

Creating dummies

INTERMEDIATE PREDICTIVE ANALYTICS IN PYTHON



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Motivation for creating dummy variables (1)

Logistic regression: $\text{logit}(a_1x_1 + a_2x_2 + \dots + a_nx_n + b)$

| donor_id | gender | country | segment |
|----------|--------|---------|---------|
| 5 | F | India | Gold |
| 3 | M | USA | Silver |
| 2 | M | India | Bronze |
| 8 | F | UK | Silver |
| 1 | F | USA | Bronze |

Motivation for creating dummy variables (2)

Logistic regression: $\text{logit}(a_1x_1 + a_2x_2 + \dots + a_nx_n + b)$

| donor_id | gender | country | segment | gender_F | gender_M |
|----------|--------|---------|---------|----------|----------|
| 5 | F | India | Gold | 1 | 0 |
| 3 | M | USA | Silver | 0 | 1 |
| 2 | M | India | Bronze | 0 | 1 |
| 8 | F | UK | Silver | 1 | 0 |
| 1 | F | USA | Bronze | 1 | 0 |

Preventing Multicollinearity (1)

| donor_id | gender | gender_F | gender_M |
|----------|--------|----------|----------|
| 5 | F | 1 | 0 |
| 3 | M | 0 | 1 |
| 2 | M | 0 | 1 |
| 8 | F | 1 | 0 |
| 1 | F | 1 | 0 |

Preventing Multicollinearity (2)

| donor_id | gender | gender_F |
|----------|--------|----------|
| 5 | F | 1 |
| 3 | M | 0 |
| 2 | M | 0 |
| 8 | F | 1 |
| 1 | F | 1 |

Preventing Multicollinearity (3)

| donor_id | country | country_USA | country_India | country_UK |
|----------|---------|-------------|---------------|------------|
| 5 | India | 0 | 1 | 0 |
| 3 | USA | 1 | 0 | 0 |
| 2 | India | 0 | 1 | 0 |
| 8 | UK | 0 | 0 | 1 |
| 1 | USA | 1 | 0 | 0 |

Preventing Multicollinearity (4)

| donor_id | country | country_USA | country_India |
|----------|---------|-------------|---------------|
| 5 | India | 0 | 1 |
| 3 | USA | 1 | 0 |
| 2 | India | 0 | 1 |
| 8 | UK | 0 | 0 |
| 1 | USA | 1 | 0 |

Adding dummy variables in Python

```
donor_id segment
0      32770  Gold
1      32776  Silver
2      32777  Bronze
3      65552  Bronze
```

```
# Create the dummy variable
dummies_segment = pd.get_dummies(basetable["segment"], drop_first=True)

# Add the dummy variable to the basetable
basetable = pd.concat([basetable, dummies_segment], axis=1)

# Delete the original variable from the basetable
del basetable["segment"]
```

```
donor_id Gold Silver
0      32770    1     0
1      32776    0     1
2      32777    0     0
3      65552    0     0
```


Let's practice!

INTERMEDIATE PREDICTIVE ANALYTICS IN PYTHON

Missing values

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Replacing missing values by an aggregate (1)

| donor_id | age |
|----------|-----|
| 5 | - |
| 3 | 25 |
| 2 | 36 |
| 8 | 40 |
| 1 | 26 |

Replacing missing values by an aggregate (2)

| donor_id | age |
|----------|-----|
| 5 | 38 |
| 3 | 25 |
| 2 | 36 |
| 8 | 40 |
| 1 | 26 |

Mean age: 38

Replacing missing values by an aggregate (3)

| donor_id | max_donation |
|----------|--------------|
| 5 | - |
| 3 | 1 000 000 |
| 2 | 100 |
| 8 | 40 |
| 1 | 120 |

Mean `max_donation` : 25 065

Median `max_donation` : 110

Replacing missing values by an aggregate (4)

| donor_id | max_donation |
|----------|--------------|
| 5 | 110 |
| 3 | 1 000 000 |
| 2 | 100 |
| 8 | 40 |
| 1 | 120 |

Mean `max_donation` : 25 065

Median `max_donation` : 110

Replacing missing values by a fixed value (1)

| donor_id | sum_donations |
|----------|---------------|
| 5 | 130 |
| 3 | 10 |
| 2 | - |
| 8 | 40 |
| 1 | 120 |

