

# Machine Learning in Python with No Strings Attached

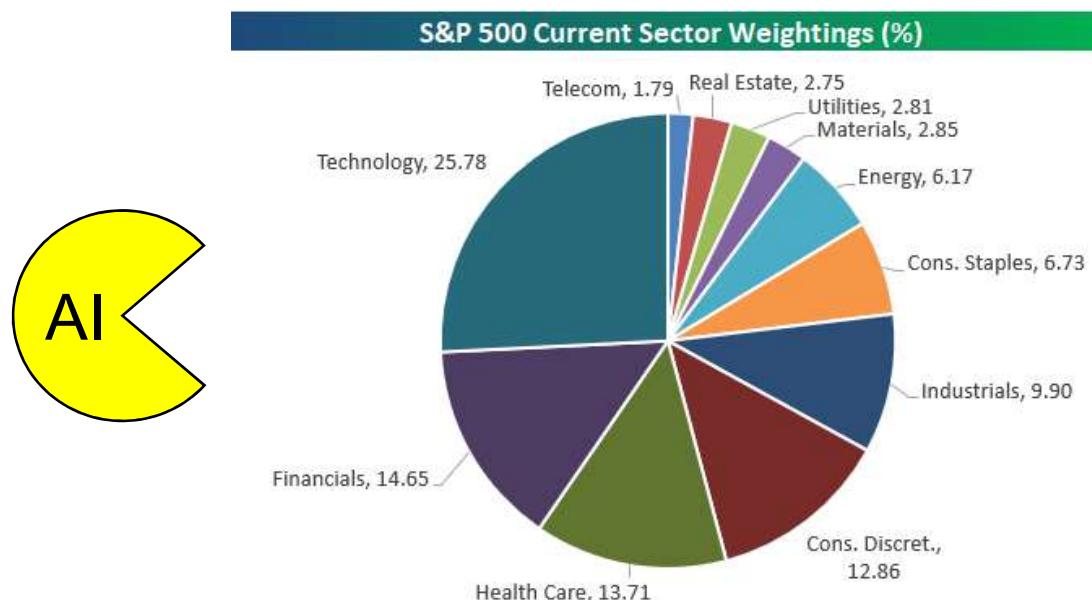
Guillaume Baudart, Martin Hirzel, Kiran Kate,  
Louis Mandel, and Avi Shinnar

IBM Research, USA

6/22/2019 talk at MAPL Workshop

# Why PL for AI?

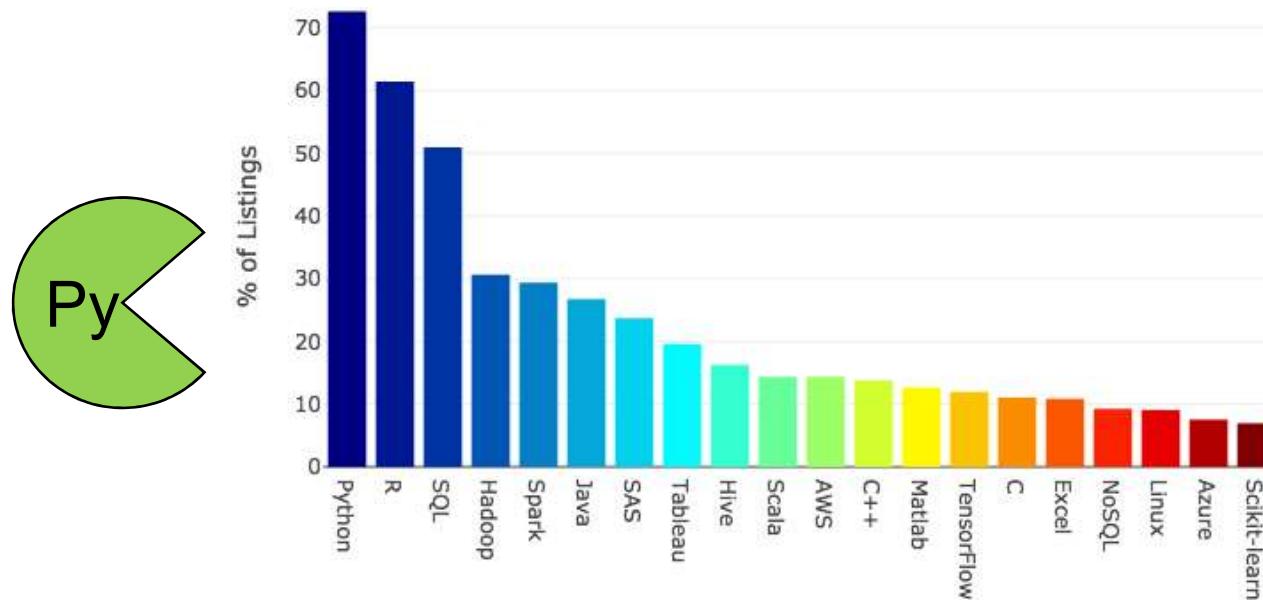
- IBM customers want AI
  - Large non-tech companies
  - Not enough skilled data scientists
- PL can lower the required skill level
  - Consistency, conciseness, error checking



