

Previously, the asymptotically fastest algorithms were due to Berlekamp (1970), Cantor and Zassenhaus (1981), and Ben-Or (1981). These algorithms are also probabilistic. Berlekamp's algorithm can be implemented with $O(n^\omega +$

$n \log q$) operations in F_q , if we can multiply two $n \times n$ matrices with $O(n^\omega)$ operations; we can choose $\omega < 2.376$ (Coppersmith & Winograd 1990). The Cantor-Zassenhaus algorithm takes $O(n^2 \log q)$ operations in F_q , and Ben-Or's algorithm also has this running time.

The contrast between the running time of our algorithm and these other algorithms is most striking when $\log q \approx n$. In this case, our algorithm takes $O(n^2)$ operations in F_q , Berlekamp's takes $O(n^\omega)$, and the Cantor-Zassenhaus algorithm takes $O(n^3)$.

Our algorithm is a variant of the Cantor-Zassenhaus algorithm, and, like that algorithm, breaks the general factoring problem into three sub-problems:

1. *squarefree factorization*, i.e., making an arbitrary polynomial squarefree;
2. *distinct-degree factorization*, i.e., splitting a squarefree polynomial into polynomials whose irreducible factors have the same degree;
3. *equal-degree factorization*, i.e., completely factoring a squarefree polynomial whose irreducible factors have the same degree.

The algorithm of Yun (1976) computes the squarefree factorization with $O(n \log q)$ operations in F_q .

Let $f \in F_q[x]$ be the polynomial to be factored, $n = \deg f$, and $R = F_q[x]/(f)$. We shall always use the symbol ξ to denote $(x \bmod f) \in R$. In solving the distinct-degree and equal-degree factorization problems, the *Frobenius map* on R , which sends $\alpha \in R$ to α^q , plays a fundamental rôle. Several algorithms for these problems compute some of the iterates $\alpha, \alpha^q, \dots, \alpha^{q^n}$ of the Frobenius map for various α in R .

At the heart of our algorithms is the following idea, which we call the *polynomial representation* of the Frobenius map. Suppose we have precomputed the special element $\beta = \xi^q \in R$. An arbitrary $\alpha \in R$ can be represented as $\alpha = (g \bmod f) \in R$, where $g \in F_q[x]$ has degree less than n . Thus $\alpha = g(\xi)$, and the problem of computing α^q can be viewed as the problem of evaluating the polynomial g at the point β , since

$$\alpha^q = g(\xi)^q = g(\xi^q) = g(\beta).$$

This polynomial representation of the Frobenius map is due to Erich Kaltofen who used it to design a factoring algorithm that takes $O(n^3 + n \log q)$ operations in F_q and only linear space. Previous algorithms with this running time used quadratic space.

Combining the polynomial representation of the Frobenius map with an algorithm for fast multi-point polynomial evaluation over the ring R allows us to compute $\alpha, \alpha^q, \dots, \alpha^{q^n}$ for a given $\alpha \in R$ with $O(n^2 + n \log q)$ operations in F_q , whereas the obvious repeated squaring algorithm uses $O(n^2 \log q)$ operations in F_q , and an algorithm based on a matrix representation of the Frobenius map uses $O(n^\omega + n \log q)$ operations in F_q . This new approach leads to our basic algorithms for solving the distinct-degree and equal-degree factorization problems, which use $O(n^2 + n \log q)$ operations in F_q . These are described in Section 3; a part of the probabilistic analysis is done in Section 4.

A high-level view of the three approaches to computing the Frobenius map is that repeated squaring treats it as a multiplicative map on R , the matrix representation as an F_q -linear function, and the polynomial representation as an F_q -algebra endomorphism.

The running time of our algorithm is fairly insensitive to the largest degree d of irreducible factors of f . However, the algorithm of Cantor and Zassenhaus actually uses only $O(nd \log q)$ operations in F_q , and hence may be faster than ours when d is small. In particular, our method does not improve root-finding.

Section 5 takes another look at equal-degree factorization. For this problem, we do not need to compute the individual iterates of the Frobenius map, but only sums of the form $\sum_{0 \leq i < d} \alpha^{q^i}$, where d is the degree of the irreducible factors of f . We call such a sum a *trace map*, as it is given by the formula for the trace from F_{q^d} to F_q . The polynomial representation of the Frobenius map allows us to compute the trace map for a given $\alpha \in R$ using only $O(n^{(\omega+1)/2} + n \log q)$ operations in F_q . This leads to an algorithm for equal-degree factorization with the same time bound, which is $O(n^{1.7} + n \log q)$.

At the present time, we have the anomaly that our algorithm for the "easy" problem of distinct-degree factorization, which uses $O(n^2 + n \log q)$ operations in F_q , is asymptotically slower than our algorithm for the "harder" problem of equal-degree factorization, which uses $O(n^{1.7} + n \log q)$ operations in F_q . The latter problem seems harder since no deterministic polynomial time algorithm for it is known if the characteristic of the field is large. In Section 6, we discuss a special version of the multi-point polynomial evaluation problem that is the bottleneck of our current algorithms for distinct-degree factorization: asymptotic improvements on it would immediately improve the time for distinct-degree factorization, and hence the time for the general factoring problem. No lower bounds are known for the factoring problem, except that factoring quadratic polynomials takes $\Omega(\log q)$ operations in F_q (von zur Gathen & Seroussi 1991).

Throughout the paper, we also discuss the space requirements—in terms of storage of elements of F_q —of our algorithms. Our basic algorithms in Section 3

use space $O(n^2)$. The equal-degree factoring algorithm in Section 5 uses space $O(n \log n)$, and we present in Section 6 a distinct-degree factorization algorithm with space $O(n^{3/2})$.

