int, and the expression result.remainder as Int. On the other hand, the compiler would reject any attempt to select any other field, such as result.fractional, for instance.

The way Selectable works is that the compiler only allows you to select the fields that were declared in the type refinement (between the braces following Record). It rewrites fields' selections to calls to the method selectDynamic, which takes the name of the selected field as a parameter of type String, and returns the value of that field.

For end-users, Selectable only is half satisfactory. On the one hand, once they get the result of the Euclidean division, they can access the quotient and remainder in a concise and type-safe way. On the other hand, the result type of the method is verbose, as illustrated by the following snippet that compares a tuple type expression, a case class reference, and a structural type expression:

#### (Int, Int)

```
EuclideanDivisionResult
Record { val quotient: Int; val remainder: Int }
```

As you can see, the structural type expression is the most verbose of the three.

For the implementers of the method that performs the Euclidean division, **Selectable** is not satisfactory at all because there is no simple and type-safe way to create a structurally-typed value. The infrastructure in the compiler is only responsible for rewriting fields' selections into calls to **selectDynamic**, but that method still needs to be implemented.

We could define **Record** as a class that internally stores fields' values in a Map and implements **selectDynamic** by performing a look-up in that Map:

### case class Record(fields: Map[String, Any]) extends Selectable: def selectDynamic(label: String): Any = fields(label)

With this **Record** data type defined, the implementers of the method that performs the Euclidean division could return a structurally-typed value as follows:

```
Record(Map("quotient" -> quotient, "remainder" -> remainder))
.asInstanceOf[Record { val quotient: Int; val remainder: Int }]
```

But that would still be much more verbose than returning a tuple or an instance of the class EuclideanDivisionResult. Furthermore, this is error-prone since any mistake in the type passed to the asInstanceOf call would not be caught by the compiler (e.g., a mismatch between the Map content and the declared record type).

Another important limitation in the current support of structural types that is not illustrated by my example is that there is no way to concatenate two structural

types (for example to add new fields to a record). The type intersection operation does almost that, but it can not detect clashes (such as extending a record with a field of the same name but with a different type). The limitations of structural types in Scala 3 have been discussed in depth by Karlsson and Haller, and motivated their proposal for extensible records, which I describe in the next subsection.

### 2.2 Extensible Records Proposal

As described in the previous section, Selectable alone does not provide satisfactory support of structural typing in Scala 3. Karlsson and Haller showed that the key missing piece is the ability to *extend* structurally-typed records in a type-safe way[1]. Indeed, type-safe extension of records also provides type-safe *creation* of records since any record can be constructed by extending the empty record with additional fields.

Here is how one can create a record containing a quotient and a remainder with their proposal:

Record() + ("quotient" ->> quotient) + ("remainder" ->> remainder)

Or, with the specific constructor for records with two fields:

Record("quotient" ->> quotient, "remainder" ->> remainder)

The type of both expressions is inferred to Record  $\{$  val quotient: Int; val remainder: Int  $\}$ .

Without going into details about their proposal, the following points are worth noting:

- to ensure soundness of record extension, a record can only be extended by one field at a time (via the operator +),
- most of the proposal is implemented as a library but some typing rules have been changed in the compiler to simplify the type expression Record { val a: A } & Record { val b: B } into Record { val a A; val b: B },
- no special syntax is introduced.

## 3 Evaluation of Extensible Records on Real World Libraries

In this section, I look at three libraries that use structural typing, and I evaluate how much the extensible records proposal would simplify them.

I consider the following axes of simplification:

- 1. implementation: do extensible records make those libraries easier to implement and maintain? For instance, by subsuming some parts of their implementation.
- 2. usage: is the user-facing API easier to navigate through? Is it easier to use? For instance, by having clearer type signatures, or a more concise syntax.

I selected the following libraries:

• Chimney, a library for transforming data structures by adding, removing, or renaming fields,

- Tyqu, a library for constructing SQL queries in a composable and type-safe way,
- Iskra, a library that adds more precise types to Spark data frames.

I rebased the implementation of extensible records (published in 2018) on top of Scala 3.3.0, re-implemented a small subset of Chimney, Tyqu, and Iskra as selfcontained tests, and looked at possible simplification with records<sup>1</sup>.

