

```
In [1]: def uniform_cdf(x: float) -> float:
        """Returns the probability that a uniform random variable is <= x"""
        if x < 0: return 0      # Uniform random is never less than 0
        elif x < 1: return x    # e.g.  $P(X \leq 0.4) = 0.4$ 
        else: return 1         # Uniform random is always less than 1
```

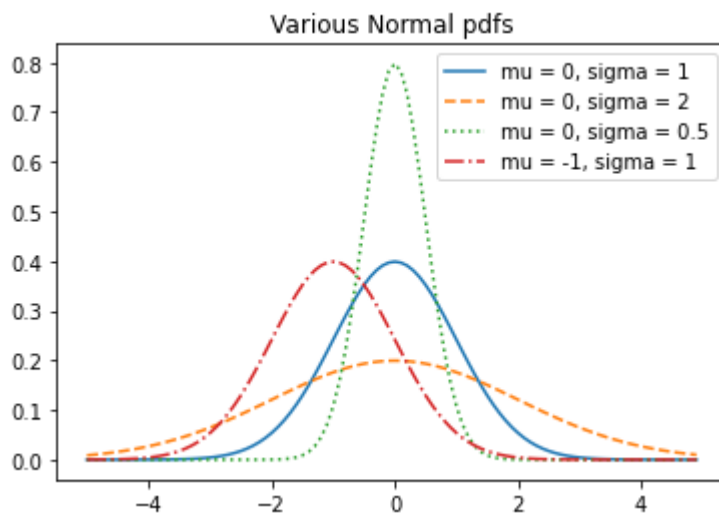
```
In [2]: import math
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In [3]: SQRT_TWO_PI = math.sqrt(2 * math.pi)
```

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In [4]: def normal_pdf(x: float, mu: float = 0, sigma: float = 1) -> float:
        return (math.exp(-(x - mu) ** 2 / 2 / sigma ** 2) / (SQRT_TWO_PI * sigma))
```

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In [5]: import matplotlib.pyplot as plt
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In [6]: xs = [x / 10.0 for x in range(-50, 50)]
plt.plot(xs, [normal_pdf(x, sigma = 1) for x in xs], '-', label = 'mu = 0, sigma = 1')
plt.plot(xs, [normal_pdf(x, sigma = 2) for x in xs], '--', label = 'mu = 0, sigma = 2')
plt.plot(xs, [normal_pdf(x, sigma = 0.5) for x in xs], ':', label = 'mu = 0, sigma = 0.5')
plt.plot(xs, [normal_pdf(x, mu = -1) for x in xs], '-.', label = 'mu = -1, sigma = 1')
plt.legend()
plt.title("Various Normal pdfs")
plt.show()
```

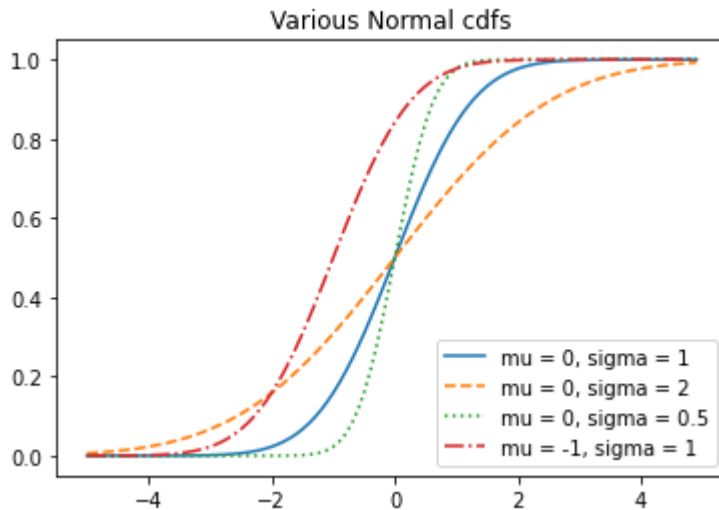


```
In [7]: # plt.savefig('various_normal_pdfs.png')
plt.gca().clear()
plt.close()
plt.clf()
```