In Section 7, we show how to test polynomials for irreducibility with $O(n^{(\omega+1)/2} + n \log q)$ operations in F_q . This improves upon the previous methods of Butler (1954), which takes $O(n^\omega + n \log q)$ operations in F_q , and Rabin (1980), using $O(n^2 \log q)$ operations in F_q . We obtain a similar improvement for finding normal bases, and—with a different method—an even faster algorithm for computing traces in extension fields.

In Section 8, we briefly discuss the usefulness of our methods under the restriction that so-called “classical” algorithms for polynomial and matrix arithmetic are used.

Our algorithms for distinct-degree factorization are deterministic, and those for equal-degree factorization are probabilistic. Indeed, it is a well-known open question to find a deterministic polynomial-time algorithm for the latter problem. In Section 9, we present a new deterministic algorithm for equal-degree factorization. Our results are best appreciated in the important special case where $q = p^k$ and p is a small, fixed prime (e.g., $p = 2$). Then our deterministic algorithm factors an arbitrary polynomial in $F_q[x]$ of degree n using $O(n^2 + n^{3/2}k)$ operations in F_q , and $O(n^2k + n^{3/2}k^2)$ bit operations. A variant extracts a single irreducible factor using just $O(n^2 + nk)$ operations in F_q , and $O(n^2k + nk^2)$ bit operations.

Among the previously known deterministic factorization algorithms, a deterministic variant of Berlekamp’s algorithm described by von zur Gathen (1987) has bit complexity $O((nk)^\omega)$, and an algorithm by Thiong ly (1989) combines the trace computations of Camion (1983) and Berlekamp (1970); its bit complexity is not stated explicitly, and appears to be $O(n^3k^2 + n^2k^3)$. An algorithm described by Shoup (1990) works only for $k = 1$ and has a bit complexity of $O(n^2)$.

2. Polynomial arithmetic

In this section, we briefly recall some facts about polynomial arithmetic, and introduce some notation.

We will let $M(n)$ denote an upper bound on the cost of polynomial multiplication. That is, we assume that we have an algorithm that multiplies two polynomials of degree n over an arbitrary ring R (commutative, containing 1) that uses $O(M(n))$ operations in R (additions, subtractions, and multiplications). We assume that this algorithm uses space for $O(n)$ elements of R .

In the analysis of some recursive algorithms, we implicitly assume that M is suitably well-behaved, namely, that M is actually a function $M: \mathbb{R}_{>0} \rightarrow \mathbb{R}_{>0}$ growing at least linearly and at most quadratically. More precisely, we insist that

$$\forall t \in \mathbb{R}_{>0} \forall a \in \mathbb{R}_{\geq 1} \quad aM(t) \leq M(at) \leq a^2M(t).$$

With the asymptotically fastest algorithms for polynomial multiplication (Schönhage & Strassen 1971, Schönhage 1977, Cantor & Kaltofen 1991), we can use $M(n) = O(n \log n \log \log n)^1$.

Let $f \in F_q[x]$ have degree n , and $R = F_q[x]/(f)$. For $g \in F_q[x]$, we denote by $(g \bmod f)$ the image of g in R , and reserve the notation $(g \text{ rem } f)$ to denote the polynomial in $F_q[x]$ obtained as the remainder on dividing g by f . For $\alpha \in R$, the *canonical representative* of α is the unique polynomial $g \in F_q[x]$ of degree less than n such that $(g \bmod f) = \alpha$. At times we use the following explicit notation: for $\alpha \in R$, $\tilde{\alpha} \in F_q[x]$ is the canonical representative of α .

To implement arithmetic in R , we use the canonical representatives, and so one addition or subtraction of elements in R takes $O(n)$ operations in F_q . Since division with remainder of polynomials of degree at most n takes $O(M(n))$ operations (see Aho *et al.* 1974, § 8.3), multiplication of elements in R takes $O(M(n))$ operations in F_q .

The greatest common divisor of two polynomials of degree n over F_q can be computed with $O(M(n) \log n)$ operations in F_q and space for $O(n)$ elements of F_q (see Aho *et al.* 1974, § 8.9, Strassen 1983). We always take the gcd of (nonzero) polynomials to be monic.

We also need an algorithm for evaluating a polynomial in $R[y]$ of degree at most n at n points in R . The “standard” fast algorithm, as in Aho *et al.* (1974, § 8.5), takes $O(M(n) \log n)$ operations in R and space for $O(n \log n)$ elements of R , which translates into $O(M(n)^2 \log n)$ operations in F_q and space for $O(n^2 \log n)$ elements of F_q . In the remainder of this section, we show how to solve this problem with only $O(M(n^2) \log n)$ operations in F_q and space for $O(n^2)$ elements of F_q ; the reader may wish to skip this on first reading.

LEMMA 2.1. *Given a polynomial $g \in R[x]$ of degree n over an arbitrary ring R , and points $\alpha_1, \dots, \alpha_m$ in R , with $m \leq n$, we can compute $g(\alpha_1), \dots, g(\alpha_m)$ using*

$$O((n/m + \log m)M(m))$$

operations in R , and additional space for $O(m)$ elements of R .

¹log denotes the natural logarithm, except that in “O”-estimates, we adopt the convention that log is defined and at least one for all positive numbers.

PROOF. We first compute the polynomial $P = (x - \alpha_1) \cdots (x - \alpha_m)$. Using a divide and conquer method, this takes $O(M(m) \log m)$ operations in R and space for $O(m)$ elements of R . Next, we compute $h = (g \bmod P)$ as follows. We write $g = \sum_{0 \leq i \leq n/m} g_i x^{mi}$, where the degree of each $g_i \in R[x]$ is less than m . Then we successively compute higher powers of x^m , multiply in the appropriate g_i , and reduce modulo P as we go. This takes $O((n/m)M(m))$ operations in R and space for $O(m)$ elements of R .