In the following subsections, I will give an overview of each library, I will explain how they are implemented, and I will assess how much they would be simplified by extensible records.

All the code examples assume following class definitions:

```
case class UserV1(name: String)
case class UserV2(name: String, age: Option[Int])
```

## 3.1 Chimney

Chimney transforms data structures by adding, removing, or renaming fields. For instance, here is a snippet that converts a value of type UserV1 into a value of type UserV2:

```
val userV1 = UserV1("Martin")
val userV2 =
    userV1
    .into[UserV2]
    .withFieldComputed(_.age, _ => None)
    .transform
```

We start with the value userV1, of type UserV1, and we call the Chimney extension method into on it, with the type of the target class UserV2.

Chimney tries to create a mapping from UserV1 to UserV2 by looking at the fields in both classes that have the same name and type. In this example, the field name of type String is present in both classes, but the field age is present only in the target class. Chimney creates a partial mapping (handling only the field name), which we need to complete to cover all the fields of the target class. We call withFieldComputed(\_.age, \_ => None) to tell Chimney to map the target field age to the value None.

Finally, when the mapping covers all the fields of the target class, we can call the method transform to perform the conversion.

The current implementation of Chimney relies on macros that analyze the structure of the source and target types to create the mapping between them. The operation withFieldComputed is also implemented via a transparent inline def that analyzes the first argument (\_.age in the example above) to extract the name of the field.



 $<sup>^1{\</sup>rm The}$  result is available at https://github.com/lampepfl/dotty/compare/main...julienrf:dotty:dotty-records-final.

On the usage side, the Chimney API is rather easy to navigate through and IDEs provide helpful suggestions. For instance, it is impossible to make typos on field names (e.g., agee instead of age) or assigning a value of an incompatible type to a field.

Furthermore, the authors of the library made a great job at providing sensible error messages. For instance, if we omit to provide the mapping for the **age** field:

```
Chimney can't derive transformation from UserV1 to UserV2
```

```
UserV2
```

age: Option[Int] - no accessor named age in source type UserV1

Consult https://chimney.readthedocs.io for usage examples.

There is a lot of overlap between what Chimney does and extensible records. Indeed, with both tools developers transform data structures by adding new fields to them.

Here is a possible sketch of how Chimney could look like if extensible records were part of the language:

```
val userV2 =
  (userV1.toRecord + ("age" ->> None))
   .into[UserV2]
```

We first convert the value userV1 to an extensible record, which we extend with a field age, and then we convert the resulting record into the case class UserV2.

Thus, the implementation of Chimney would be only responsible for providing conversions between case classes and records. The user-facing API would be reduced to the following two extension methods:

```
extension [A](value: A)
  def into[B](using Conversion[A, B]): B =
    value.convert
extension [A](recordLike: A)
  def toRecord[B <: Record](using Conversion[A, B]): B =
    recordLike.convert</pre>
```

In addition to those methods, Chimney would still be responsible for synthesizing given conversions between case classes and their corresponding record type.

A drawback of the records-based approach is that typos and type-mismatch errors in added fields would be caught only at the point of calling the finalizer method .into[UserV2].

## 3.2 Tyqu

Tyqu provides an embedded DSL to describe SQL queries from aggregation, projection, and selection operations. Here is a snippet that shows how to look up in a database the users whose name is "Martin":

object UsersV1 extends Table: val name = Column[String]()

```
val results: Seq[Result { def firstName: String }] =
from(UsersV1)
.filter(user => user.name === "Martin")
.map(user => user.name.as("firstName"))
.run(dbConnection)
```

We start by defining the schema of the UsersV1 table via the object of the same name. It has just one column, name, of type String.

Then, we create a query that looks up in the UsersV1 table by calling from(UsersV1), and we transform that query by calling its filter and map methods. Ultimately, we execute the query by calling the run method with an actual database connection.

Tyqu provides a type-safe way to reference and manipulate table columns. Both operations filter and map take as parameter a function that takes a model of the table schema and returns a SQL expression.

For example, we can refer to the column name by writing user.name, which has type Expression[String]. This lets us write the condition user.name === "Martin", which has type Expression[Boolean] as expected by the filter operation. Tyqu will translate those expressions into proper SQL when it runs the query.

