Lecture 6: The Hafner-McCurley Class Group Algorithm
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The goal of this lecture is to prove the following theorem due to Hafner and McCurley [3].
Theorem. Under the Extended Riemann Hypothesis, there is a Las Vegas algorithm for computing the ideal class group of an imaginary quadratic order.

In the following, we identify ideal classes with reduced quadratic forms of discriminant $-d$ for $d>0$ such that $-d$ is a quadratic discriminant. Given two forms $f_{1}, f_{2}$, the operation $f_{1} f_{2}$ is the composition and reduction of the result, thus giving the ideal class represented by the product of the ideal classes represented by $f_{1}, f_{2}$.

### 6.1 Overview

Let us use $f_{i}$ to denote the equivalence class $\left.\left[\left(p_{i}, b_{i},.\right)\right)\right]$. These equivalence classes are called as prime forms. We define the subexponential function $L$ by

$$
L(d):=e^{\sqrt{\log d \log \log d}}
$$

Let $n=L(d)^{z+o(1)}$ for a fixed positive number $z$. Then the classes $\left[\left(p_{i}, b_{i},.\right)\right], 1 \leq i \leq n$ generate the class group $\mathrm{Cl}(-d)$ from Bach's bounds. we can define a homomorphism $\varphi: \mathbb{Z}^{n} \rightarrow \mathrm{Cl}(-d)$ by

$$
\varphi\left(x_{1}, \ldots, x_{n}\right)=\prod_{i=1}^{n} f_{i}^{x_{i}}
$$

An integer relation on $f_{1}, \ldots, f_{n}$ is the vector $\left(x_{1}, \ldots, x_{n}\right) \in \mathbb{Z}^{n}$ such that $\varphi\left(x_{1}, \ldots, x_{n}\right)=\prod_{i=1}^{n} f_{i}^{x_{i}}=1_{\mathrm{Cl}(-d)}$ where $1_{\mathrm{Cl}(-d)}$ is the identity element of the class group $C(-d)$. Relations in $f_{1}, \ldots, f_{n}$ form an additive subgroup of $\mathbb{Z}^{n}$ (i.e. a Euclidean lattice) which we denote as $\Lambda$. Since $\varphi$ is surjective, we have

$$
\mathbb{Z}^{n} / \Lambda \cong \mathrm{Cl}(-d)
$$

Therefore, the computation of $\mathrm{Cl}(-d)$ reduces to the search for relations between the $f_{1}, \ldots, f_{n}$. Once enough relations are collected to generate $\Lambda$, a polynomial time linear algebra phase yields the quotient $\mathbb{Z}^{n} / \Lambda$, and therefore the ideal class group $\mathrm{Cl}(-d)$.

The subexponential algorithm of Hafner and McCurley consists in the choice of a factor basis $f_{1}, \ldots, f_{n}$, and the resolution of the following two main tasks

- Finding a generating set of elements of $\Lambda$, the lattice of relations between factor basis elements.
- Computing the quotient $\mathbb{Z}^{n} / \Lambda$.

The quotient computation is well understood, and essentially corresponds to the computation of the Smith Normal Form (SNF) of the matrix representing a basis for $\Lambda$.

Making a formal case for the run time without the heuristic that elements sampled in $\Lambda$ behave randomly demands a little bit of care. We need 3 different phases.

- Phase 1: For each $k=1, \ldots, n$, we compute a relation whose $k$-th coefficient is significantly larger than the others. This ensures the fact that at the end of the collection of the first $n$ relations, the lattice $\Lambda_{0}$ they generate has full rank.
- Phase 2: We construct additional relations in order to ensure that at the end of this phase, the lattice $\Lambda_{1}$ they generate satisfies $\operatorname{det}\left(\Lambda_{1}\right) \in 2^{O\left(\log ^{4} d\right)}$.
- Phase 3: Once we have $\operatorname{det}\left(\Lambda_{1}\right) \in 2^{O\left(\log ^{4} d\right)}$, we use an expensive randomization process to find the few extra relations needed to generate $\Lambda$.