Source: <https://seekingalpha.com/article/4172093-s-and-p-500-sector-weightings-tech-nears-26-percent>

# Why Python?

- Good PL design: consistency, conciseness, ...
- Many libraries: Keras, Sklearn, Spark, ...
- Jupyter notebooks
- Popularity
  - Top 20 technology skills in data science job listings



Source: <https://towardsdatascience.com/the-most-in-demand-skills-for-data-scientists-4a4a8db896db>

# Why Embedded Languages?

## Build computational graph

- Host language: Python
- Guest language: Keras, Sklearn, Spark SQL, ...

## Evaluate

- Train vs. predict
- Run on GPU or cluster
- Save for later reuse
- Automatic differentiation
- Back-propagation
- Optimization
- Monte-Carlo sampling



Lazy evaluation

Just a library –*but*– also a language.

# Keras Example

```
In [2]: import keras
from keras.models import Model
from keras.layers import Dense, Dropout, Input

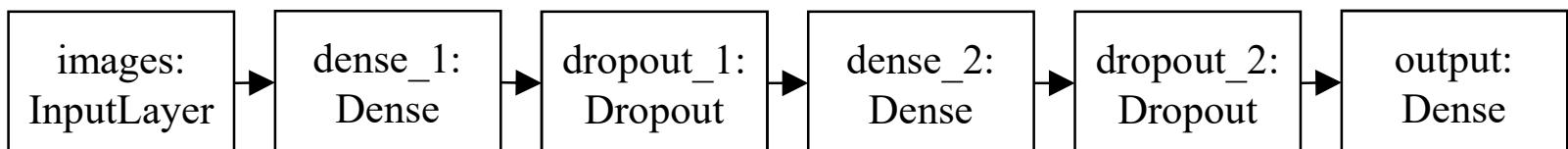
images = Input(shape=(N_PIXELS,), name='images')
hidden1 = Dense(512, activation='relu')(images)
dropped1 = Dropout(0.2)(hidden1)
hidden2 = Dense(512, activation='relu')(dropped1)
dropped2 = Dropout(0.2)(hidden2)
output = Dense(N_LABELS, activation='softmax',
               name='output')(dropped2)

model = Model(inputs=images, outputs=output)
```

1  
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3  
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11  
12  
13

```
In [3]: keras_utils.plot_model(model)
```

Out[3]:



```
In [5]: model.compile(loss='categorical_crossentropy',
                    optimizer='RMSprop',
                    metrics=['accuracy'])
model.fit({'images': x_train}, {'output': y_train},
          batch_size=128, epochs=3,
          validation_data=(x_test, y_test))
```

1  
2  
3  
4  
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6

# Outline

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Strings Considered  
Harmful

