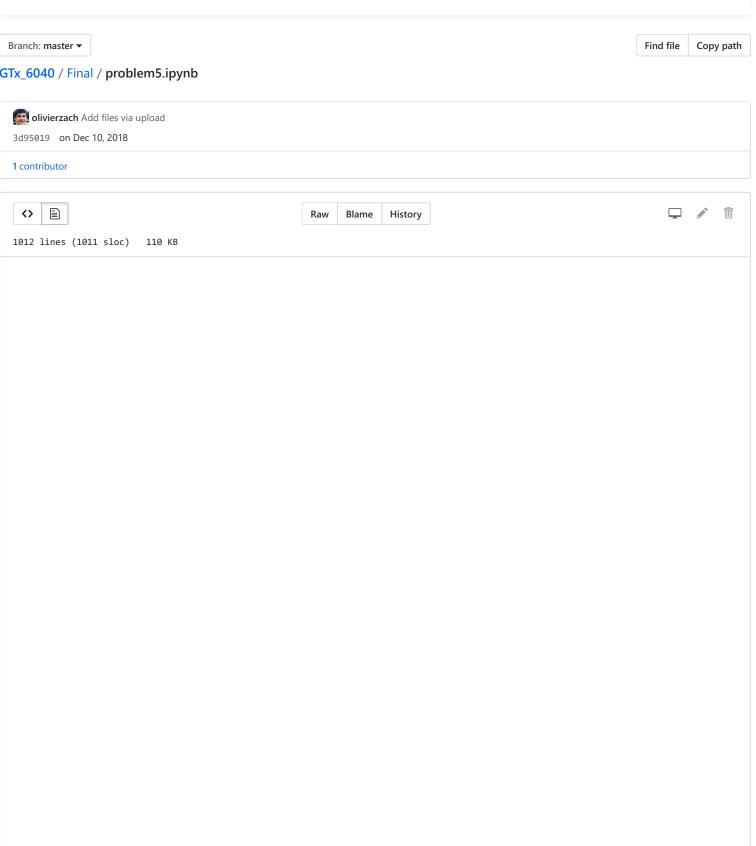
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Problem 5: Market-clearing prices

In this problem, you will simulate a simple economic market of buyers and sellers who wish to trade a certain product. This problem assesses your ability to translate a procedure into efficient code using elementary Python. There are two (2) exercises worth a total of ten (10) points.

This notebook includes some simple plots to help you understand and debug the output. Run the cell below now, which will define the plotting routines. (You do not need to understand the code inside this cell.)

```
In [2]: import matplotlib.pyplot as plt
        %matplotlib inline
        # Adapted from: https://matplotlib.org/gallery/lines bars and markers/barchart.html
        def autolabel(rects, xpos='center', ax=None):
            if ax is None:
                ax = plt.gca()
            xpos = xpos.lower() # normalize the case of the parameter
            ha = {'center': 'center', 'right': 'left', 'left': 'right'}
            offset = {'center': 0.5, 'right': 0.57, 'left': 0.43} # x_txt = x + w*off
            for rect in rects:
                 height = rect.get_height()
                 ax.text(rect.get_x() + rect.get_width()*offset[xpos], 1.01*height,
                          {}'.format(height), ha=ha[xpos], va='bottom')
        def viz market(market):
            x = range(len(market))
            y = [p for _, p in market]
            is_buyer = [t == 'buyer' for t, _ in market]
            colors = ['blue' if is_blue else 'gray' for is_blue in is_buyer]
            def split_filter(x, f):
                x true = [xi for xi, fi in zip(x, f) if fi]
                 x_false = [xi for xi, fi in zip(x, f) if not fi]
                 return x true, x false
            x_buyers, x_sellers = split_filter(x, is_buyer)
            y_buyers, y_sellers = split_filter(y, is_buyer)
            buyer_bars = plt.bar(x_buyers, y_buyers, color='blue', label='buyers')
            seller bars = plt.bar(x sellers, y sellers, color='lightgray', label='sellers')
            plt.xlabel('Person ID')
            plt.title('Price ($)')
            plt.legend()
            autolabel(buyer_bars)
            autolabel(seller_bars)
        def fn(fn_base, dirname=None):
            from os.path import isdir
            if dirname is None:
                 if isdir('.voc'):
                    dirname = '../resource/asnlib/publicdata/'
                 else:
                    dirname = ''
            assert isdir(dirname)
            return '{}{}'.format(dirname, fn_base)
```

A simple economic market

Consider the following model of an (economic) market.