### 6.2 Phase 1

In this section, we show how to create $n$ linearly independent relations between the $\left(f_{i}\right)_{i \leq n}$. We ensure that the matrix $\left(a_{i, j}\right)$ whose rows are the relation vectors satisfies $\left|a_{i i}\right|>\sum_{j \neq i}\left|a_{i, j}\right|$, which in turns guarantees that the matrix $\left(a_{i, j}\right)$ has full rank. To reduce the run time of the relation search, we use the fact that the Cayley graph of $\mathrm{Cl}(-d)$ is an expander graph. Let $n_{0}$ be such that

$$
f_{1}, \ldots, f_{n_{0}}=\left\{\text { Prime forms corresponding to } p \leq \log ^{2+\varepsilon}(d) \text { and }\left(\frac{d}{p}\right) \neq 1\right\}
$$

Choosing $t=\log (d) \gg C \frac{\log |\mathrm{Cl}(-d)|}{\log \log d}$, we draw random vectors $\vec{x}$ of $\ell_{1}$-norm $t$ until $f \cdot\left(\prod_{i \leq n_{0}} f_{i}^{x_{i}}\right)$ factors as a product of elements of $\mathcal{B}$ (i.e. is $\mathcal{B}$-smooth).

Proposition 6.1. Under the ERH, there is a Las Vegas algorithm that takes as input a reduced quadratic form $f$, and returns $\vec{x} \in \mathbb{Z}^{n}$ of $\ell_{1}$-norm bounded by $2 \log d$ such that $f=\prod_{i \leq n} f_{i}^{x_{i}}$ in time

$$
L(d)^{1 / 4 z+o(1)}+L(d)^{z+o(1)}
$$

Its probability of success is at least $1-\frac{1}{d^{1+o(1)}}$.

Proof. Each attempt at drawing $\vec{y} \in \mathbb{Z}^{n_{0}}$ of $\ell_{1}$-norm $\log d$ such that $\prod_{i \leq n_{0}} \cdot f$ is $\mathcal{B}$ can be viewed as a random walk in the Cayley graph of $\mathrm{Cl}(-d)$ of length $\log d$. It has a probability at least $\frac{|S|}{2|\mathrm{Cl}(-d)|}$ of landing in a subset $S \subseteq \mathrm{Cl}(-d)$. We choose $S$ to be the classes corresponding to the smooth reduced quadratic forms. For this, it was shown by Seyssen [5] that the probability is at least $1 / L(d)^{1 / 4 z+o(1)}$. This means that we can repeat this experiment $L(d)^{1 / 4 z+o(1)}$ times to have a probability $1-\frac{1}{d^{1+o(1)}}$ of success.

Now rather than testing the smoothness of each of the $L(d)^{1 / 4 z+o(1)}$ reduced forms we collect, we run Bernstein's batch smoothness test [1] only once on the whole set, which has a run time of $L(d)^{1 / 4 z+o(1)}+$ $L(d)^{z+o(1)}$.

Finally, once a suitable $\vec{y}$ such that $f \cdot \prod_{i} f_{i}^{y_{i}}=f^{\prime}$ for a reduced smooth form $f^{\prime}$ is found, we decompose $f^{\prime}=\prod_{i} f_{i}^{z_{i}}$, and we obtain the relation $\prod_{i} f_{i}^{z_{i}-y_{i}}=1_{\mathrm{Cl}(-d)}$. Since the norm of $f^{\prime}$ is less than $\sqrt{d}$, we have that the $\ell_{1}$-norm of $\vec{z}$ is less than $\log d$ and thus the $\ell_{1}$ norm of $\vec{x}:=\vec{z}-\vec{y}$ is less than $2 \log d$.

Given the above building block (which will be reused in subsequent phases, and even for applications in future lectures such as DLP, ideal decomposition etc ...), we can easily compute a full rank matrix of relations by choosing $f=f_{i}^{2 n d}$ for each of the $n$ elements $f_{i} \in \mathcal{B}$. This ensures that the $i$-th row $\left(a_{i, j}\right)_{j \leq n}$ has a dominant $i$-th coefficient as requested.

Proposition 6.2. Under the $E R H$, there is an algorithm that outputs $n$ linearly independent relations between elements of $\mathcal{B}$ with probability at least $1-\frac{1}{d^{1+o(1)}}$ in time

$$
L(d)^{z+o(1)}\left(L(d)^{z+o(1)}+L(d)^{1 / 4 z+o(1)}\right)
$$

### 6.3 Phase 2

At the end of Phase 1, we have a sublattice $\Lambda_{0} \subseteq \Lambda$ of full rank with $\operatorname{det}\left(\Lambda_{0}\right)<n^{5 n / 2} d^{n}$ by Hadamard bound. Then we add new relations hoping that they do not belong to the previous sublattice of relations. Starting with $\Lambda_{1}=\Lambda_{0}$, each time we find $\vec{x} \notin \Lambda_{1}$, and update $\Lambda_{1}$ by doing