---

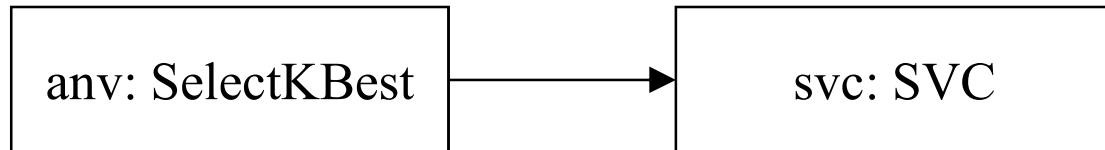
Name Reflection:  
Keras.na

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Reinterpreted  
Python: YAPS

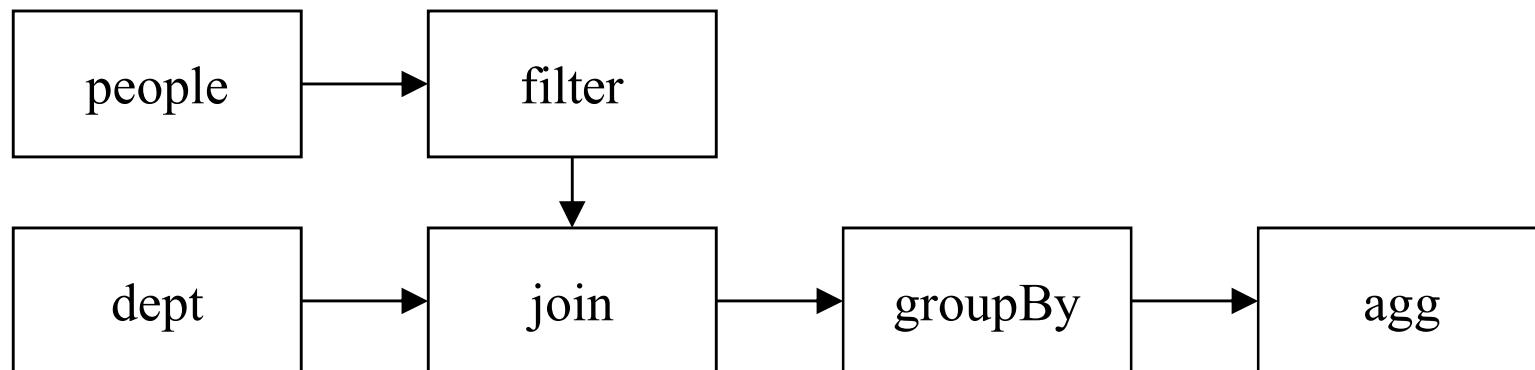
# Sklearn Example

```
1 filter = SelectKBest(f_regression, k=5)
2 clf = svm.SVC(kernel='linear')
3 anv_svm = Pipeline([('anv', filter), ('svc', clf)])
4 anv_svm.set_params(anv__k=10, svc__C=.1).fit(X, y)
```



# Spark SQL Example

```
1 people = sqlContext.read.parquet("people.parquet")
2 dept = sqlContext.read.parquet("dept.parquet")
3 people.filter(people.age > 30) \
4     .join(dept, people.deptId == dept.id) \
5     .groupBy(dept.name, "gender") \
6     .agg({"salary": "avg", "age": "max"})
```

