Name:

1 FA23 Midterm Problem 4

In this question, we will continue to work with the **donkeys** dataset from Problem 3. The first few rows of the table column descriptions are shown again below for convenience.

	id	BCS	Age	Weight	WeightAlt	id	A unique identifier for each donkey (d01, d02,	
0	d01	3.0	<2	77	NaN		etc.).	
1	d02	2.5	<2	100	NaN	BCS	Body condition score: from 1 (emaciated) to $(h_{12}(h_{12})) = (h_{12}(h_{12}))$	
2	d03	1.5	<2	74	NaN	Age	(healthy) to 5 (obese) in increments of 0.5. Age in years: <2 , $2-5$, $5-10$, $10-15$, $15-20$, and	
							over 20 years.	
						Weight	Weight in kilograms.	
						WeightAlt	Second weight measurement taken for 30 don- keys. NaN if the donkey was not reweighed	
						WeightAlt	second weight measurement taken for 30 don- keys. NaN if the donkey was not reweighed.	

Alan wants to see whether donkeys with $BCS \ge 3$ have larger Weight values on average compared to donkeys that have BCS < 3. To generate a single sample under his null hypothesis, Alan should (choose one):

- Resample 744 donkeys with replacement from donkeys.
- Resample 372 donkeys with replacement from donkeys with BCS < 3, and another 372 donkeys with $BCS \ge 3$.
- Randomly permute the Weight column.

Doris wants to use multiple imputation to fill in missing values in WeightAlt. She knows that WeightAlt is MAR on BCS and Age, so she will perform multiple imputation conditional on BCS and Age – each missing value will be filled in with values from a random WeightAlt value from a donkey with the same BCS and Age. Assume that all BCS and Age combinations have observed WeightAlt values. Fill in the blanks in the code below to estimate the median of WeightAlt using multiple imputation conditional on BCS and Age with 100 repetitions.

2 WI23 Final Exam Problem 1

The DataFrame sat contains one row for **most** combinations of Year and State, where Year ranges between 2005 and 2015 and State is one of the 50 states (not including the District of Columbia). Assume sat does not contain any duplicate rows — that is, there is only one row for every unique combination of Year and State that is in sat – and that sat does not contain any null values.

	Year	State	# Students	Math	Verbal
0	2014	Washington	41277	519	510
1	2013	Arizona	22283	529	522
2	2006	Kansas	2545	591	582
3	2011	North Dakota	219	612	586
4	2009	New Mexico	2209	548	553

The data description stated that there is one row in **sat** for most combinations of **Year** (between 2005 and 2015, inclusive) and **State**. It turns out that there are 11 rows in sat for all 50 states, except for one state. Fill in the blanks below so that **missing_years** evaluates to an array, sorted in any order, containing the years for which that one state does not appear in **sat**.

```
state_only = sat.groupby("State").filter(_____)
merged = sat["Year"].value_counts().to_frame().merge(
    state_only,
    _____)
missing_years =
    ______.
to_numpy()
```

The following DataFrame contains summary statistics for all SAT takers in New York and Texas from 2005 to 2015. Suppose we want to run a statistical test to assess whether the distributions of the number of students between 2005 and 2015 in New York and Texas are significantly different.

	mean	median	std	
State				
New York	157950.818182	157989.0	3430.986500	
Texas	155035.909091	148102.0	22509.092685	

Given the above DataFrame, which test statistic is **most likely** to yield a significant difference?

- A. mean number of students in Texas mean number of students in New York
- B. |mean number of students in Texas mean number of students in New York|
- C. |median number of students in Texas median number of students in New York|
- D. The Kolmogorov-Smirnov statistic