$$
\Lambda_{1} \leftarrow \Lambda_{1}+\mathbb{Z} \vec{x}
$$

the determinant of $\Lambda_{1}$ gets divided by at least a factor 2 .
To create relations, we first draw a vector $\vec{y}$ uniformly at random in $W_{n}\left(d^{2}\right)$ for

$$
W_{n}(t):=\left\{x: x \in \mathbb{Z}^{n},\|x\|_{\infty} \leq t\right\}
$$

and we compute the reduced form $f=\prod_{i} f_{i}^{y_{i}}$. Then we use Proposition 6.1 to create $\vec{x} \in \mathbb{Z}^{n}$ such that $f=\prod_{i} f_{i}^{x_{i}}$. Then we have that $\vec{y}-\vec{x} \in \Lambda$ is a relation. The question is "does it belong to $\Lambda_{1}$ ?". For $\vec{y}-\vec{x}$ to be outside of $\Lambda_{1}$, it suffices that the random vector $\vec{y}$ drawn from $W_{n}\left(d^{2}\right)$ be sufficiently far from $\Lambda_{1}$. Indeed, we know that the $\ell_{1}$-norm of $\vec{x}$ is less than $2 \log d$, therefore it suffices to draw $\vec{y}$ at distance $2 \log d+\varepsilon$ from $\Lambda_{1}$ to guarantee that the resulting relation is not in $\Lambda_{1}$. In other words, we need to draw $\vec{y}$ in $W_{n}\left(d^{2}\right) \backslash V$ for

$$
V:=\left\{\bigcup B(x,(2+\varepsilon) \log d): x \in \Lambda_{1}\right\}
$$

where $B(x, r)$ denotes the $n$-dimensional sphere of radius $r$ for the Euclidean distance, centered at $x$.
To evaluate the odds of drawing $\vec{y}$ outside of $V$, we compute an upper bound on the number of integer points in $V$. We need to answer two questions:

- How many lattice points of $\Lambda_{1}$ are there in $W_{n}\left(d^{2}\right)$ ?
- What is an upper bound on the number of integer vectors in each $B(x,(2+\varepsilon) \log d)$ ?

The first question is answered by [3, Lem. 1] which implies that

$$
\left|\Lambda_{1} \cap W_{n}\left(d^{2}\right)\right| \in \frac{\left(2 d^{2}\right)^{n}}{\operatorname{det}\left(\Lambda_{1}\right)} \cdot\left(1+O\left(\frac{n^{3}}{d}\right)\right)
$$

Then, according to [4, Corollary 1.4], the number of integer elements contained inside each individual $n$ dimensional sphere is bounded from above by $\frac{3 \cdot e^{\pi \cdot k^{2} \cdot r^{2}}}{2}$, where $r$ is the radius of the sphere and $k=10(\log n+$ 2). Choosing $r=(2+\varepsilon) \log d$ yields

$$
\left|W_{n}\left(d^{2}\right) \backslash V\right| \geq\left(2 d^{2}\right)^{n}-\frac{\left(2 d^{2}\right)^{n}}{e^{\log ^{4} d(1+o(1))}}(1+o(1))
$$

whenever $\operatorname{det}\left(\Lambda_{1}\right) \geq e^{\log ^{4} d(1+o(1))}$. In these conditions, the probability of drawing $\vec{y}$ in $W_{n}\left(d^{2}\right) \backslash V$ is at least $1+\frac{1}{e^{\log ^{4} d(1+o(1))}}$.
We repeat this process $\log _{2}\left(n^{5 n / 2} d^{n}\right)=n^{1+o(1)}$ times to ensure that even with the pessimistic estimates on $\operatorname{det}\left(\Lambda_{0}\right)$ and even if we decrease the determinant by a factor 2 at a time, we end up with $\operatorname{det}\left(\Lambda_{1}\right)<e^{\log ^{4} d}$ at the end of the phase. We compute the determinant of $\Lambda_{1}$ in time $n^{3+o(1)}$ by reducing it with [6, Th. 58] to the case of a square matrix and then by using the Smith Normal Form algorithm for square nonsingular matrices of [2].