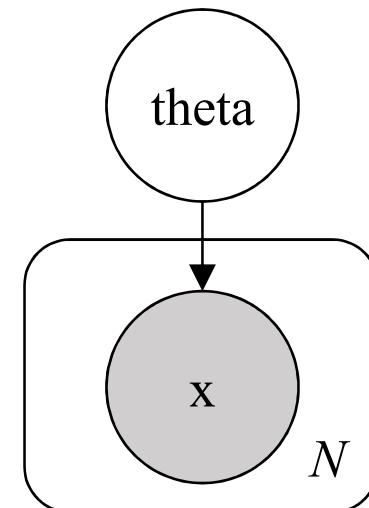
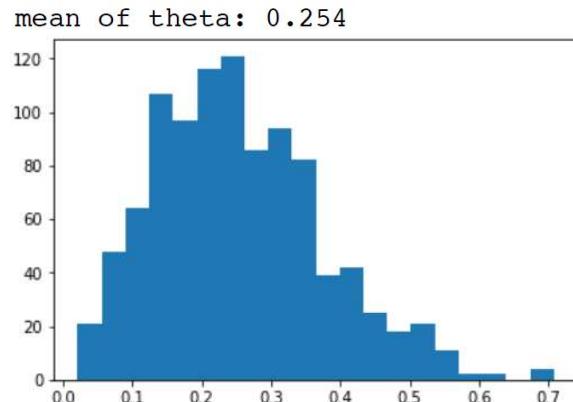


# PyStan Example

```
In [2]: model = Model(''  
    data {  
        int<lower=0,upper=1> x[10];  
    }  
    parameters {  
        real<lower=0,upper=1> theta;  
    }  
    model {  
        theta ~ uniform(0,1);  
        for (i in 1:10)  
            x[i] ~ bernoulli(theta);  
    }'')  
1  
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10  
11  
12
```

```
In [3]: flips = np.array([0, 1, 0, 0, 0, 0, 0, 0, 0, 1])  
run = model.sample(data={'x': flips}, random_=  
1  
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11  
12
```

```
In [4]: theta = run['theta']  
plt.hist(theta, bins=20)  
print('mean of theta: {:.3f}'.format(theta.mean()))  
1  
2  
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11  
12
```



# Outline

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Harmful

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Name Reflection:  
Keras.na

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Reinterpreted  
Python: YAPS

# Keras with Fewer Strings

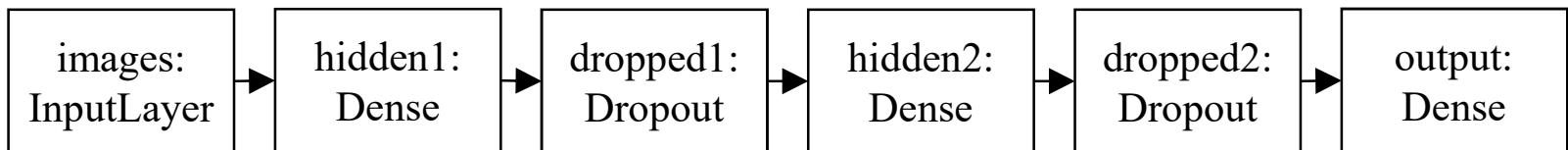
```
In [2]: import keras
        from keras.models import Model
        from keras.layers import Dense, Dropout, Input
        keras_utils.wrap_layers(globals())

        images = Input(shape=(N_PIXELS,))
        hidden1 = Dense(512, activation='relu')(images)
        dropped1 = Dropout(0.2)(hidden1)
        hidden2 = Dense(512, activation='relu')(dropped1)
        dropped2 = Dropout(0.2)(hidden2)
        output = Dense(N_LABELS, activation='softmax')(dropped2)

        model = Model(inputs=images, outputs=output)
```

```
In [3]: keras_utils.plot_model(model)
```

Out[3]:



```
In [5]: model.compile(loss='categorical_crossentropy',
                    optimizer='RMSprop',
                    metrics=['accuracy'])
        model.fit({images: x_train}, {output: y_train},
                  batch_size=128, epochs=3,
                  validation_data=(x_test, y_test))
```

# Name Reflection

```
1 import traceback
2 tb = traceback.extract_stack()
3 file_name, line_number, function_name, text = tb[-2]
4 import ast
5 tree = ast.parse(text, file_name)
6 assert isinstance(tree, ast.Module)
7 if isinstance(tree.body[0], ast.Assign):
8     targets = tree.body[0].targets
9     if len(targets) == 1 and isinstance(targets[0], ast.Name):
10        name = targets[0].id
```

