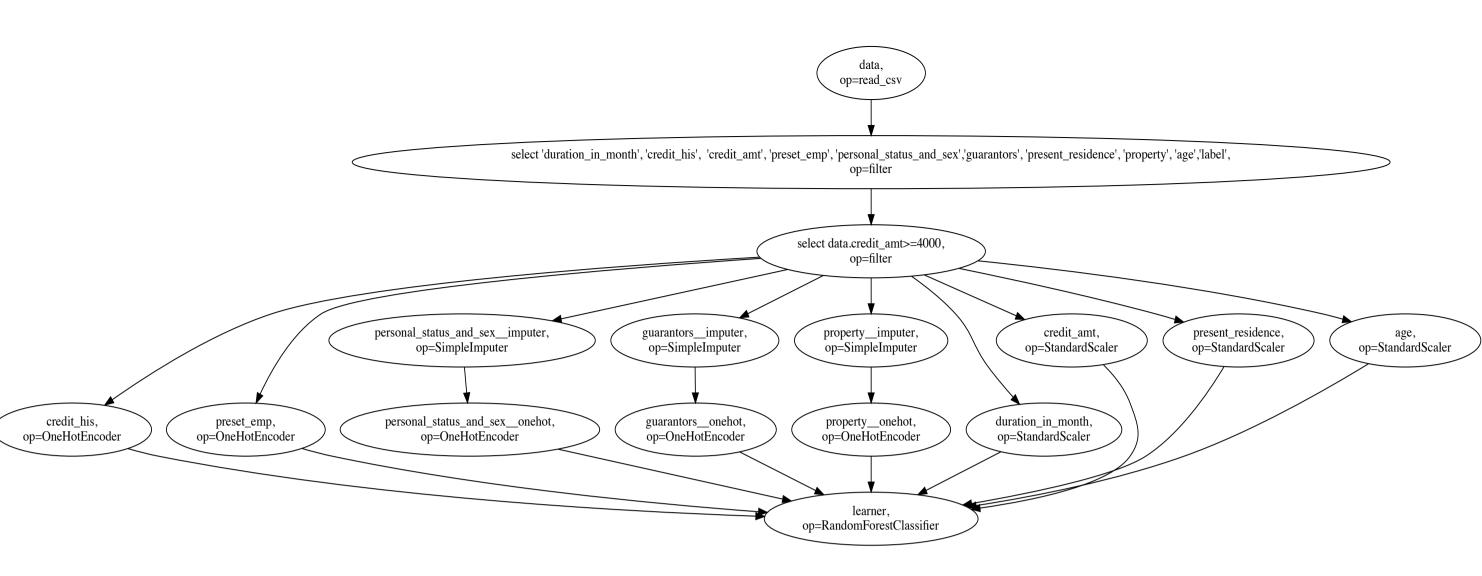
```
@tracer(cat_col = ['personal_status_and_sex'], numerical_col = ['age'])
def german_pipeline_easy(f_path = '../data/german_titled.csv'):
   data = pd.read_csv(f_path)
   # projection
   data = data[['duration_in_month', 'credit_his', 'credit_amt', 'preset_emp', 'personal_status_and_sex',
                 'guarantors', 'present_residence', 'property', 'age', 'label']]
   # filterina
   data = data.loc[(data.credit_amt>=4000)]
   #start sklearn pipeline
   one_hot_and_impute = Pipeline([
        ('imputer', SimpleImputer(strategy='most_frequent')),
        ('onehot', OneHotEncoder())
   featurizer = ColumnTransformer(transformers=[
        ('onehot', OneHotEncoder(), ['credit_his', 'preset_emp']),
        ('impute_onehot', one_hot_and_impute, ['personal_status_and_sex', 'guarantors', 'property']),
       ('std_scaler', StandardScaler(), ['duration_in_month', 'credit_amt', 'present_residence', 'age'])
   pipeline = Pipeline([
        ('features', featurizer),
        ('learner', RandomForestClassifier())
   return pipeline
```



	missing_count	num_class	class_count	class_percent
personal_status_and_sex	0.0	0.0	{'A93': -384, 'A92': -251, 'A91': -37, 'A94': -82}	{'A93': 0.1187, 'A92': -0.0702, 'A91': 0.0028, 'A94': -0.0513}
******** 		.credit a	mt>=4000)]	

Operations SimpleImputer on personal_status_and_sex

Operations OneHotEncoder on personal_status_and_sex

Changes in categorical features!

personal_status_and_sex

 missing_count
 0

 num_class
 -2

 class_count
 {0.0: 233, 1.0: 13}

 class_percent
 {0.0: 0.9472, 1.0: 0.0528}

Operations StandardScaler on age

Changes in numerical features!

	age
count	0.0000
missing_count	0.0000
median	-33.7344
mad	-10.1331
range	-50.1208