To achieve this, Tyqu synthesizes a type that models the table schema (ie, the type of the user parameter). It does that via a macro that introspects the content of the object UsersV1 and finds all its fields of type Column. It returns the structural type Selectable { def name: Expression[String] }.

Another point worth noting is how projection operations can transform the structural type of the underlying schema. For instance, in the call to map we return user.name.as("firstName"), which creates the alias firstName for the column name. Subsequent transformations of the query would now have to use the identifier firstName to refer to the column instead of name.

When we run the query, Tyqu returns a result of the structural type Result { def firstName: String } according to the alias we created (the type Result, defined by Tyqu, is also a subtype of Selectable).

There is some overlap between what Tyqu does and extensible records. The most obvious one is the **Result** type defined by Tyqu, which is literally the same as **Record**. Additionally, both Tyqu and extensible records implement a way to create values of structural types. Tyqu achieves that via the **as** operation (as in user.name.as("firstName")).

Here is a possible sketch of how Tyqu could look like if extensible records were part of the language:

```
object UsersV1Table:
  val name = Column[String]()
val results: Seq[Record { def firstName: String }] =
  from(UsersV1Table)
  .filter(user => user.name === "Martin")
```

```
.map(user => Record("firstName" ->> user.name))
.run(connection)
```

The differences with status quo are the following:

- executing a query now returns standard records of type Record instead of Result,
- the creation of the alias firstName is achieved by constructing a standard record.

The benefits of the second points are debatable, as some users may prefer using the original **as** method because it looks more similar to SQL. I will elaborate further on that point in the Discussion section.

Extensible records would not have a significant impact on the implementation of Tyqu. Indeed, the main tasks related to structural types in Tyqu consists of computing the type modeling table schemas from table object definitions, and computing the type of a query result based on the structural type of the query schema. Both tasks might be achieved with some form of generic programming support on records, but such features are not included in the extensible records proposal.

### 3.3 Iskra

Spark SQL allows developers to perform relational operations on data structured in columns. A drawback of Spark SQL is that its main abstraction, DataFrame, does not provide a precisely typed schema of the manipulated data, making it easy to do mistakes such as typos in column names, or applying an operation incompatible with the underlying column type.

The goal of Iskra is to be as source compatible as possible with Spark's DataFrame while providing more precise types in order to catch the type errors mentioned above at compile-time. Here is a simple example of Spark SQL program that prints the users whose name is "Martin", and then the same program with Iskra:

```
// Spark SQL
(users: DataFrame)
  .select($"name".as("user_name"))
  .where($"user_name" === "Martin")
  .show()
// Iskra
(users: DataFrame[UserV1])
  .select($.name.as("user_name"))
  .where($.user_name === "Martin")
  .show()
```

The only difference is that column selection is typechecked in Iskra: the expression \$.name has type Column[String].

Also shown in this example is the definition of column aliases, similar to the Tyqu example. The result of the select operation is a DataFrame with one column named user\_name and whose type is String. As a consequence, we refer to that column in the following where call by using \$.user\_name.

Iskra computes a structural type modeling the schema of the manipulated DataFrame that is pretty similar to what Tyqu does. Iskra would marginally benefit from extensible records, like Tyqu.

Here is a possible sketch of how Iskra could look like if extensible records were part of the language:

```
(users: DataFrame[UserV1])
  .select(Record("user_name" ->> $.name))
  .where($.user_name === "Martin")
  .show()
```

Like with Tyqu, we use standard records to manipulate structurally-typed values, which provides the same debatable benefit: the resulting user-facing API would be pretty different from the original Spark SQL API. I will discuss further that point in the next section.

The other main tasks related to structural types in Iskra consists of computing the type modeling the DataFrame schema, and computing the type of the result of evaluating a computation. Like with Tyqu, both may be achievable with some support of generic programming on records, but such a feature is not included in the extensible records proposal.

## 4 Discussion

In this section, I summarize the impacts of the extensible records proposal on the three studied libraries, and I summarize the remaining common challenges when working with structural types. Then, I make new proposals to address those challenges.

## 4.1 Pros and Cons of Extensible Records

The following table summarizes the impact of extensible records on the studied libraries ("+" means a marginal positive impact, and "++" means a significant positive impact):

	Chimney	Tyqu	Iskra
Implementation	++	+	+
Usage	=	+	+

#### 4.1.1 Impact on Implementation

The only case where the benefits are clear is the implementation of Chimney. Indeed, 80% of Chimney's implementation is a DSL to extend data structures. That part is subsumed by the extensible records proposal. The remaining 20% is an infrastructure to convert between case classes and records.