# Naming Heuristics

Case	Example	Name mapping
If explicit name argument	<code>aux_out = Dense(..., name='output2') (input)</code>	output2
Else if assigned to variable with unique name	<code>aux_out = Dense(...) (input)</code>	aux_out
Else	<code>h = Dense(...) (input) h = Dense(...) (h)</code>	h dense_2

# Reverse Name Mapping

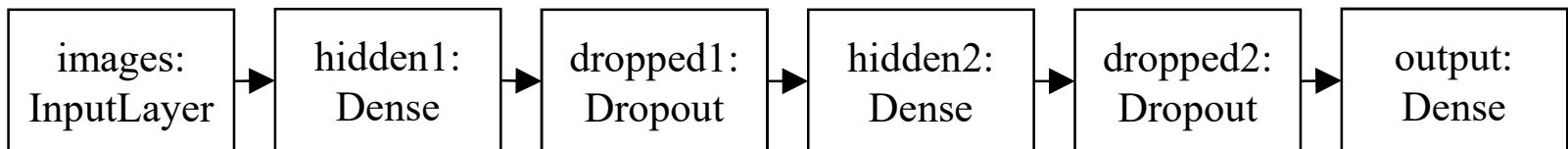
```
In [2]: import keras
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from keras.layers import Dense, Dropout, Input
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hidden2 = Dense(512, activation='relu')(dropped1)
dropped2 = Dropout(0.2)(hidden2)
output = Dense(N_LABELS, activation='softmax')(dropped2)

model = Model(inputs=images, outputs=output)
```

```
In [3]: keras_utils.plot_model(model)
```

Out[3]:



```
In [5]: model.compile(loss='categorical_crossentropy',
                    optimizer='RMSprop',
                    metrics=['accuracy'])
model.fit({images: x_train}, {output: y_train},
          batch_size=128, epochs=3,
          validation_data=(x_test, y_test))
```

- Names images, output are in scope (dict key does not start scope)
- Model wrapper re-keys the dictionary with strings from reflection

# Keras.na Error Messages

```
In [1]: import keras
        from keras.models import Model
        from keras.layers import Input, Conv2D
        import keras_utils
        keras_utils.wrap_layers(globals())
1
2
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        img = Input(shape=(28,28))
        conv = Conv2D(filters=10, kernel_size=3)(img)
        model = keras.models.Model(inputs=img, outputs=conv)
        model.compile(optimizer='sgd', loss='mean_squared_error')
```

Using TensorFlow backend.

**ValueError:** Input 0 is incompatible with layer conv: expected ndim=4, found ndim=3

# Outline

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Strings Considered  
Harmful

---

Name Reflection:  
Keras.na

---

Reinterpreted  
Python: YAPS

# Why Stan Probabilistic PL

- Widely adopted in the sciences
- Hundreds of examples
- Text books
- Dedicated conference
- High-level and self-contained language

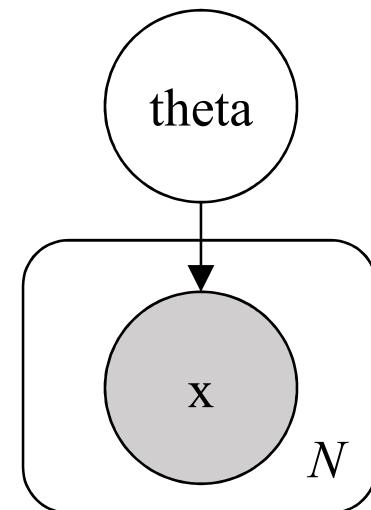
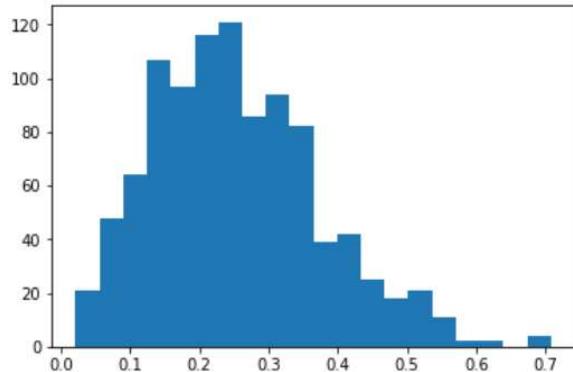
# YAPS Example

```
In [2]: @yaps.model
def coin(x: int(lower=0, upper=1)[10]):
    theta: real(lower=0, upper=1) <- uniform(0, 1)
    for i in range(1,11):
        x[i] <- bernoulli(theta)
```