Replacing missing values by a fixed value (2)

| donor_id | sum_donations |
|----------|---------------|
| 5 | 130 |
| 3 | 10 |
| 2 | 0 |
| 8 | 40 |
| 1 | 120 |

Replacing missing values in Python

```
# Replace missing values by 0
replacement = 0
basetable["donations_last_year"] =
    basetable["donations_last_year"].fillna(replacement)

# Replace missing values by mean
replacement = basetable["age"].mean()
basetable["age"] = basetable["age"].fillna(replacement)
```

Missing value dummies

```
donor_id email
0      32770 person32770@provider.com
1      32776 nan
2      32777 person32777@provider.com
3      65552 nan
```

```
basetable["no_email"] = pd.Series(
    [0 if email==email else 1
     for email in basetable["email"]])
```

```
donor_id email          no_email
0      32770 person32770@provider.com  0
1      32776 nan                      1
2      32777 person32777@provider.com  0
3      65552 nan                      1
```

Let's practice!

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Handling outliers

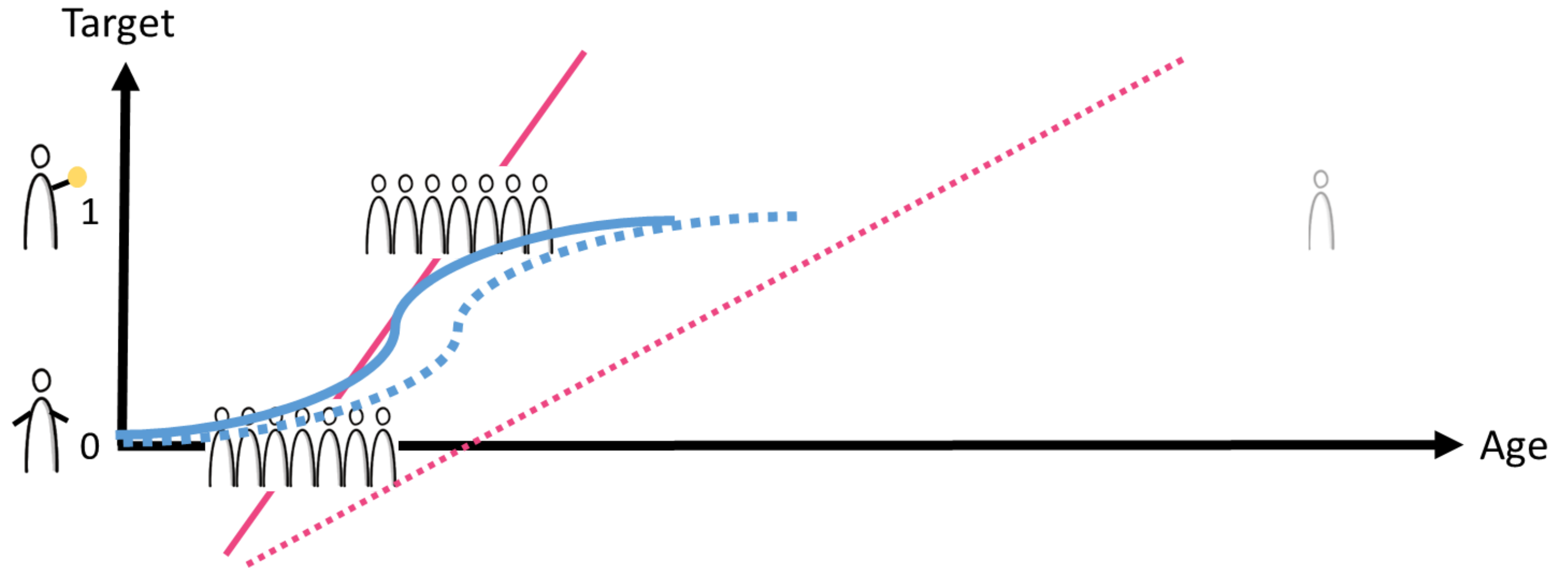
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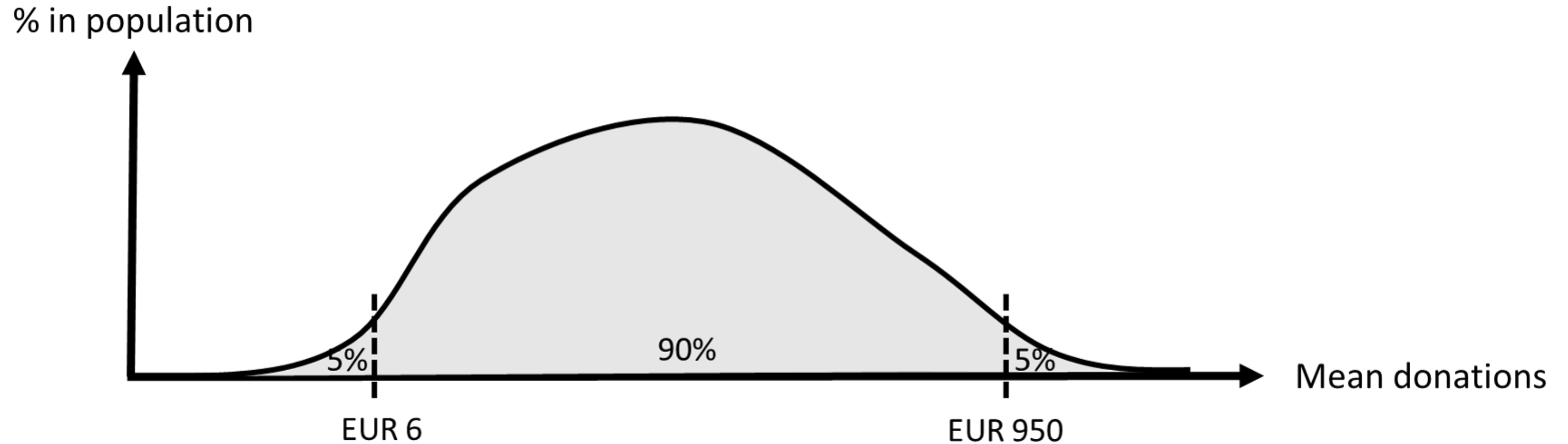
Influence of outliers on predictive models