The impact of the extensible records on the implementation of Tyqu and Iskra is only marginal. Indeed, most of the work related to structural types performed by Tyqu and Iskra consists of computing the types of SQL table schemas (or DataFrame schemas, respectively), and computing the result types of running the queries (or evaluating the computations, respectively).

For example, in Iskra, calling the map method with a function that returns a record of type Record { val a: Column[A]; val b: Column[B] } produces a DataFrame[Schema { val a: Column[A]; val b: Column[B] }]. ie, the schema of the resulting DataFrame has the same columns as the provided record. Note that it might be possible to just reuse the Record type itself to model a schema instead of using a bespoke Schema type, but the current implementation in Iskra requires the Schema type to have specific additional methods that are not on Record.

Likewise, evaluating a computation on a DataFrame[Schema { val a: Column[A]; val b: Column[B] }] produces a result of type Seq[Record { val a: A; val b: B }]. Here, the structural type of the result has the same field names as the DataFrame schema, but their types do not contain the Column type constructor anymore.

Similar examples could be constructed for Tyqu.

Unfortunately, the extensible records proposal is of no help to compute such types.

#### 4.1.2 Impact on Usage

In all the cases, the impact of extensible records on the usability of the libraries is marginal.

Nevertheless, a general improvement is that once records become standard in Scala, then there is a unique, standard way to create and extend records that is shared by all the libraries. The user experience is more consistent across the libraries, and fewer concepts need to be learned to get started with a new library.

Other than that, the current design of the extensible records proposal still suffers from usability issues.

A minor issue is that in the proposed design, records can only be extended by using the + symbolic operation. This does not work well when mixed with non-infix calls. Compare for instance the following typical usage of Chimney:

foo

```
.into[Bar]
.withFieldConst(_.baz, 42)
.withFieldComputed(_.quux, _.x + 1)
.transform
```

With the following equivalent program based on my records-based adaptation of Chimney:

#### (foo

```
.toRecord
+ ("baz" ->> 42)
+ ("quux" ->> foo.x + 1))
.into[Bar]
```

The combination of infix operators and dot-notation does not play well.

This problem can easily be solved by providing a non-symbolic alias to the operation +, as I will show in section 4.2.1.

Another, more important issue with the extensible records design is that the syntax to define a field is not intuitive. To define a field name, developers write a **String** literal. This is inconsistent with the usual way of creating bindings in Scala, which consists of writing an unquoted identifier in a binding position. Fixing this problem would require changes to the language. Obviously, the current implementations of Tyqu and Iskra also suffer from this problem.

The last important issue is also related to syntax. As shown in the introduction, the syntax of structural types is very verbose. As a reminder, here is the type of records containing the quotient and remainder of a division:

Record { val quotient: Int; val remainder: Int }

It is obviously much more verbose than a tuple containing the same data.

This problem is important since, unlike classes, structural types can not be referred to by a simple identifier such as EuclideanDivisionResult. Defining a type alias would not really solve this issue because structural type expressions would still show up in signatures in the API documentation or in IDEs. Structural types are meant to be referred to by their type expression. Defining such an alias would defeat their purpose.

In the next section, I propose some ideas to fix this issue and the other aforementioned issues.

Finally, another minor usability issue with the current implementation of extensible records is that the type name Record clashes with java.lang.Record, introduced in JDK 16.

### 4.2 Proposals to Address the Remaining Issues

I propose three possible additions to the original extensible records proposal to address the issues described in the previous section: a non-symbolic operation to extend records, a dedicated syntax for record types and record literals, and compiler-synthesized mirrors for record types.

#### 4.2.1 Non-Symbolic Record Extension Operation

As mentioned in the previous section, in the extensible records proposal, the only way to extend a record is to use a symbolic + operation, which does not play well with the dot-notation. We can fix this issue by adding a non-symbolic alias withField to the + operation on records. With this change, the initial example becomes:

foo

```
.toRecord
.withField("baz" ->> 42)
.withField("quux" ->> foo.x + 1)
.into[Bar]
```

## 4.2.2 Dedicated Syntax

The syntax for records should be concise, readable, and lead to as few ambiguities as possible.