Sellers and buyers. Suppose there are *n* people, who wish to trade some product, like coffee mugs. The people come in two types: **sellers**, who have a coffee mug and wish to sell it, and **buyers**, who do not have coffee mugs but wish to acquire one.

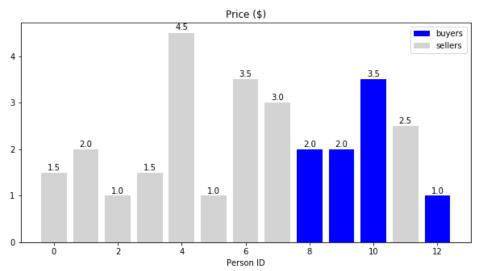
Product values and prices. Each person values a coffee mug differently. That means each seller wants to sell his or her coffee mug at one price, while each buyer wants to buy at (possibly) some other price. These prices might differ from person to person, that is, different sellers want to sell at different prices, and similarly for buyers.

Example. Suppose n = 13 people, of whom $n_s = 9$ are sellers and $n_b = 4$ buyers. Each seller and buyer values the product, a single coffee mug, at some price.

The code cell below defines this market, which is stored in a list named market_demo. Each person i is represented by market_demo[i] = (who_i, price_i), a tuple where who_i is either 'seller' or 'buyer', and price_i is the value that person i has assigned to a coffee mug that he or she wishes to sell or buy.

Run this code cell now to "create" this market.

```
[('seller', 1.5), ('seller', 2.0), ('seller', 1.0), ('seller', 1.5), ('seller', 4.5), ('seller', 1.0), ('seller', 3.5), ('seller', 3.0), ('buyer', 2.0), ('buyer', 2.0), ('buyer', 3.5), ('seller', 2.5), ('buyer', 1.0)]
```



Observe that there are 9 sellers, shown as light gray bars, and 4 buyers, shown as darker blue bars. Each bar is labeled by the corresponding person's price. For example, persons 0 and 3 wish to sell their coffee mugs for \$1.50, whereas person 10 is willing to pay \$3.50, while person 12 is only willing to pay \$1.00.

The market-clearing price

Now suppose buyers and sellers start trading their coffee mugs, according to the following procedure:

- 1. Initially, all buyers and sellers are "unmatched."
- 2. Pick any unmatched buyer with the highest price. Similarly, pick any unmatched seller with the lowest price.
- 3. A *match* occurs between these two **if** the buyer's price is at least as large as the seller's price. If there is a match, they exchange coffee mugs. This transaction is also called a *trade*. Once these two people have executed this trade, they are no longer part of the market.
- 4. Repeat this matching process (steps 2 and 3) among the remaining buyers and sellers, until no matches remain.

What if there is a tie, meaning there are multiple unmatched buyers with the same highest price or multiple unmatched sellers with the same lowest price? In this case, the buyer or seller is selected arbitrarily.

For instance, go back to the market_demo example.

- · Initially, there are no matches.
- The buyer with the highest price is Person 10, whose value is \$3.50.
- There are two sellers with the same lowest price, Persons 2 and 5, who both value coffee mugs at \$1.00. Pick either one; let's say, Person 2. Thus, Persons 10 and 2 are "matched." They drop out of the market.
- Among remaining buyers, both Persons 8 and 9 have the highest price, who each have a value of \$2.00. So one of these will be

matched against Person 5; let's say it's Person 8. Persons 5 and 8 are matched and drop out.

• Then Person 9 will be matched against either Persons 0 or 3, who have the same price of \$1.50; let's say it's Person 0.

Here is what we have so far:

Buyer \\$	Seller \\$	Who?
3.5	1.0	10 ← 2
2.0	1.0	8 ← 5
2.0	1.5	9 = 0

As it happens, that is the last possible trade in this market! Person 12 is the only unmatched buyer, but his or her value is only \$1.00. By contrast, the next unmatched seller is Person 3 with a value of \$1.50. Since the asking price of \$1.50 exceeds the buyer's price of \$1.00, they cannot trade. We say the market has **reached equilibrium**.

The (market-)clearing price. Looking at the trades, consider the highest selling price, which in the preceding example was \$1.50. We refer to this value as the *market-clearing price*, or just *clearing price*: it is the selling price at which the market has reached an equilibrium and no further trades are possible.

There are other possible definitions of clearing price, but for this problem, please use this one.

Exercises

Exercise 0 (2 points). You do not need to write any code in this exercise. However, you do need to read some code, for which you get a "almost-free" 2 points! (You **must** submit the problem to the autograder to get these two points.)