Finally, we evaluate h at the points $\alpha_1, \dots, \alpha_m$. This can be done using $O(M(m) \log m)$ operations in R using the algorithm in Aho *et al.* (1974, § 8.5), but that algorithm uses space for $O(m \log m)$ elements of R . We can circumvent this problem as follows. We split the points $\alpha_1, \dots, \alpha_m$ into about $\log m$ blocks, each of size about $m/\log m$. We use the algorithm in Aho *et al.* to evaluate h at the points in the first block, and then at the points in the second block, and so on. It is easy to verify that this method uses $O(M(m) \log m)$ operations in R and space for only $O(m)$ elements of R . \square

LEMMA 2.2. Let $f \in F_q[x]$ be a polynomial of degree n , and $R = F_q[x]/(f)$.

- (i) Two polynomials in $R[y]$ of degree at most m can be multiplied using $O(M(mn))$ operations in F_q .
- (ii) We can evaluate a polynomial in $R[y]$ of degree at most n at m points in R , where $m \leq n$, using

$$O((n/m + \log m)M(mn))$$

operations in F_q and space for $O(nm)$ elements in F_q .

PROOF. (i) Let $g, h \in R[y]$ be the polynomials to be multiplied. We can view the coefficients of g and h as elements of $F_q[x]$, compute the product in $F_q[x, y]$, and reduce each coefficient in the product modulo f . To compute the product polynomial in $F_q[x, y]$, we can perform the well-known Kronecker substitution $y \mapsto x^{2n-1}$, reducing our problem to that of multiplying two univariate polynomials in $F_q[x]$ of degree at most $(2n - 1)m$.

Using this method, the cost of computing the product (as a polynomial in $F_q[x, y]$) is $O(M(mn))$ operations in F_q , and the cost of reducing the coefficients in the product polynomial modulo f is $O(mM(n))$ operations in F_q , which is $O(M(mn))$.

Assertion (ii) follows directly from (i) and the proof of Lemma 2.1. \square

REMARK. The problem of multiplying polynomials in $R[y]$ of degree at most m is in fact asymptotically equivalent to the problem of multiplying polynomials in $F_q[x]$ of degree at most mn , in the sense that an algorithm for the former problem can be used to solve the latter problem using the same number of operations in F_q (up to a constant multiplicative factor).

3. A factoring algorithm

The purpose of this section is to describe and analyze an algorithm that will factor a polynomial $f \in F_q[x]$ of degree n with $O(n^2 + n \log q)$ operations in F_q .

The squarefree factorization of f is of the form $f = f_1 f_2^2 f_3^3 \cdots f_n^n$, where f_1, \dots, f_n are monic, squarefree, and pairwise relatively prime. Thus, f_i is the product of those monic irreducible polynomials in $F_q[x]$ that divide f exactly to the power i . Yun's (1976) algorithm computes this squarefree factorization with $O(M(n) \log n + n \log q)$ operations in F_q and space for $O(n)$ elements of F_q (see Knuth 1981, Exercise 4.6.2–36). So we may now assume that f is squarefree.

In most of the paper, we will use the following notation.

$$q \text{ is a prime power, } f \in F_q[x] \text{ is a squarefree monic polynomial of degree } n, R = F_q[x]/(f), \text{ and } \xi = (x \bmod f) \in R. \quad (3.1)$$

Our basic algorithms for distinct-degree and equal-degree factorization use as a subroutine an algorithm for computing iterates of the Frobenius map on R , which is based on the following observation. If for some $m \leq n$ we have computed $\xi, \xi^q, \dots, \xi^{q^m}$ and $g \in F_q[x]$ is the canonical representative of ξ^{q^m} , then we can calculate $\xi^{q^{m+1}}, \dots, \xi^{q^{2m}}$ by evaluating the polynomial g at the points ξ^q, \dots, ξ^{q^m} , since $g(\xi^{q^i}) = \xi^{q^{m+i}}$. This “doubling step” can be done using a fast multi-point evaluation algorithm with $O(n)$ operations in R . Thus, if we first compute ξ^q using repeated squaring—which takes $O(\log q)$ operations in R —and then perform the above “doubling step” $O(\log n)$ times, we can compute all of the elements $\xi, \xi^q, \dots, \xi^{q^n}$ with $O(n + \log q)$ operations in R , and hence $O(n^2 + n \log q)$ operations in F_q .

ALGORITHM 3.1. *Iterated Frobenius.*

Input: f, α, β , and m , where, in the notation (3.1), α and β are elements of R with $\beta = \xi^t$ for some power t of q , and m is a positive integer with $m \leq n$.

Output: The elements $\alpha, \alpha^t, \dots, \alpha^{t^m}$ in R .

1. Set $\gamma_0 = \xi \in R$, and $\gamma_1 = \beta$. Let $\ell = \lceil \log_2 m \rceil$.
2. For $i = 1, \dots, \ell$ do the following. [After stage i , $\gamma_j = \xi^{t^j}$ has been computed for $0 \leq j \leq 2^i$.]
 Compute $\gamma_{2^{i-1}+j} = \tilde{\gamma}_{2^{i-1}}(\gamma_j)$ for $j = 1, \dots, 2^{i-1}$ using a fast multi-point evaluation algorithm.
3. Compute $\gamma'_j = \tilde{\alpha}(\gamma_j)$ for $j = 0, \dots, m$ using a fast multi-point evaluation algorithm.
4. Return $\gamma'_0, \gamma'_1, \dots, \gamma'_m$.