Proposition 6.3. Under the ERH, Phase 2 produces a sublattice $\Lambda_{1}$ such that $\operatorname{det}\left(\Lambda_{1}\right)<e^{\log ^{4} d}$ with probability at least $1-\frac{1}{d^{1+o(1)}}$ in time

$$
L(d)^{3 z+o(1)}+L(d)^{z+1 / 4 z+o(1)}
$$

Proof. The cost of each relation is that of the computation of $f$, which is in $L(d)^{z+o(1)}$, and of the execution of the relation search of Proposition 6.1 which is in $L(d)^{1 / 4 z+o(1)}+L(d)^{z+o(1)}$. This is repeated $L(d)^{z+o(1)}$ times for a total cost of $L(d)^{z+1 / 4 z+o(1)}+L(d)^{2 z+o(1)}$. Then the final determinant computation runs in time $L(d)^{3 z+o(1)}$ thus proving our statement on the run time.
The probability of success is at least that of succeeding $n^{1+o(1)}$ times at creating a relation outside of $\Lambda$ that satisfies $\operatorname{det}\left(\Lambda_{1}\right) \geq e^{\log ^{4} d}$. This means it is at least

$$
\left(1-\frac{1}{e^{\log ^{4} d(1+o(1))}}\right)^{n^{1+o(1)}} \cdot\left(1-\frac{1}{d^{1+o(1)}}\right)^{n^{1+o(1)}}=1-\frac{1}{d^{1+o(1)}}
$$

### 6.4 Phase 3

At the beginning of Phase 3 , we have $\Lambda_{1} \subseteq \Lambda$ with $|\Lambda / \Lambda|<e^{\log ^{4} d}$. Then we create a tower of sublattices $\left(\Lambda_{i}\right)_{2 \leq i \leq m}$ of the lattice of relations such that

$$
\Lambda_{0} \subseteq \Lambda_{1} \subseteq \ldots, \subseteq \Lambda_{m}=\Lambda
$$

The key observation proved in [3, Lem. 2] is that when relations are obtained simply by testing the $\mathcal{B}$ smoothness of elements $f=\prod_{i} f_{i}^{x_{i}}$ for a vector $\vec{x}$ drawn uniformly at random in $W_{d}\left(d^{2}\right)$, the probability that they belong to a given coset in $\Lambda / \Lambda_{1}$ is essentially given by $\operatorname{det}(\Lambda) / \operatorname{det}\left(\Lambda_{1}\right)$. This means that new relations have somewhat comparable chances of landing in different cosets of $\Lambda / \Lambda_{1}$. Once every cost has been hit at least once, the relation collection is complete. We can show that only $\log ^{4} d(1+o(1))$ such steps are required to generate the whole lattice $\Lambda$ with good enough probability. This point is important because unlike in Phase 2, the cost of finding each individual relation is $L(d)^{z+1 / 4 z+o(1)}$ due to the fact that we recompute a new $f$ for each form tested for smoothness. At the end, we use the Smith Normal Form algorithm described in the previous section that relies on $[2,6]$ to produce $d_{1}, \ldots, d_{n}$ such that

$$
\mathrm{Cl}(-d)=\mathbb{Z} / d_{1} \mathbb{Z} \oplus \mathbb{Z} / d_{2} \mathbb{Z} \oplus \ldots \oplus \mathbb{Z} / d_{n} \mathbb{Z}
$$

Theorem 6.4. Under the $E R H$, there is a Las Vegas algorithm to compute $\mathrm{Cl}(-d)$ in time $L(d)^{3 / \sqrt{8}+o(1)}$ with probability at least $1-\frac{1}{d^{1+o(1)}}$.

Proof. As mentioned above, each relation obtained by drawing $\vec{x} \in W_{n}\left(d^{2}\right)$ and test it for smoothness (using [1]) takes time $L(d)^{z+1 / 4 z+o(1)}$. We can also show using [3, Lem. 2] that the probability that the
resulting relation is in any given coset of $\Lambda / \Lambda_{1}$ is in $\frac{\operatorname{det}(\Lambda)}{\operatorname{det}\left(\Lambda_{1}\right)}(1+o(1))$. Then, as observed in [3, Lem. 2] we generate all of $\Lambda$ with probability at least $1-\frac{1}{d}$ after collecting $m$ relations where $m$ satisfies

$$
m \geq \frac{\log \left|\Lambda / \Lambda_{1}\right|+\log d}{\log (2 / \alpha)}
$$

for some $\alpha=1+O\left(\frac{n^{3}}{d}\right)$. This means that a polynomial number of relations is required, which proves that the run time is again

$$
L(d)^{3 z+o(1)}+L(d)^{z+1 / 4 z+o(1)}
$$

This value is minimized for $z=1 / \sqrt{8}$, which yields a total run time of $L(d)^{3 / \sqrt{8}+o(1)}$.

## References

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