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In [1]: from typing import Tuple
import math
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```
In [2]: def normal_approximation_to_binomial(n: int, p: float) -> Tuple[float, float]:
        """Returns mu and sigma corresponding to a Binomial(n, p)"""
        mu = p * n
        sigma = math.sqrt(p * (1 - p) * n)
        return mu, sigma
```

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In [5]: from scratch.probability import normal_cdf
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In [6]: # The normal_cdf Is the Probability That the Variable Is Below a Certain Threshold
normal_probability_below = normal_cdf
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In [7]: # It's Above the Threshold If It's Not Below the Threshold
def normal_probability_above(lo: float,
                             mu: float = 0,
                             sigma: float = 1) -> float:
    """The probability that a N(mu, sigma) is greater than lo."""
    return 1 - normal_cdf(lo, mu, sigma)
```

```
In [8]: # It's Between the Threshold If It's Less Than "hi", but Not Less Than "lo"
def normal_probability_between(lo: float,
                               hi: float,
                               mu: float = 0,
                               sigma: float = 1) -> float:
    """The probability that a N(mu, sigma) is between lo and hi."""
    return normal_cdf(hi, mu, sigma) - normal_cdf(lo, mu, sigma)
```

```
In [9]: # It's Outside the Threshold If It's Not Between
def normal_probability_outside(lo: float,
                               hi: float,
                               mu: float = 0,
                               sigma: float = 1) -> float:
    """The probability that a N(mu, sigma) is not between lo and hi."""
    return 1 - normal_probability_between(lo, hi, mu, sigma)
```

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In [10]: from scratch.probability import inverse_normal_cdf
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In [11]: def normal_upper_bound(probability: float,
                                mu: float = 0,
                                sigma: float = 1) -> float:
    """Returns the z for which P(Z <= z) = probability"""
    return inverse_normal_cdf(probability, mu, sigma)
```

```
In [12]: def normal_lower_bound(probability: float,
                                mu: float = 0,
                                sigma: float = 1) -> float:
    """Returns the z for which  $P(Z \geq z) = \text{probability}$ """
    return inverse_normal_cdf(1 - probability, mu, sigma)
```

```
In [13]: def normal_two_sided_bounds(probability: float,
                                        mu: float = 0,
                                        sigma: float = 1) -> Tuple[float, float]:
    """
    Returns the symmetric (about the mean) bounds that contain the specified probability
    """
    tail_probability = (1 - probability) / 2

    # The Upper Bound Should Have a tail_probability Above It
    upper_bound = normal_lower_bound(tail_probability, mu, sigma)

    # The Lower Bound Should Have a tail_probability Below It
    lower_bound = normal_upper_bound(tail_probability, mu, sigma)

    return lower_bound, upper_bound
```

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In [14]: mu_0, sigma_0 = normal_approximation_to_binomial(1000, 0.5)
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In [15]: assert mu_0 == 500
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In [16]: assert 15.8 < sigma_0 < 15.9
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In [17]: # (469, 531)
lower_bound, upper_bound = normal_two_sided_bounds(0.95, mu_0, sigma_0)
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In [18]: assert 468.5 < lower_bound < 469.5
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In [19]: assert 530.5 < upper_bound < 531.5
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In [20]: # 95% Bounds Based On the Assumption That p Is 0.5
lo, hi = normal_two_sided_bounds(0.95, mu_0, sigma_0)
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In [21]: # The Actual mu and sigma Are Based On p = 0.55
mu_1, sigma_1 = normal_approximation_to_binomial(1000, 0.55)
```

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In [22]: # A Type 2 Error Means We Fail to Reject the Null Hypothesis, Which Will Happen When X
type_2_probability = normal_probability_between(lo, hi, mu_1, sigma_1)
power = 1 - type_2_probability # 0.887
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In [23]: assert 0.886 < power < 0.888
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In [24]: hi = normal_upper_bound(0.95, mu_0, sigma_0)
# Is 526 (< 531, Since We Need More Probability In the Upper Tail)
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In [25]: type_2_probability = normal_probability_below(hi, mu_1, sigma_1)
power = 1 - type_2_probability # 0.936
```

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In [26]: assert 526 < hi < 526.1
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In [27]: assert 0.9363 < power < 0.9364
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```
In [28]: def two_sided_p_value(x: float, mu: float = 0, sigma: float = 1) -> float:
        """
        How likely are we to see a value at least as extreme as x (in either direction) if
        """
        if x >= mu:
            # x Is Greater Than the Mean, so the Tail Is Everything Greater Than x
            return 2 * normal_probability_above(x, mu, sigma)
        else:
            # x Is Less Than the Mean, so the Tail Is Everything Less Than x
            return 2 * normal_probability_below(x, mu, sigma)
```

```
In [29]: two_sided_p_value(529.5, mu_0, sigma_0) # 0.062
```