To help you get started, we are giving you one function called p, $n = analyze_market(m)$ that implements the procedure above. Given a market m, it returns two values: the clearing price p and the number of trades t. If no trades are possible at all, it returns p=0 and t=0 (both integers, even though p could be fractional in general).

Read the code for analyze_market() and convince yourself that it implements the trading procedure described previously. As a reminder (and to save you some scrolling), here is that trading procedure, repeated verbatim:

- 1. Initially, all buyers and sellers are "unmatched."
- 2. Pick any unmatched buyer with the highest price. Similarly, pick any unmatched seller with the lowest price.
- 3. A match occurs between these two if the buyer's price is at least as large as the seller's price. If there is a match, they exchange coffee mugs. This transaction is also called a trade. Once these two people have executed this trade, they are no longer part of the market.
- 4. Repeat this matching process (steps 2 and 3) among the remaining buyers and sellers, until no matches remain.

What if there is a tie, meaning there are multiple unmatched buyers with the same highest price or multiple unmatched sellers with the same lowest price? In this case, the buyer or seller is selected arbitrarily.

```
In [4]: # Test cell: `ex0_freebie` (2 points)
        def analyze market(market, verbose=False):
            buy_prices = [price for who, price in market if who == 'buyer']
            sell_prices = [price for who, price in market if who == 'seller']
            trades = [] # Tracks trades
            unmatched = buy_prices and sell_prices
            while unmatched:
                i buyer = buy prices.index(max(buy prices))
                i_seller = sell_prices.index(min(sell_prices))
                if buy_prices[i_buyer] >= sell_prices[i_seller]: # A match!
                    trades.append((buy_prices[i_buyer], sell_prices[i_seller]))
                    del buy_prices[i_buyer]
                    del sell_prices[i_seller]
                     unmatched = buy_prices and sell_prices
                    unmatched = False # Stops trading
            if verbose: print(trades)
            if trades:
                return trades[-1][1], len(trades)
```

```
return int(0), int(0)

clearing_price_demo, num_trades_demo = analyze_market(market_demo, verbose=True)
print("The clearing price is ${:.2f}.".format(clearing_price_demo))
print("There were {} trades.".format(num_trades_demo))

print("\n(Passed!)")

[(3.5, 1.0), (2.0, 1.0), (2.0, 1.5)]
The clearing price is $1.50.
There were 3 trades.

(Passed!)
```

Creating a random market. For the next few exercises, we'll need a function that can create a random market.

The function create_random_market(num_people, prob_buyer, max_price) will randomly generate a market in the form of a list of tuples formatted just like market_demo, above. Its parameters are:

- num_people: The number of people in the market
- prob_buyer: The probability that a given person should be a buyer, rather than a seller.
- max_price: The maximum value that can be assigned to any buyer or seller.

Each value (price) will be an integer drawn uniformly at random from the *closed* interval [1, max_price], that is, inclusive of 1 and max price.

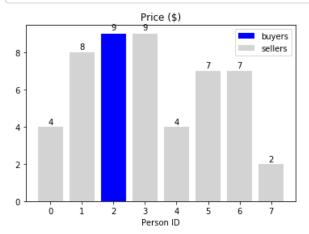
Recall that market_demo looked like the following (run the next cell):

And here is the create_random_market function; as a quick demo, run it a few times.

```
In [6]: def random_person(prob_buyer, max_price):
    from random import uniform, randrange
    who = 'buyer' if uniform(0, 1) < prob_buyer else 'seller'
    price = randrange(1, max_price+1)
    return (who, price)

def create_random_market(num_people, prob_buyer, max_price):
    return [random_person(prob_buyer, max_price) for _ in range(num_people)]

# Demo:
market_demo2 = create_random_market(8, 0.3, 10)
viz_market(market_demo2)</pre>
```



Exercise 1 (2 points). Next, you will implement a function that simulates market trading.

In particular, you will define this function:

```
def simulate(prob_buyer, num_people, max_price, num_trials):
```

. . .

One input is num trials, which is the number of simulation trials to run. In each trial, your code should:

- Randomly generate a market with num_people people, where the probability of being a buyer is prob_buyer and the maximum price for any buyer or seller is max_price, drawn uniformly at random from 1 to max_price, inclusive. *Hint: Use a function that appears earlier in this notebook.*
- Call analyze_market() to analyze that market.
- Record the clearing price and number of trades as a tuple (pair).
- · Return all of these pairs in a list.