THEOREM 3.2. Algorithm 3.1 works correctly as specified, and uses

$$O\left(\frac{n}{m} M(mn) \log m\right)$$

operations in F_q and space for $O(nm)$ elements of F_q . In particular, the running time can be bounded by

$$O(n^2 \log n \log \log n \log m)$$

operations in F_q .

PROOF. For the correctness, we first prove by induction on i the assertion made in step 2 of the algorithm, namely that $\gamma_j = \xi^{t^j}$ for $0 \leq j \leq 2^i$. We have for $1 \leq j \leq 2^{i-1}$

$$\begin{aligned} \gamma_{2^{i-1}+j} &= \tilde{\gamma}_{2^{i-1}}(\gamma_j) = \tilde{\gamma}_{2^{i-1}}(\xi^{t^j}) = (\tilde{\gamma}_{2^{i-1}}(\xi))^{t^j} \\ &= (\gamma_{2^{i-1}})^{t^j} = (\xi^{t^{2^{i-1}}})^{t^j} = \xi^{t^{2^i-1}+j}, \end{aligned}$$

and thus for $0 \leq j \leq m$

$$\gamma'_j = \tilde{\alpha}(\gamma_j) = \tilde{\alpha}(\xi^{t^j}) = (\tilde{\alpha}(\xi))^{t^j} = \alpha^{t^j}.$$

By Lemma 2.2(ii), stage i of step 2 of takes

$$O\left((n/2^i + i)M(2^i n)\right)$$

operations in F_q ; moreover, the cost of step 3 is asymptotically bounded by the cost of the last stage of step 2. Summing over $1 \leq i \leq \ell$ yields the running time bound in the statement of the theorem. The assertions about space are clear. \square

For distinct-degree factorization, we use a variant of an algorithm that already appears in Arwin (1918). As an aside, we remark that Arwin's fascinating paper foreshadows many other methods used in modern polynomial factorization, including trace computation (McEliece 1969, Ben-Or 1981, Camion 1983), reduction to root-finding via resultants (Berlekamp 1970), and distinguishing roots by their order (Moenck 1977, von zur Gathen 1987, Mignotte & Schnorr 1988, Menezes et al. 1992). Arwin's algorithm is based on the following fact (see Lidl & Niederreiter 1983, Theorem 3.20).

FACT 3.3. For $d \geq 1$, the polynomial $x^{q^d} - x \in F_q[x]$ is the product of the monic irreducible polynomials in $F_q[x]$ whose degree divides d .

ALGORITHM 3.4. Distinct-degree factorization.

Input: A squarefree monic polynomial f over F_q of degree n .

Output: The set of all pairs (g, d) such that g is the product of all monic irreducible factors of f of degree d with $g \neq 1$.

1. For ξ as in (3.1), compute $\xi^q \in R$ using a repeated squaring algorithm.
2. Using Algorithm 3.1 with arguments (f, ξ, ξ^q, n) , compute $\alpha_i = \xi^{q^i}$ for $i = 1, \dots, n$.
3. Put $S = \emptyset$, put $f^* = f$, and for $i = 1, \dots, n$, do the following.
 - (i) Compute $g = \gcd(f^*, \alpha_i - x)$.
 - (ii) If $g \neq 1$, add (g, i) to S .
 - (iii) Replace f^* with f^*/g .
4. Return S .

THEOREM 3.5. Algorithm 3.4 works correctly as specified, and uses

$$O(M(n^2) \log n + M(n) \log q)$$

operations in F_q and space for $O(n^2)$ elements of F_q . In particular, the running time can be bounded by

$$O((n^2 \log n + n \log q) \cdot \log n \log \log n)$$

operations in F_q .

PROOF. Note that $\tilde{\alpha}_i - x \equiv x^{q^i} - x \pmod{f}$, so that the correctness of the algorithm follows from Fact 3.3. The time and space bounds are easy to verify. \square

ALGORITHM 3.6. *Equal-degree factorization.*

Input: A squarefree polynomial f over F_q of degree n and a positive integer d , where f is the product of r irreducible factors, each of degree d (so that $n = rd$).

Output: The set of monic irreducible factors of f .

1. If $\deg f = d$, then return $\{f\}$; if $\deg f = 0$, then return \emptyset .

2. For ξ as in (3.1), compute ξ^q using repeated squaring.

3. Pick $\alpha \in R$ at random and compute

$$\beta = \sum_{0 \leq i < d} \alpha^{q^i},$$

using Algorithm 3.1 with arguments $(f, \alpha, \xi^q, d-1)$ to get $\alpha, \alpha^q, \dots, \alpha^{q^{d-1}}$, and then summing these elements.

4. If q is odd, then do the following:

(i) Compute $\gamma = \beta^{(q-1)/2}$ using repeated squaring.

(ii) Compute $h_1 = \gcd(\gamma, f)$, $h_2 = \gcd(\gamma - 1, f)$, and $h_3 = f/(h_1 h_2)$.

(iii) Recursively factor h_1 , h_2 , and h_3 , and return the union of these three sets of factors.

Else, if $q = 2^k$, do the following:

(i) Compute $\gamma = \sum_{0 \leq i < k} \beta^{2^i}$.

(ii) Compute $h_1 = \gcd(\gamma, f)$ and $h_2 = f/h_1$.

(iii) Recursively factor h_1 and h_2 , and return the union of these two sets of factors.

THEOREM 3.7. Algorithm 3.6 works correctly as specified, and uses an expected number of

$$O(M(nd)r \log d + M(n) \log r \log q)$$

operations in F_q , an expected number of $O(n \log r)$ random elements of F_q , and space for $O(nd)$ elements of F_q . In particular, the expected running time can be bounded by

$$O((n^2 \log d + n \log r \log q) \cdot \log n \log \log n)$$

operations in F_q .