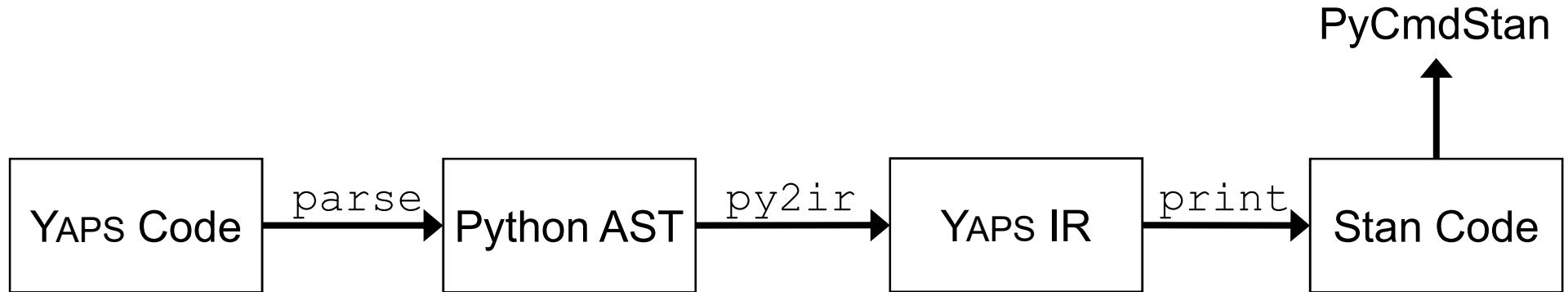
```
In [3]: flips = np.array([0, 1, 0, 0, 0, 0, 0, 0, 0, 1])
coin_obs = coin(x=flips)
coin_obs.sample(data=coin_obs.data, random_='seed=42')
```

```
In [4]: theta = coin_obs.posterior.theta
plt.hist(theta, bins=20)
print('mean of theta: {:.3f}'.format(theta.mean()))
```

mean of theta: 0.254



# Reinterpreted Python



- User interacts with Yaps just like with any other Python library.
- Decorator `@yaps.model` replaces function, internally triggers Stan code generation.
- PyCmdStan takes care of the rest, invoking external Stan and C++ compilers.

# Standard Python Tooling

```
@yaps.model
def coin(x: int(lower=0,
theta: real(lower=0,
for i in range(1, 11) Quick Fix... Peek Problem
    x[i] is bernoulli(theta)
```

# Reverse Source Mapping

```
In [2]: @yaps.model
1
def coin(x: int(lower=0, upper=1)[10]):
2
    for i in range(1,11):
3
        x[i] <~ bernoulli(theta)
4
```