For example,

```
simulate(0.5, 8, 10, 5)
```

might return the list of pairs, [(10, 2), (3, 3), (10, 2), (9, 1), (15, 2)], which has one entry per trial and 5 trials in all, and each entry is a (clearing price, number of trades) pair. (This is just an example of the format of the output; since the markets will be generated randomly, you will see different values.)

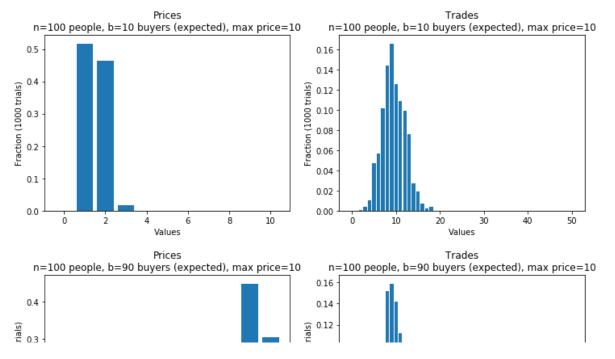
There are two test cells, so you can get partial credit. The first is just a quick demo; if your simulation is on the right track, you should see that when there are many more sellers than buyers, then the clearing price is low; and in the opposite scenario, the clearing price will be high.

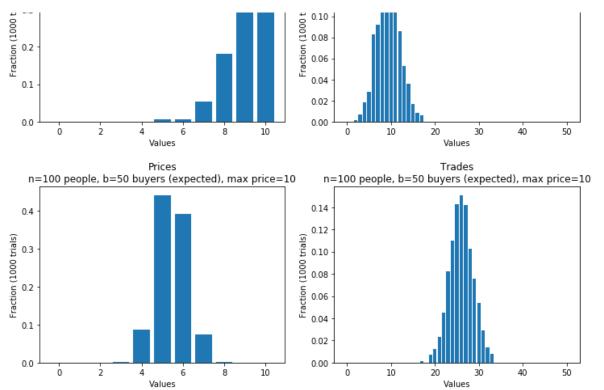
```
In [10]: def simulate(prob_buyer, num_people, max_price, num_trials):
    sims = []
    for i in range(num_trials):
        c = create_random_market(num_people, prob_buyer, max_price)
        clearing_price, num_trades = analyze_market(c)
        sims.append((clearing_price, num_trades))
    return sims
```

```
In [11]: | # Test cell 0: `ex1_0_quick_run` (1 point)
          def avg clearing price(sim results):
             return sum([price for price, _ in sim_results]) / len(sim_results)
          def avg_trades(sim_results):
             return sum([trades for _, trades in sim_results]) / len(sim_results)
          def viz counts(values, counts, norm=True, ax=None, title=''):
             if ax is None: ax = plt.gca()
             if norm:
                 s = sum(counts)
                 y = [c/s for c in counts]
             else:
                 y = counts
             bars = ax.bar(values, y)
             ax.set xlabel('Values')
             ax.set_ylabel('Fraction ({} trials)'.format(s) if norm else 'Counts')
             ax.set_title(title)
             return bars
          def get counts sorted by value(x, values=None):
             from collections import Counter
             counts_table = Counter(x)
             if values is None:
                  pairs = sorted(counts_table.items(), key=lambda x: x[0])
                  values = [v for v, _ in pairs]
                 counts = [c for _, c in pairs]
             else:
                  values = list(values)
                  counts = [counts_table.get(v, 0) for v in values]
             return values, counts
         NUM_PEOPLE = 100
         MAX_PRICE = 10
         NUM_TRIALS = 1000
         CASES_AVG_BUYERS = [10, 90, 50]
```

```
all_results = []
for avg_buyers in CASES__AVG_BUYERS:
    results = simulate(avg_buyers/NUM_PEOPLE, NUM_PEOPLE, MAX_PRICE, NUM_TRIALS)
    print("""
______
Consider a market of {} people, of whom about {} are buyers on average.
Suppose a given product is worth at most ${} to any buyer or seller.
Then, {} trials of your simulation suggests that ...
""".format(NUM_PEOPLE, avg_buyers, MAX_PRICE, NUM_TRIALS))
   print("* the average clearing price is ${}; and".format(avg_clearing_price(results)))
    print("* the average number of trades is {}.".format(avg trades(results)))
    all results.append(results)
    price_values, price_counts = get_counts_sorted_by_value([p for p, _ in results],
                                                          values=range(0, MAX_PRICE+1))
    trade_values, trade_counts = get_counts_sorted_by_value([t for _, t in results],
                                                          values=range(0, (NUM_PEOPLE+2)//2))
    fig, ax = plt.subplots(1, 2, figsize=(12, 4))
    subtitle = '\nn={} people, b={} buyers (expected), max price={}'.format(NUM_PEOPLE, avg buyers,
MAX PRICE)
    viz counts(price values, price counts, ax=ax[0], title='Prices{}'.format(subtitle))
    viz_counts(trade_values, trade_counts, ax=ax[1], title='Trades{}'.format(subtitle))
print("\n(Passed!)")
_____
Consider a market of 100 people, of whom about 10 are buyers on average.
Suppose a given product is worth at most $10 to any buyer or seller.
Then, 1000 trials of your simulation suggests that ...
* the average clearing price is $1.501; and
* the average number of trades is 9.507.
Consider a market of 100 people, of whom about 90 are buyers on average.
Suppose a given product is worth at most $10 to any buyer or seller.
Then, 1000 trials of your simulation suggests that ...
* the average clearing price is $8.975; and
* the average number of trades is 9.386.
Consider a market of 100 people, of whom about 50 are buyers on average.
Suppose a given product is worth at most $10 to any buyer or seller.
Then, 1000 trials of your simulation suggests that ...
* the average clearing price is $5.452; and
* the average number of trades is 26.043.
```