PROOF. *Analyzing a single recursive invocation.* We consider first a single invocation of the body of this recursive algorithm. Step 2 takes $O(M(n) \log q)$ operations in F_q , step 3 takes $O(rM(nd) \log d)$ operations in F_q , and step 4 takes $O(M(n) \log n + M(n) \log q)$ operations in F_q .

Let $f_1, \dots, f_r \in F_q[x]$ be the irreducible factors of f . Then by the Chinese Remainder Theorem, we have an isomorphism from R onto $\bigoplus_{i=1}^r F_q[x]/(f_i) \cong \bigoplus_{i=1}^r F_{q^d}$ that maps F_q onto the diagonal. For $\alpha \in R$, suppose that $(\alpha_1, \dots, \alpha_r)$, with each $\alpha_i \in F_{q^d}$, is the image of α under this isomorphism.

For α randomly chosen in step 3, $\alpha_1, \dots, \alpha_r$ are independently and uniformly distributed random elements in F_{q^d} . If $T: F_{q^d} \rightarrow F_q$ denotes the trace, then $\beta_i = T(\alpha_i)$ for all i . Since T is F_q -linear and surjective, β_1, \dots, β_r are independently and uniformly distributed random elements in F_q .

In step 4, suppose first that q is odd. Each γ_i is 0 with probability $1/q$, and otherwise is ± 1 with equal probability. The polynomials h_1, h_2, h_3 are the products of those f_i with γ_i equal to 0, 1, -1 , respectively. If q is even, each component of γ is 0 or 1 with equal probability. In this case, h_1 and h_2 are the products of those f_i with γ_i equal to 0 and 1, respectively.

Analyzing the entire computation. One easily checks that the recursion tree has expected depth $O(\log r)$, since any pair of factors is "split" in one invocation with probability at least $1/2$. From this, one obtains a bound of

$$O(\log r \cdot rM(nd) \log d + M(n) \log r \log q)$$

on the expected number of operations in F_q . This proves the theorem, except for the additional factor of $\log r$ in the first summand.

We now give a more detailed analysis of the algorithm that eliminates this extra factor of $\log r$. We model the probabilistic behavior of the algorithm by a certain type of "balls and bins" game; we refer to the following section for a description of this game.

Suppose that q is odd. We calculate first the expected number of operations in F_q spent performing step 3 in *all* recursive invocations of the algorithm.

This corresponds to a balls and bins game in the following way. The “balls” correspond to the irreducible factors of f . The “bins” correspond to the polynomials h_1 , h_2 , and h_3 . A “ball landing in bin i ” corresponds to an irreducible factor of f dividing h_i . The probabilities associated with the three bins are $p_1 = 1/q$, $p_2 = (q-1)/2q$, and $p_3 = (q-1)/2q$.

The cost function g of the game is chosen so that $g(r)$ is an upper bound on the number of operations performed in step 3 in a single recursive invocation of the algorithm. We can take

$$g(r) = c \cdot rM(rd^2)(\log d + 1),$$

for an appropriately chosen constant c . Clearly, g satisfies the strong super-linearity condition of Lemma 4.1(ii), and we then find (with $\ell = 3$, $w = 1/2$, and $\epsilon = 1$) that the expected number of operations spent performing step 3 in all recursive invocations is $O(rM(nd)\log d)$.

By similar reasoning, we can bound the expected number of operations spent in steps 2 and 4 in all recursive invocations by $O(M(n)\log n \log r + M(n)\log r \log q)$. Note that we incur the extra factor of $\log r$ because the associated cost function will not necessarily satisfy the hypothesis of Lemma 4.1(ii), and so only (i) applies.

If q is even, we can argue similarly, but this time the behavior of the algorithm is modeled by a balls and bins game with two bins, each with associated probability $1/2$. \square

We summarize the results of this section in the following theorem, which is the main result of this paper.

THEOREM 3.8. *We have a probabilistic algorithm that factors a polynomial of degree n over F_q using an expected number of*

$$O((n^2 + n \log q) \cdot (\log n)^2 \log \log n)$$

operations in F_q .

4. A game of balls and bins

In the proof of Theorem 3.7, it is convenient to model the probabilistic behavior of Algorithm 3.6 by a balls and bins game of the following type. We have a fixed number ℓ of bins, and when a ball is tossed, the probability that it lands in bin i is p_i for $1 \leq i \leq \ell$. A ball must land in some bin, i.e., $\sum_{1 \leq i \leq \ell} p_i = 1$,

and $p_i \leq w < 1$ for all i . Associated with the game is a “cost function” $g: \mathbb{R}_{>0} \rightarrow \mathbb{R}_{>0}$.

The game is played with $r \geq 2$ balls as follows. First, the player pays $g(r)$ units. Second, the player throws the balls into the bins. Third, for each bin that contains at least two balls, the player recursively plays the game with the balls in that bin.

Let $G(r)$ be the total number of units that the player pays during the game, including the units paid in all recursive games. We want to estimate the expected value $E[G(r)]$ of $G(r)$.

LEMMA 4.1. (i) *If g grows at least linearly, i.e.,*

$$\forall t \in \mathbb{R}_{>0} \forall a \in \mathbb{R}_{\geq 1} \quad g(at) \geq ag(t),$$

then

$$\forall r \geq 2 \quad E[G(r)] \leq c_1 g(r) \log r,$$

where c_1 is a constant depending only on w .

(ii) *If g grows at a strongly super-linear rate, i.e., for some constant $\epsilon \in \mathbb{R}_{>0}$*

$$\forall t \in \mathbb{R}_{>0} \forall a \in \mathbb{R}_{\geq 1} \quad g(at) \geq a^{1+\epsilon} g(t),$$

then

$$\forall r \geq 2 \quad E[G(r)] \leq c_2 g(r),$$

where c_2 is a constant depending only on w , ℓ , and ϵ .

