## Math 341: Probability Eighth Lecture (10/6/09)

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## Independence

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## Independence of events

Two events $A$ and $B$ are independent if $\mathbb{P}(A \cap B)=\mathbb{P}(A) \mathbb{P}(B)$.

As $\mathbb{P}(A \cap B)=\mathbb{P}(A \mid B) \mathbb{P}(B)$, if $\mathbb{P}(B)>0$ this is equivalent to $\mathbb{P}(A \mid B)=\mathbb{P}(A)$, or that knowledge of one happening does not affect knowledge of the other happening.

## Independence (continued)

## Independence of random variables

Two random variables $X$ and $Y$ are independent if for all $x, y$ :

- Discrete case: events $\{X=x\}$ and $\{Y=y\}$ are independent.
- Continuous case: events $\{X \leq x\}$ and $\{Y \leq y\}$ are independent.

Non-trivial example (from book): Toss a coin with probability $p$ of heads $N$ times, where $N$ is a Poisson random variable with parameter $\lambda$. Then the number of heads and the number of tails are independent random variables.

## Main result

## Key Lemma

Let $g, h: \mathbb{R} \rightarrow \mathbb{R}$ and assume $X$ and $Y$ are independent random variables. Then $g(X)$ and $h(Y)$ are independent.

- The proof involves real analysis, specifically properties of the $\sigma$-fields.
- Assume $g$, $h$ continuous and strictly increasing (so $g^{-1}, h^{-1}$ exist) and $X, Y$ continuous random variables.
- Then $\{g(X) \leq a\}$ and $\{h(Y) \leq b\}$ are the same as $\left\{X \leq g^{-1}(a)\right\}$ and $\left\{Y \leq h^{-1}(b)\right\}$.
- As latter two sets are independent (due to independence of $X, Y$ ), we see $g(X)$ and $h(Y)$ independent.


## Sections 3.3 \& 4.3:

Expectation

## Definition

## Expectation (mean value, average)

$X$ random variable with density / mass function $f_{X}$, then expected value is

- Discrete case: $\mathbb{E}[X]:=\sum_{x} x f_{X}(x)$ if sum converges absolutely.
- Continuous case: $\mathbb{E}[X]:=\int_{-\infty}^{\infty} x f_{X}(x) d x$ if integral converges absolutely.

Notation:

- Often use integral notation for both.
- Set $\mathbb{E}[g(X)]$ equal to $\int_{-\infty}^{\infty} g(x) f_{X}(x) d x$ if exists.


## Definition (continued)

## Moments

Let $X$ be a random variable. We define

- $k^{\text {th }}$ moment: $m_{k}:=\mathbb{E}\left[X^{k}\right]$ (if converges absolutely).

Assume $X$ has a finite mean, which we denote by $\mu$ (so $\mu=\mathbb{E}[X]$ ). We define

- $k^{\text {th }}$ centered moment: $\sigma_{k}:=\mathbb{E}\left[(X-\mu)^{k}\right]$ (if converges absolutely).


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- $k^{\text {th }}$ centered moment: $\sigma_{k}:=\mathbb{E}\left[(X-\mu)^{k}\right]$ (if converges absolutely).
- Be alert: Some books write $\mu_{k}^{\prime}$ for $m_{k}$ and $\mu_{k}$ for $\sigma_{k}$.
- Call $\sigma_{2}$ the variance, write it as $\sigma^{2}$.
- Note $\sigma^{2}=\mathbb{E}\left[(X-\mu)^{2}\right]=\mathbb{E}\left[X^{2}\right]-\mathbb{E}[X]^{2}$.


## Clicker Questions

## Prime divisors

## Number of prime divisors

Let $N$ be a large number. If we choose an integer of size approximately $N$, on average about how many distinct prime factors do we expect $N$ to have (as $N \rightarrow \infty$ )? It might be useful to recall the Prime Number Theorem: The number of primes at most $x$ is about $x / \log x$.

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- (a) At most 10.
- (b) Around $\log \log \log N$.
- (c) Around $\log \log N$.
- (d) Around $\log N$.
- (e) Around $\log N \log \log N$.
- (f) Around $(\log N)^{2}$.
- (g) This is an open question.


## Fermat Primes

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If $F_{n}=2^{2^{n}}+1$ is prime, we say $F_{n}$ is a Fermat prime. About how many Fermat primes are there less than $x$ as $x \rightarrow \infty$ ?

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About how many Fermat primes are there less than $x$ as $x \rightarrow \infty$ ?

- (a) 5
- (b) 10
- (c) Between 11 and 20.
- (d) Between 21 and 100.
-(e) $\log \log \log x$.
-(f) $\log \log x$.
- (g) $\log x$.
- (h) More than $\log x$.
- (i) This is an open problem.


## $3 x+1$ Problem

## $3 x+1$ : Iterating to the fixed point

Define the $3 x+1$ map by $a_{n+1}=\frac{3 a_{n}+1}{2^{k}}$ where $2^{k}| | 3 a_{n}+1$. Choose a large integer $N$ and randomly choose a starting seed $a_{0}$ around $N$. About how many iterations are needed until we reach 1 (equivalently, about how large is the smallest $n$ such that $a_{n}=1$ ) ?

## $3 x+1$ Problem

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There is a constant $C$ so that the answer is about

- (a) At most 10.
- (b) Around $C \log \log \log N$.
- (c) Around $C \log \log N$.
- (d) Around $C \log N$.
- (e) Around $C \log N \log \log N$.
- (f) Around $C(\log N)^{2}$.
- (a) This is an open question.