(Passed!)





```
In [12]: # Test cell: `ex1_1_random_tests` (1 point)
                        def check_dist(dist, obs, title=None):
                                 from scipy.stats import kstest, ks_2samp
                                  from numpy import array
                                  exp_obs = []
                                  for v, c in zip(dist.index, dist['freq']):
                                            exp_obs += [v] * int(c * len(obs))
                                  D, p = ks_2samp(obs, array(exp_obs))
                                  if title is not None: # Verbose mode
                                            print("{}: D={}, p={}".format(title, D, p))
                                  \textbf{assert} \ p \ > \ 0.1, \ \textbf{"There is something fishy about the values produced by your simulation. Keep try}
                        ing! (D={}, p={})".format(D, p)
                        def read_dist(filepath):
                                  from pandas import read_csv
                                  pmf = read_csv(filepath)
                                  cdf = pmf.set_index('value')
                                  cdf['cfreq'] = cdf['freq'].cumsum()
                                  return cdf
                        def check_sim_results(avg_buyers, results, title=None):
                                  prices_dist = read_dist(fn('prices--n{}--eb{}--p{}.csv'.format(NUM_PEOPLE, avg_buyers, MAX_PRICE
                        )))
                                 prices_obs = [p for p, _ in results]
                                  check_dist(prices_dist, prices_obs, title)
                                  trades\_dist = read\_dist(fn('trades--n\{\}--p\{\}.csv'.format(NUM\_PEOPLE, avg\_buyers, MAX\_PRICE, avg\_buyers, Avg\_buy
                       )))
                                  trades_obs = [t for _, t in results]
                                  check_dist(trades_dist, trades_obs, title)
                        NUM PEOPLE = 100
                       MAX PRICE = 10
                        NUM_TRIALS = 1000
                        CASES_AVG_BUYERS = [10, 90, 50]
                        for avg_buyers in CASES__AVG_BUYERS:
                                  results = simulate(avg_buyers/NUM_PEOPLE, NUM_PEOPLE, MAX_PRICE, NUM_TRIALS)
                                  assert len(results) == NUM_TRIALS, "Did not return the correct number of results."
                                  check_sim_results(avg_buyers, results, title='avg_buyers={}'.format(avg_buyers))
                        print("\n(Passed!)")
```

```
avg_buyers=90: D=0.01769477911646586, p=0.9974383845541924
avg_buyers=90: D=0.019475806451612843, p=0.9910140159191321
avg_buyers=50: D=0.008189568706118311, p=0.99999999999976
avg_buyers=50: D=0.010474268415741728, p=0.999999997885006
(Passed!)
```

Timing. Let's measure how long it takes to run analyze market() for a "large" market, i.e., one with many people.

```
In [15]: market_for_timing = create_random_market(20000, 0.5, 100)
%timeit analyze_market(market_for_timing)
1.33 s ± 4.8 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
```

Exercise 2 (6 points). The analyze market() function is slow. Come up with a more efficient implementation.

There will be both a correctness test (1 point) and a timing test (5 points). The timing test below will check that your implementation is at least **50 times faster** than analyze_market() on an input comparable to market_for_timing, as generated above. There is no partial credit for anything that does not beat this threshold.