PROOF. Corresponding to any game is a tree representing all of the recursive games played, where each node in the tree is labeled with the number of units paid for that particular game. The quantity $G(r)$ is just the sum of all of the values labeling the nodes of the tree. Let D be the number of levels in the game tree.

We first bound the expected value $E[D]$ of D . For any pair of balls, the probability that they land in the same bin in any one game is at most w . Thus, the probability that this pair of balls has not been split after level number t in the game tree is at most w^t . Since the number of such pairs is less than r^2 , we have $\Pr[D \geq t] < r^2 w^{t-1}$. Then one calculates that

$$\begin{aligned} E[D] &= \sum_{t \geq 1} \Pr[D \geq t] \\ &\leq \frac{2 \log r}{\log(1/w)} + 1 + \sum_{t > 2 \log r / \log(1/w) + 1} r^2 w^{t-1} \\ &\leq \frac{2 \log r}{\log(1/w)} + 1 + 1/(1-w). \end{aligned}$$

Our assumption that g grows at least linearly implies that the values labeling the nodes on any one level of the tree sum to no more than $g(r)$. Thus, $E[G(r)] \leq g(r) \cdot E[D]$, and (i) follows with

$$c_1 = \frac{2}{\log(1/w)} + \frac{1 + 1/(1-w)}{\log 2}.$$

For (ii), we first note that $\sum_{1 \leq i \leq \ell} p_i^{1+\epsilon} \leq w^\epsilon < 1$. Let $\gamma = (1 + w^\epsilon)/2$. Clearly, we can choose $\delta > 0$ —depending only on w , ϵ and ℓ —in such a way that $\sum_{1 \leq i \leq \ell} (p_i + \delta)^{1+\epsilon} < \gamma$.

Call a toss of the balls “bad” if more than $p_i r + \delta r$ balls land in bin i for some $i \leq \ell$. By using well-known bounds on the tail of the binomial distribution (see Cormen *et al.* 1989, Corollary 6.7) one easily calculates that the probability of a bad toss is no more than $\ell e^{-\delta^2 r}$.

Let c_1 be as above, $A = \max\{c_1 \log r \ell e^{-\delta^2 r} : r \geq 2\}$, and $c_2 = (1+A)/(1-\gamma)$. We prove by induction on r that $E[G(r)] \leq c_2 g(r)$ for all $r \geq 2$. One easily verifies the base case $r = 2$. Then, for $r > 2$, we have

$$\begin{aligned} E[G(r)] &= E[G(r) \mid \text{bad toss}] \cdot \Pr[\text{bad toss}] \\ &\quad + E[G(r) \mid \text{good toss}] \cdot \Pr[\text{good toss}] \\ &\leq g(r) + c_1 g(r) \log r \cdot \ell e^{-\delta^2 r} + \sum_{1 \leq i \leq \ell} c_2 g(p_i r + \delta r) \\ &\leq g(r)(1+A) + c_2 g(r) \sum_{1 \leq i \leq \ell} (p_i + \delta)^{1+\epsilon} \\ &\leq g(r)(1+A+c_2 \gamma) = c_2 g(r). \quad \square \end{aligned}$$

5. Equal-degree factorization

In this section, we consider again the equal-degree factorization problem. We describe a method that is more efficient than Algorithm 3.6, in terms of both time and space. One variant of this algorithm uses $O^*(n^2 + n \log q)$ operations in F_q and space for $O(n)$ elements of F_q . Another variant uses $O^*(n^{(\omega+1)/2} + n \log q)$ operations in F_q and space for $O(n \log n)$ elements of F_q .

We now denote the cost of matrix multiplication by $MM(n)$ rather than n^ω ; that is, we assume that we have an algorithm that multiplies two $n \times n$ matrices over an arbitrary ring R using $O(MM(n))$ operations in R . We assume that the algorithm uses space for $O(n^2)$ elements of R , and that MM is actually a function $MM: R_{>0} \rightarrow R_{>0}$ such that

$$\forall t \in R_{>0} \forall a \in R_{\geq 1} \quad a^2 MM(t) \leq MM(at) \leq a^3 MM(t).$$

We can choose $MM(n) = n^{2.376}$ (Coppersmith & Winograd 1990).

A basic problem for the algorithms in this section is *modular composition*: computing $g(h) \bmod f$, where f , g , and h are polynomials over F_q . We summarize what is known about the complexity of this problem.

FACT 5.1. *Let f , g , and h be polynomials over F_q of degree at most n . Then we can compute $g(h) \bmod f$ in the following time and space bounds (time is measured in operations in F_q , and space is measured in storage for elements of F_q):*

- (i) time $O(nM(n))$ and space $O(n)$,
- (ii) time $O(n^{1/2}(M(n) + MM(n^{1/2})))$ and space $O(n^{3/2})$,
- (iii) time $O((MM(n^{1/2})n^{1/2} \log n + n^{3/2} \log n M(\ell) \ell^{-1} + M(n)(\log n)^3)$ and space $O(n \log n)$, where $\ell = \lceil \log n / \log q \rceil$.

PROOF. (i) is proved by viewing the problem as that of evaluating a polynomial of degree n at a single point in the ring $R = F_q[X]/(f)$. Using Horner's rule, this takes $O(n)$ operations in R and space for $O(1)$ elements of R . Algorithm 2.1 in Brent & Kung (1978) solves this problem for the special case where $f = x^n$ within the time and space bounds stated in (ii). In fact, this algorithm works as stated for arbitrary f . (iii) is proved in Shoup & Smolensky (1992). \square

The complexity of modular composition is not well understood. In the special case where $f = x^n$, Algorithm 2.2 in Brent & Kung (1978) solves this problem using $O^*(n^{3/2})$ operations. Moreover, no super-linear lower bounds are known for this problem.