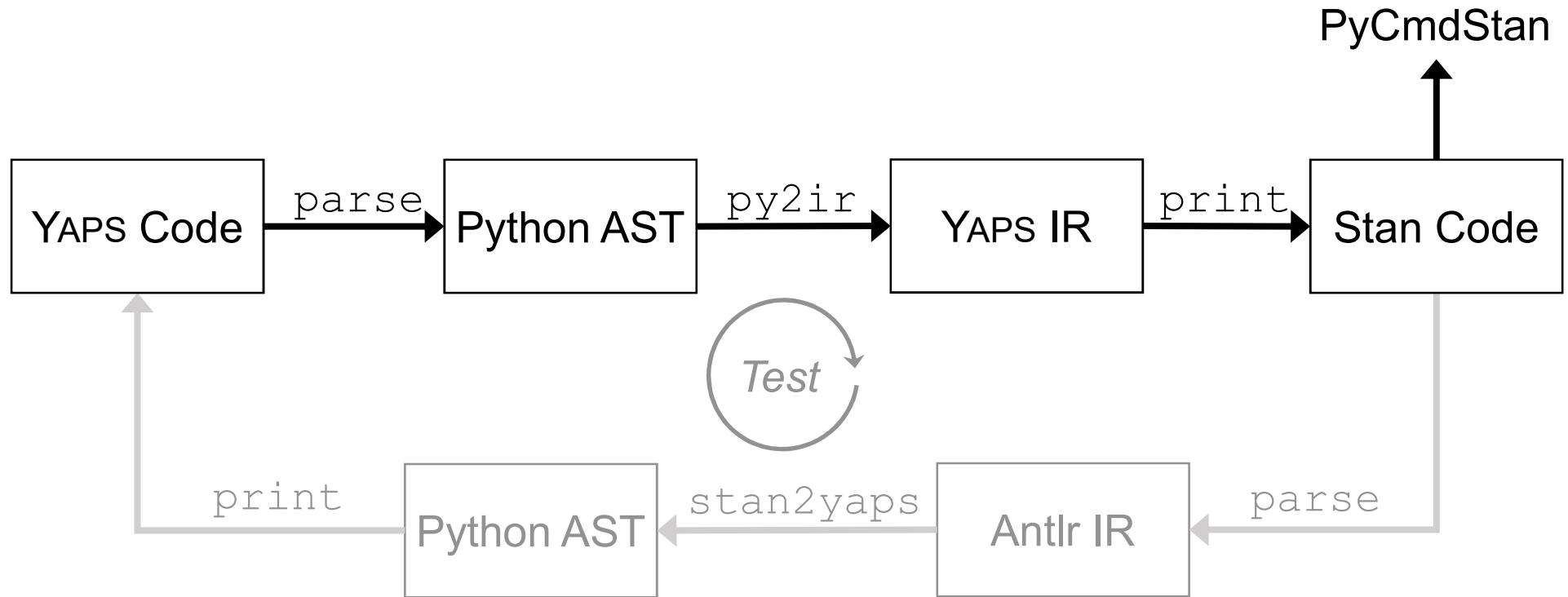
```
In [3]: import numpy as np
1
flips = np.array([0, 1, 0, 0, 0, 0, 0, 0, 0, 1])
2
coin_obs = coin(x=flips)
3
coin_obs.sample(data=coin_obs.data)
4
```

SYNTAX ERROR, MESSAGE(S) FROM PARSER:

```
variable "theta" does not exist.
error in '/Users/[REDACTED]/.cache/pycmdstan/model-3aed19a9.stan'
' at line 4, column 27
-----
```

```
1: @yaps.model
2: def coin(x: int(lower=0, upper=1)[10]):
3:     for i in range(1,11):
4:         x[i] <~ bernoulli(theta)
^
-----
```

# Roundtrip Validation



	Total	Success
User manual	61	100%
Dev repo	721	97%
Examples repo	500	82%

The failed cases used deprecated Stan syntax.

# YAPS Discussion

- Runtime validation
  - 13 example Stan models with datasets
  - Translated to YAPS, then back to Stan
  - Ran inference, found identical results
- Limitations
  - Typevar declaration for dependent types
  - Name clash with Python and Stan keywords
  - Debugger integration
- Open-source
  - <https://github.com/ibm/yaps/>
  - <https://pypi.org/project/yaps/>

# Conclusions

Strings attached	No strings attached	Where to find
Keras	Keras.na	This paper
PyStan	YAPS	This paper; github
Sklearn	LALE	arXiv:1906.03957
Spark SQL	?	
?		