Hint. You may be tempted to start with the analyze_market() but replace lists with Numpy arrays. While that may work (we didn't try it), our sample solution handily beats the target threshold using only standard Python (no Numpy!) but with a better algorithm. In particular, rather than literally implementing the trading procedure, as analyze_market() does, see if you can find a different way to perform the same analysis.

```
In [16]: def analyze_market_faster(market, verbose=False):
    return analyze_market(market)

clearing_price0, trades0 = analyze_market(market_demo, verbose=True)
    clearing_price1, trades1 = analyze_market_faster(market_demo, verbose=True)
    print("Baseline: The clearing price is ${} with {} trades.".format(clearing_price0, trades0))
    print("Your method: The clearing price is ${} with {} trades.".format(clearing_price1, trades1))

[(3.5, 1.0), (2.0, 1.0), (2.0, 1.5)]
    Baseline: The clearing price is $1.5 with 3 trades.
    Your method: The clearing price is $1.5 with 3 trades.

In [17]: # This code cell times your method and is here for debugging purposes
%timeit analyze_market_faster(market_for_timing)

1.52 s ± 268 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
```

The following two cells are the tests for Exercise 2 that you must pass to get points.

```
In [18]: # Test cell 0: `analyze_market_faster_correctness` (1 point)
         import random
         print("Checking corner cases...")
         assert analyze_market([]) == analyze_market_faster([])
         assert analyze_market([(1, 'buyer')]) == analyze_market_faster([(1, 'buyer')])
         assert analyze_market([(1, 'seller')]) == analyze_market_faster([(1, 'seller')])
         print("Checking random cases...")
         for trial_num in range(1, 11): # 10 random trials
             num_people_for_testing = random.randrange(1, 10)
             avg_buyers_for_testing = random.randrange(num_people_for_testing+1)
             prob_buyer_for_testing = avg_buyers_for_testing / num_people_for_testing
             max_price_for_testing = random.randrange(5, 20)
             print("Trial {}: {} people, {} buyers (expected; prob={}), max price of ${}".format(trial_num,
                                                                                                   num_people_f
         or_testing,
                                                                                                   avg_buyers_f
         or_testing,
                                                                                                   prob_buyer_f
         or_testing,
                                                                                                   max_price_fo
         r_testing))
             market_for_testing = create_random_market(num_people_for_testing, prob_buyer_for_testing, max_pr
         ice_for_testing)
             naive result = analyze market(market for testing)
```

```
your_result = analyze_market_faster(market_for_testing)
             assert your result == naive result, "Results do not match on this market (`market for testing`):
         \n\t{}.\nYour result is {} while the baseline produces {}.".format(market_for_testing, your_result,
         naive_result)
         print("\n(Passed!)")
         Checking corner cases...
         Checking random cases...
         Trial 1: 2 people, 1 buyers (expected; prob=0.5), max price of $18
         Trial 2: 2 people, 1 buyers (expected; prob=0.5), max price of $19
         Trial 3: 4 people, 3 buyers (expected; prob=0.75), max price of $11
         Trial 4: 4 people, 1 buyers (expected; prob=0.25), max price of $6
         Trial 5: 5 people, 2 buyers (expected; prob=0.4), max price of $17
         Trial 6: 6 people, 5 buyers (expected; prob=0.833333333333334), max price of $19
         Trial 7: 6 people, 3 buyers (expected; prob=0.5), max price of $10
         Trial 8: 9 people, 0 buyers (expected; prob=0.0), max price of $10
         Trial 9: 7 people, 4 buyers (expected; prob=0.5714285714285714), max price of $14
         Trial 10: 2 people, 1 buyers (expected; prob=0.5), max price of $10
         (Passed!)
In [19]: # Test cell 1: `analyze_market_faster_speed` (5 points)
         market for timing2 = create random market(20000, 0.5, 100)
         print("Timing the naive method...")
         t_naive = %timeit -o analyze_market(market_for_timing2)
         print("\nTiming your method...")
         t_you = %timeit -o analyze_market_faster(market_for_timing2)
         speedup = t_naive.average / t_you.average
         print("\nYour method is {:.1f}x faster than the baseline.".format(speedup))
         assert speedup >= 50, "Sorry, not yet fast enough!"
         print("\n(Passed!)")
         Timing the naive method...
         1.32 s ± 15.5 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
         Timing your method...
         1.32 s ± 1.12 ms per loop (mean ± std. dev. of 7 runs, 1 loop each)
```