In what follows we shall use $C_T(n)$ and $C_S(n)$ to denote the time and space bounds for a modular composition algorithm. We assume that C_T is a function $C_T: R_{>0} \rightarrow R_{>0}$ such that

$$\forall t \in R_{>0} \forall a \in R_{\geq 1} \quad a C_T(t) \leq C_T(at) \leq a^3 C_T(t).$$

We also assume that C_S is nondecreasing, and that $C_S(n) \geq n$.

The bottleneck in Algorithm 3.6 is the application of the iterated Frobenius algorithm in step 3 to compute $\alpha, \alpha^q, \dots, \alpha^{q^{d-1}}$. All we really need to compute is the sum of these elements, i.e., the trace map $\sum_{0 \leq i < d} \alpha^{q^i}$. We now present a different algorithm for computing trace maps, based on the following observation. Recall (3.1) and the notation $\bar{\alpha}$ for the canonical representative of an element α of R . For $m \geq 1$, let $\beta_m = \sum_{0 \leq i < m} \alpha^{q^i}$ and $\gamma_m = \xi^{q^m}$. Then $\beta_{2m} = \beta_m + \beta_m(\gamma_m)$

and $\gamma_{2m} = \tilde{\gamma}_m(\gamma_m)$. We can therefore obtain β_{2m} and γ_{2m} from β_m and γ_m at a cost of two modular compositions (plus one addition in R). Thus, if d is a power of 2, we can easily compute the trace map $\beta_d = \sum_{0 \leq i < d} \alpha^{q^i}$ as follows: first, compute ξ^q using repeated squaring, and then repeat the above “doubling step” $\log_2 d$ times. The total cost of this method is $O(n \log q)$ operations in F_q plus $O(\log d)$ modular compositions. The algorithm is easily adapted to deal with arbitrary d .

ALGORITHM 5.2. *Computing trace maps.*

Input: f , α , β , and m , where, in the notation (3.1), α and β are elements of R with $\beta = \xi^t$ for some power t of q , and m is a positive integer.

Output: The elements α^{t^m} and $\sum_{0 \leq j \leq m} \alpha^{t^{2^j}}$ in R .

1. Let $m = \sum_{0 \leq i \leq \ell} b_i 2^i$ be the binary representation of m , with $\ell = \lfloor \log_2 m \rfloor$ and $b_0, \dots, b_\ell \in \{0, 1\}$.

2. (stage 0):

$$\begin{aligned} \tau &\leftarrow \check{\alpha}(\beta), & \tau' &\leftarrow \begin{cases} \alpha & \text{if } b_0 = 0, \\ \alpha + \tau & \text{if } b_0 = 1, \end{cases} \\ \mu &\leftarrow \beta, & \mu' &\leftarrow \begin{cases} \xi & \text{if } b_0 = 0, \\ \beta & \text{if } b_0 = 1. \end{cases} \end{aligned}$$

3. For $i = 1, \dots, \ell$, do the following (stage i):

$$\begin{aligned} \tau &\leftarrow \tau + \check{\tau}(\mu), & \tau' &\leftarrow \begin{cases} \tau' & \text{if } b_i = 0, \\ \tau' + \check{\tau}(\mu') & \text{if } b_i = 1, \end{cases} \\ \mu &\leftarrow \check{\mu}(\mu), & \mu' &\leftarrow \begin{cases} \mu' & \text{if } b_i = 0, \\ \check{\mu}(\mu') & \text{if } b_i = 1. \end{cases} \end{aligned}$$

4. Return $\check{\alpha}(\mu')$ and τ' .

LEMMA 5.3. *Algorithm 5.2 works correctly as specified, and uses $O(C_T(n) \cdot \log m)$ operations in F_q and space for $O(C_S(n))$ elements of F_q .*

PROOF. The time and space bounds are clear, as the algorithm performs $O(\log m)$ modular compositions and additions of polynomials of degree at most n .

To prove correctness, let $m_i = \sum_{0 \leq j \leq i} b_j 2^j$ for $0 \leq i \leq \ell$, and let $\tau_i, \mu_i, \tau'_i, \mu'_i$ denote the values of the variables τ, μ, τ', μ' at the end of stage i . We show by

induction on i that the following identities hold:

$$\begin{aligned} \tau_i &= \sum_{1 \leq j \leq 2^i} \alpha^{t^j}, & \tau'_i &= \sum_{0 \leq j \leq m_i} \alpha^{t^j}, \\ \mu_i &= \xi^{t^{2^i}}, & \mu'_i &= \xi^{t^{m_i}}. \end{aligned}$$

The base case $i = 0$ is easily verified. Now assume that the identities hold after stage i ; we show that they also hold after stage $i + 1$. We have

$$\begin{aligned} \tau_{i+1} &= \tau_i + \check{\tau}_i(\mu_i) = \tau_i + \check{\tau}_i(\xi^{t^{2^i}}) = \tau_i + \tau_i^{t^{2^i}} \\ &= \sum_{1 \leq j \leq 2^i} \alpha^{t^j} + \sum_{2^i < j \leq 2^{i+1}} \alpha^{t^j} = \sum_{1 \leq j \leq 2^{i+1}} \alpha^{t^j}, \\ \mu_{i+1} &= \check{\mu}_i(\mu_i) = \check{\mu}_i(\xi^{t^{2^i}}) = \mu_i^{t^{2^i}} = \xi^{t^{2^{i+1}}}. \end{aligned}$$