The syntax for record types and record literals should be homogeneous, just like the syntax of tuple types and tuple literals, and the syntax of case class definitions and case class construction.

Tuples, records, and case classes are all means of structuring information. Ideally, the syntax of records should share similarities with the syntax of tuples and case classes.

With these principles in mind, I propose the following syntax to define a record containing the quotient and remainder of a Euclidean division:

(quotient: Int, remainder: Int)

This type would expand to the following:

Record { val quotient: Int; val remainder: Int }

Constructing a record literal of that type looks as follows:

(quotient = dividend / divisor, remainder = dividend % divisor)

Extending a record looks as follows:

(foo = 42) + (bar = "hello")

Or, with the non-symbolic operator:

(foo = 42).withField((bar = "hello"))

Just like auto-tupling allows developers to call a function taking a tuple as if it was a function taking several parameters, "auto-recording" allows developers to call a function taking a record as if it was a function taking several named parameters:

(foo = 42).withField(bar = "hello")

"auto-recording" can be a source of ambiguities, though, and more work is needed to find the right set of rules governing its application.

Another benefit of having a dedicated syntax for record literals is that it allows the extension of a record with more than one field at a time, as in:

(foo = 42) + (bar = "hello", baz = true)

This was not possible in the original extensible records proposal because of soundness issues.

Note that extending a record with a non-literal would still be forbidden:

val r = (bar = "hello", baz = true)
(foo = 42) + r // error

Last, records should also support named field patterns, as in SIP-44. With this syntax, the code examples presented earlier can be rewritten as follows:

```
userV1
  .toRecord
  .withField(age = None)
  .into[UserV2]
foo
  .toRecord
  .withField(baz = 42)
  .withField(quux = foo.x + 1)
  .into[Bar]
val results: Seq[(firstName: String)] =
  from(UsersV1Table)
    .filter(user => user.name === "Martin")
    .map(user => (firstName = user.name))
    .run(connection)
users
  .select(user_name = $.name)
  .show()
```

I believe the simpler type signatures and literal syntax would make it easier to work with structural types.

An open question is whether the dedicated syntax should apply to records only, or to structural types in general. Even if records become standard in Scala, the underlying, more general, mechanism based on Selectable supports other forms of structurally-typed values. Probably, a more general syntax that would also apply to any form of structural type would be useful. A possibility could be to support a shorthand syntax for type refinements: the type A(x: X, y: Y) would be a shorthand for A { val x: X; val y: Y }.

#### 4.2.3 Generic Programming Support for Records

The last challenge related to using structural types, faced by both Tyqu and Iskra, is to derive a structural type from another (possibly structural) type.

Let us remind one of the examples presented earlier. In Iskra, the type DataFrame[Schema { val a: Column[A]; val b: Column[B] }] describes a computation that produces results of type Record { val a: A; val b: B }. Computing the result type is mechanical: for every field f: Column[T] of the schema type, there is a corresponding field f: T in the resulting record. A generic solution to perform this type computation requires a way to iterate over the fields of a structural type.

In Scala 3, mirrors allow developers to iterate over the fields of case classes. However, this technique can not currently be used with structural types because mirrors are not synthesized for them. In the remainder of this section, I propose new rules to synthesize mirrors for structural types and assess how much that would simplify the implementation of Iskra, Tyqu, and Chimney. Currently, the compiler synthesizes a subtype of Mirror.Product for every case class definition:

```
trait Product extends Mirror:
  type MirroredElemLabels <: Tuple
  type MirroredElemTypes <: Tuple
  type MirroredLabel <: String
  type MirroredMonoType
  def fromProduct(product: scala.Product): MirroredMonoType
```

For instance, here is a case class definition and its mirror type as synthesized by the compiler:

```
case class EuclideanDivisionResult(quotient: Int, remainder: Int)
```

#### // Synthesized by the compiler

```
trait EuclideanDivisionResultMirror extends Mirror.Product:
  type MirroredElemLabels = ("quotient", "remainder")
  type MirroredElemTypes = (Int, Int)
  type MirroredLabel = "EuclideanDivisionResult"
  type MirroredMonoType = EuclideanDivisionResult
  def fromProduct(product: scala.Product): EuclideanDivisionResult =
    EuclideanDivisionResult(
      product.productElement(0).asInstanceOf[Int],
      product.productElement(1).asInstanceOf[Int]
    )
```

The type members of EuclideanDivisionResultMirror model the structure of the class EuclideanDivisionResult. The mirror also implements the method fromProduct, which provides a constructor for EuclideanDivisionResult. It is worth noting that this constructor is position-based: it iterates over the elements of the provided product and passes them as parameters to the primary constructor of the class EuclideanDivisionResult.