Causes of outliers

- Human errors
- Measuring errors
- Truly extreme values
- ...

Winsorization concept



Winsorization in Python

```
from scipy.stats.mstats import winsorize
basetable["variable_winsorized"] =
    winsorize(
        basetable["variable"],
        limits = [0.05, 0.01])
```


Standard deviation method concept



Standard deviation method in Python

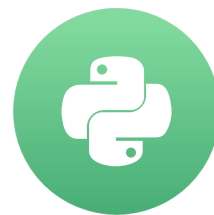
```
mean_age = basetable["age"].mean()
sd_age = basetable["age"].std()
lower_limit = mean_age - 3*sd_age
upper_limit = mean_age + 3*sd_age
basetable["age_no_outliers"] = pd.Series(
    [min(max(a, lower_limit), upper_limit)
     for a in basetable["age"]]
)
```

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Transformations

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Motivation for transformations

100€



Alice

1 100€



Bob

10 000€



Carol

11 000€

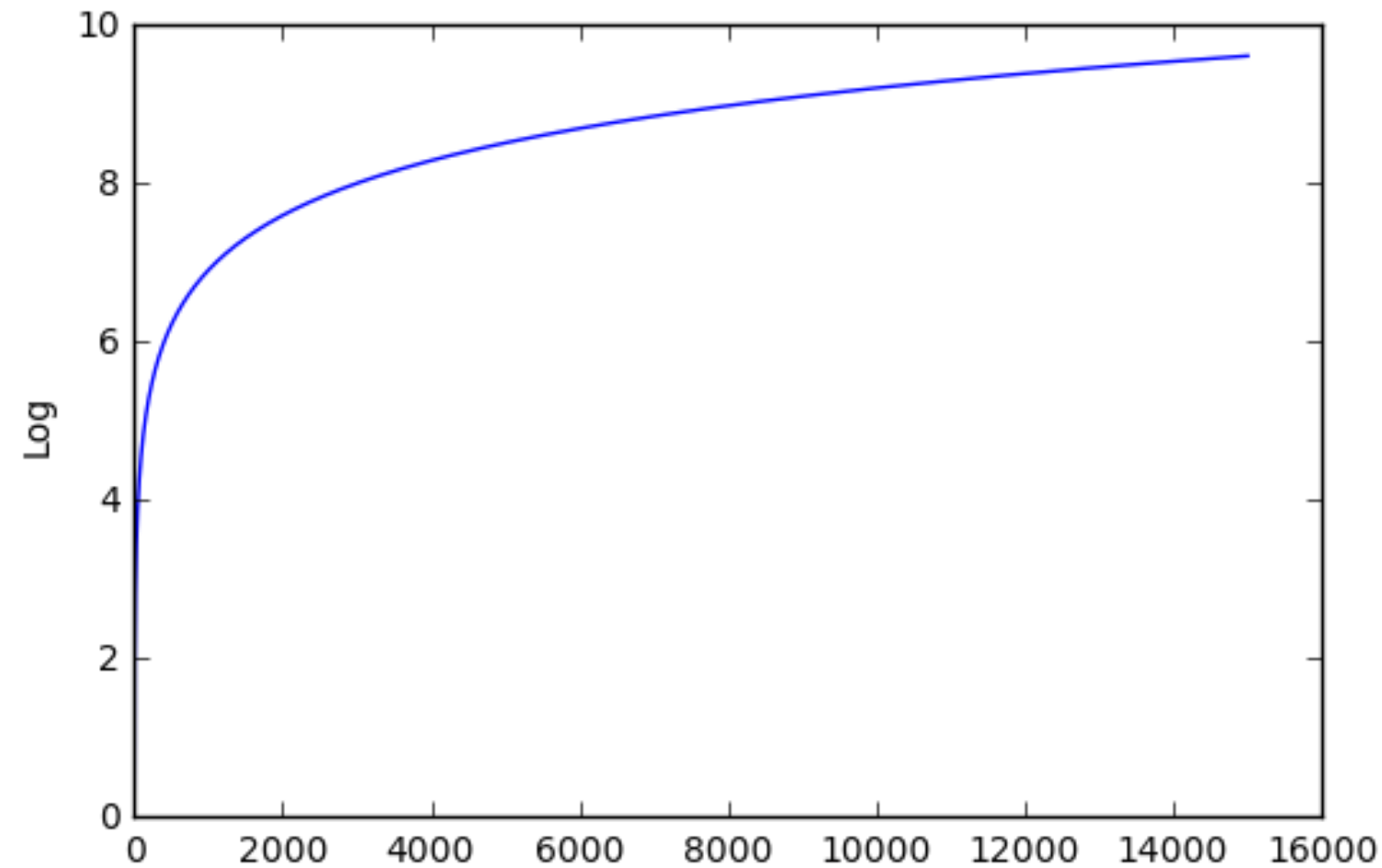


Dave

Log transformation



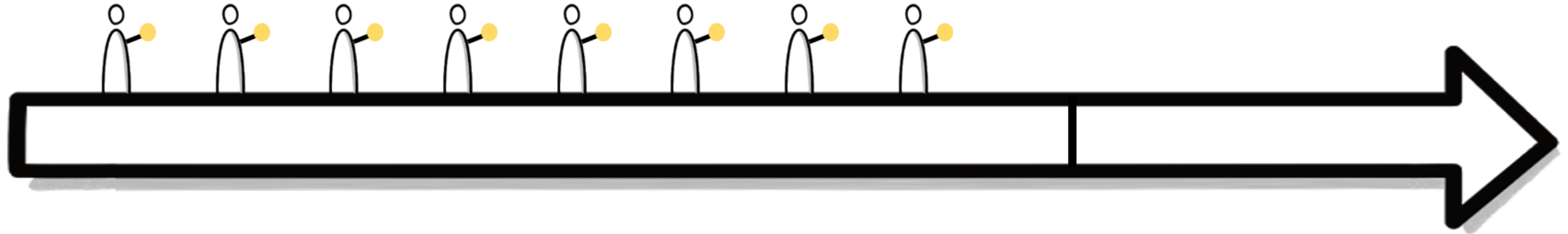
Log transformation



```
import numpy as np
basetable["log_variable"] = np.log(basetable["variable"])
```

Interactions

Likely to donate soon



Unlikely to donate soon



Interactions in Python

```
basetable["number_donations_int_recency"] =  
    basetable["number_donations"] * basetable["recency"]
```

Let's practice!

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