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Out[29]: 0.06207721579598835
```

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In [30]: import random
```

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In [31]: extreme_value_count = 0
for _ in range(1000):
    num_heads = sum(1 if random.random() < 0.5 else 0 # Counts the number of heads
                    for _ in range(1000))           # in 1000 flips
    if num_heads >= 530 or num_heads <= 470:         # and count how often
        extreme_value_count += 1                    # the number is "extreme"
```

```
In [ ]: # p-Value Was 0.062 => ~62 Extreme Values Out of 1000
assert 59 < extreme_value_count < 65, f"{extreme_value_count}"
```

```
In [33]: two_sided_p_value(531.5, mu_0, sigma_0) # 0.0463
```

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Out[33]: 0.046345287837786575
```

```
In [34]: tspv = two_sided_p_value(531.5, mu_0, sigma_0)
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In [35]: assert 0.0463 < tspv < 0.0464
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```
In [36]: upper_p_value = normal_probability_above  
lower_p_value = normal_probability_below
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```
In [37]: upper_p_value(524.5, mu_0, sigma_0) # 0.061
```

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Out[37]: 0.06062885772582072
```

```
In [38]: upper_p_value(526.5, mu_0, sigma_0) # 0.047
```

```
Out[38]: 0.04686839508859242
```

```
In [39]: p_hat = 525 / 1000  
mu = p_hat  
sigma = math.sqrt(p_hat * (1 - p_hat) / 1000) # 0.0158
```

```
In [40]: normal_two_sided_bounds(0.95, mu, sigma) # [0.4940, 0.5560]
```

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Out[40]: (0.4940490278129096, 0.5559509721870904)
```

```
In [41]: p_hat = 540 / 1000  
mu = p_hat  
sigma = math.sqrt(p_hat * (1 - p_hat) / 1000) # 0.0158  
normal_two_sided_bounds(0.95, mu, sigma) # [0.5091, 0.5709]
```

```
Out[41]: (0.5091095927295919, 0.5708904072704082)
```

```
In [42]: from typing import List
```

```
In [43]: def run_experiment() -> List[bool]:  
    """Flips a fair coin 1000 times, True = heads, False = tails"""  
    return [random.random() < 0.5 for _ in range(1000)]
```

```
In [44]: def reject_fairness(experiment: List[bool]) -> bool:  
    """Using the 5% significance levels"""  
    num_heads = len([flip for flip in experiment if flip])  
    return num_heads < 469 or num_heads > 531
```

```
In [45]: random.seed(0)  
experiments = [run_experiment() for _ in range(1000)]  
num_rejections = len([experiment  
    for experiment in experiments  
    if reject_fairness(experiment)])
```

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In [46]: assert num_rejections == 46
```

```
In [47]: def estimated_parameters(N: int, n: int) -> Tuple[float, float]:  
    p = n / N  
    sigma = math.sqrt(p * (1 - p) / N)  
    return p, sigma
```

```
In [48]: def a_b_test_statistic(N_A: int, n_A: int, N_B: int, n_B: int) -> float:  
    p_A, sigma_A = estimated_parameters(N_A, n_A)  
    p_B, sigma_B = estimated_parameters(N_B, n_B)  
    return (p_B - p_A) / math.sqrt(sigma_A ** 2 + sigma_B ** 2)
```

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In [49]: z = a_b_test_statistic(1000, 200, 1000, 180)    # -1.14
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In [50]: assert -1.15 < z < -1.13
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In [51]: two_sided_p_value(z)                                # 0.254
```

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Out[51]: 0.254141976542236
```

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In [52]: assert 0.253 < two_sided_p_value(z) < 0.255
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In [53]: z = a_b_test_statistic(1000, 200, 1000, 150)    # -2.94  
two_sided_p_value(z)                                     # 0.003
```

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Out[53]: 0.003189699706216853
```

```
In [54]: def B(alpha: float, beta: float) -> float:  
    """A normalizing constant so that the total probability is 1"""  
    return math.gamma(alpha) * math.gamma(beta) / math.gamma(alpha + beta)
```

```
In [55]: def beta_pdf(x: float, alpha: float, beta: float) -> float:  
    if x <= 0 or x >= 1:                # No weight outside of [0, 1]  
        return 0  
    return x ** (alpha - 1) * (1 - x) ** (beta - 1) / B(alpha, beta)
```