That mechanics would not work as is on structural types because position is not significant in structural types. Instead, the constructor should be name-based:

#### def fromFields(fields: Map[String, Any]): MirroredMonoType

The complete template type for record type mirrors would be the following:

```
object Mirror:
  trait Record extends Mirror:
   type MirroredElemLabels <: Tuple
   type MirroredElemTypes <: Tuple
   type MirroredLabel = Nothing
   type MirroredMonoType <: scala.Record
   def fromFields(fields: Map[String, Any]: MirroredMonoType
```

Then, the compiler would synthesize the following mirror type for the record type modeling the result of a Euclidean division:

```
trait RecordQuotientRemainderMirror extends Mirror.Record:
  type MirroredElemLabels = ("quotient", "remainder")
  type MirroredElemTypes = (Int, Int)
  type MirroredMonoType =
    Record { val quotient: Int; val remainder: Int }
  def fromFields(
    fields: Map[String, Any]
  ): Record { val quotient: Int; val remainder: Int } =
    Record()
    + ("quotient" ->> fields("quotient").asInstanceOf[Int])
    + ("remainder" ->> fields("remainder").asInstanceOf[Int])
```

Note that such mirrors could not be synthesized for Selectable in general, because Selectable also supports structural method calls via a method applyDynamic, not just structural field selection via selectDynamic.

The benefits of record mirrors would be to provide a standard, high-level, way to iterate over the fields of a record type.

Currently, it is already possible to iterate over the fields of structural types by iterating over the Refinement nodes of the TypeRepr model of the structural type, with TASTy reflection. However, such refinements may also include arbitrary refinements such as { type X; def f(x: Int): X }.

Record mirrors would simplify the work of the developers by removing the need to check the structure of the refinements: the compiler would synthesize a mirror only if the refinements all have the form val <name>: <type>, and that none of the field types uses another field name as a type prefix (as in { val x: X; val y: x.Y }).

Even though our motivating example only mentions Iskra and Tyqu, it is likely that record mirrors would also benefit to a library like Chimney to implement the conversions from records to case classes. The developers would summon both the mirror of the source record type and the target case class, check that all the fields of the case class are present in the record type and have a compatible type, and finally construct the case class instance by retrieving all the fields' values by calling selectDynamic on the underlying record instance.

That being said, record mirrors would also have some drawbacks. Unlike case class mirrors, which have almost no compile-time and run-time footprint, record mirror types cannot be cached by the compiler as efficiently as case class mirrors, and their usage would require the allocation of extra objects.

In conclusion, record mirrors may provide additional benefits on the implementation of libraries manipulating structural types, but more work is necessary to more precisely assess their impact.

## 5 Conclusion

In this paper, I reviewed the support of structural typing in Scala 3.3.0 and identified possible ways to improve it.

The first issue is that there is no standard and type-safe way to create structurallytyped values. I assessed that that issue is fully solved by the extensible records proposal from Karlsson and Haller.

My analysis showed that the next most important issue is that the syntax of structural type expressions is too verbose. Structural types, by definition, do not have a short name that can be used to reference them. Therefore, they need a concise syntax to be usable at scale. I proposed a syntax for record type expressions, and a consistent syntax for record literals.

I also investigated other minor issues and their possible solutions, such as generic programming on records via compiler-synthesized mirrors.

## References

 Karlsson, O., Haller, P.: Extending scala with records: design, implementation, and evaluation. In: Proceedings of the 9th ACM SIGPLAN International Symposium on Scala, SCALA@ICFP 2018, St. Louis, MO, USA, September 28, 2018, pp. 72– 82. ACM, New York, NY, USA (2018). https://doi.org/10.1145/3241653.3241661