<Figure size 432x288 with 0 Axes>

```
In [8]: def normal_cdf(x: float, mu: float = 0, sigma: float = 1) -> float:
        return (1 + math.erf((x - mu) / math.sqrt(2) / sigma)) / 2
```

```
In [9]: xs = [x / 10.0 for x in range(-50, 50)]
plt.plot(xs, [normal_cdf(x, sigma = 1) for x in xs], '-', label = 'mu = 0, sigma = 1')
plt.plot(xs, [normal_cdf(x, sigma = 2) for x in xs], '--', label = 'mu = 0, sigma = 2')
plt.plot(xs, [normal_cdf(x, sigma = 0.5) for x in xs], ':', label = 'mu = 0, sigma = 0.5')
plt.plot(xs, [normal_cdf(x, mu = -1) for x in xs], '-.', label = 'mu = -1, sigma = 1')
plt.legend(loc = 4) # Bottom right
plt.title("Various Normal cdfs")
plt.show()
```



```
In [10]: plt.close()
plt.gca().clear()
plt.clf()
```

<Figure size 432x288 with 0 Axes>

```
In [11]: def inverse_normal_cdf(p: float,
    mu: float = 0,
    sigma: float = 1,
    tolerance: float = 0.00001) -> float:
    """Find the approximate inverse using binary search"""

    # If Not Standard, Then Compute Standard and Rescale
    if mu != 0 or sigma != 1:
        return mu + sigma * inverse_normal_cdf(p, tolerance = tolerance)

    low_z = -10.0 # normal_cdf(-10) is (very close to) 0
    hi_z = 10.0 # normal_cdf(10) is (very close to) 1
    while hi_z - low_z > tolerance:
        mid_z = (low_z + hi_z) / 2 # Consider the midpoint
        mid_p = normal_cdf(mid_z) # and the cdf's value there
        if mid_p < p:
            low_z = mid_z # If the midpoint is too low, then search above
        else:
            hi_z = mid_z # If the midpoint is too high, then search below

    return mid_z
```

```
In [12]: import random
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```
In [13]: def bernoulli_trial(p: float) -> int:
        """Returns 1 with probability p and 0 with probability 1 - p"""
        return 1 if random.random() < p else 0
```

```
In [14]: def binomial(n: int, p: float) -> int:
        """Returns the sum of n bernoulli(p) trials"""
        return sum(bernoulli_trial(p) for _ in range(n))
```

```
In [15]: from collections import Counter
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```
In [16]: def binomial_histogram(p: float, n: int, num_points: int) -> None:
        """Picks points from a Binomial(n, p) and plots their histogram"""
        data = [binomial(n, p) for _ in range(num_points)]

        # Use a Bar Chart to Show the Actual Binomial Samples
        histogram = Counter(data)
        plt.bar([x - 0.4 for x in histogram.keys()],
                [v / num_points for v in histogram.values()],
                0.8,
                color = '0.75')

        mu = p * n
        sigma = math.sqrt(n * p * (1 - p))

        # Use a Line Chart to Show the Normal Approximation
        xs = range(min(data), max(data) + 1)
        ys = [normal_cdf(i + 0.5, mu, sigma) - normal_cdf(i - 0.5, mu, sigma)
              for i in xs]
        plt.plot(xs, ys)
        plt.title("Binomial Distribution vs. Normal Approximation")
        # plt.show()
```

```
In [17]: def main():
        import enum, random

        # An Enum Is a Typed Set of Enumerated Values
        # We Can Use Them to Make Our Code More Descriptive and Readable
        class Kid(enum.Enum):
            BOY = 0
            GIRL = 1

        def random_kid() -> Kid:
            return random.choice([Kid.BOY, Kid.GIRL])

        both_girls = 0
        older_girl = 0
        either_girl = 0

        random.seed(0)

        for _ in range(10000):
            younger = random_kid()
```

```

older = random_kid()
if older == Kid.GIRL:
    older_girl += 1
if older == Kid.GIRL and younger == Kid.GIRL:
    both_girls += 1
if older == Kid.GIRL or younger == Kid.GIRL:
    either_girl += 1

print("P(both | older):", both_girls / older_girl)    # 0.514 ~ 1/2
print("P(both | either): ", both_girls / either_girl) # 0.342 ~ 1/3

assert 0.48 < both_girls / older_girl < 0.52
assert 0.30 < both_girls / either_girl < 0.35

def uniform_pdf(x: float) -> float:
    return 1 if 0 <= x < 1 else 0

```

In [18]:

```

if __name__ == "__main__": main()

```

```

P(both | older): 0.5007089325501317
P(both | either): 0.3311897106109325

```