If $b_{i+1} = 0$, it is clear that τ'_{i+1} and μ'_{i+1} will have the correct values, so assume that $b_{i+1} = 1$. Then we have

$$\begin{aligned} \tau'_{i+1} &= \tau'_i + \check{\tau}_{i+1}(\mu'_i) = \tau'_i + \check{\tau}_{i+1}(\xi^{t^{m_i}}) = \tau'_i + \tau_{i+1}^{t^{m_i}} \\ &= \tau'_i + \left(\sum_{1 \leq j \leq 2^{i+1}} \alpha^{t^j} \right)^{t^{m_i}} = \sum_{0 \leq j \leq m_i} \alpha^{t^j} + \sum_{m_i < j \leq m_i + 2^{i+1}} \alpha^{t^j} = \sum_{0 \leq j \leq m_{i+1}} \alpha^{t^j}, \\ \mu'_{i+1} &= \check{\mu}_{i+1}(\mu'_i) = \check{\mu}_{i+1}(\xi^{t^{m_i}}) = \mu'_{i+1} = \xi^{t^{m_i + 2^{i+1}}} = \xi^{t^{m_{i+1}}}. \quad \square \end{aligned}$$

We now modify Algorithm 3.6 by replacing step 3 with the following.

3'. Pick $\alpha \in R$ at random, and compute $\beta = \sum_{0 \leq i < d} \alpha^{q^i}$ using Algorithm 5.2 with arguments $(f, \alpha, \xi^q, d - 1)$.

THEOREM 5.4. *With the above modification, Algorithm 3.6 uses an expected number of*

$$O(C_T(n) \log d \log r + M(n) \log n \log r + M(n) \log r \log q)$$

operations in F_q and space for $O(C_S(n))$ elements of F_q . Moreover, if C_T is one of the bounds in Fact 5.1, then this modified version of Algorithm 3.6 uses an expected number of

$$O(C_T(n) \log d + M(n) \log r \log q)$$

operations in F_q . In particular, we have the following time and space bounds (time is measured in operations in F_q and space in storage for elements of F_q):

- (i) time $O(n^2 \log n \log \log n \log d + n \log n \log \log n \log r \log q)$ and space $O(n)$;
(ii) time $O(n^{1.7} + n \log n \log \log n \log r \log q)$ and space $O(n \log n)$.

PROOF. The proof is essentially the same as that of Theorem 3.7, and we omit the details. The exponent in (ii) comes from $(2.376 + 1)/2 = 1.688 < 1.7$.

□

6. Distinct-degree factorization

In this section, we describe another algorithm for distinct-degree factorization that uses $O(n^2 + n \log q)$ operations in F_q , like Algorithm 3.4, but space for only $O(n^{3/2})$ elements of F_q .

LEMMA 6.1. Let $m \leq n$, $g_1, \dots, g_m \in F_q[x, y]$ with $\deg_x g_i \leq 1$ and $\deg_y g_i < n$ for all $i \leq m$, and $h \in F_q[y]$ be monic with $\deg h \leq n$. Then

$$P = \left(\prod_{1 \leq i \leq m} g_i \right) \bmod h \in F_q[x, y]$$

can be computed with

$$O(M(mn) \log m)$$

operations in F_q and space for $O(mn)$ elements of F_q .

PROOF. The polynomial P is obtained by taking the remainder modulo h of each coefficient in $\prod_{1 \leq i \leq m} g_i$ at a power of x .

To compute it, we use a binary tree with $\lceil \log_2 m \rceil$ levels and g_1, \dots, g_m at the leaves, i.e., level zero. At each node, we multiply the two children and take the remainder modulo h of each coefficient of a power of x . The degree in x at level $j - 1$ is at most 2^{j-1} , and that in y is less than n . The claims follow from Lemma 2.2(i), with the rôles of x and y interchanged and $R = F_q[x]/(h)$. □

We now describe a subroutine used by our distinct-degree factorization algorithm.

ALGORITHM 6.2. Range decomposition.

Input: f , a , b , α , β , where a and b are positive integers with $a \leq b$, and $\alpha = \xi^{a^q}$ and $\beta = \xi^{b^q}$, in the notation (3.1); it is assumed that the degrees of all irreducible factors of f lie in the range a, \dots, b .

Output: The set of all pairs (g, d) such that g is the product of all monic irreducible factors of f of degree d with $g \neq 1$.

1. If $a > n$, return \emptyset ; if $b > n$, replace b with n .
2. Use Algorithm 3.1 with arguments $(f, \alpha, \beta, b - a)$ to compute $\alpha_i = \xi^{a^i}$ for $i = a, \dots, b$.
3. Put $S = \emptyset$, $f^* = f$, and do the following for $i = a, \dots, b$.
 - (i) Compute $g = \gcd(f^*, \alpha_i - x)$.
 - (ii) If $g \neq 1$, add (g, i) to S .
 - (iii) Replace f^* with f^*/g .
4. Return S .

LEMMA 6.3. Algorithm 6.2 works correctly as specified, and uses

$$O\left(\frac{n}{m} M(mn) \log m\right)$$

operations in F_q and space for $O(nm)$ elements of F_q , where $m = \min\{n, b - a + 1\}$. In particular, the running time can be bounded by

$$O(n^2 \log n \log \log n \log m)$$

operations in F_q .

PROOF. The correctness follows from Fact 3.3. The time and space bounds can be easily checked. □

Our distinct-degree factorization algorithm works by dividing the interval $1, \dots, n$ into roughly \sqrt{n} intervals each of size about \sqrt{n} . For each interval, it finds the product of the irreducible factors whose degree lies in the that interval, and then applies Algorithm 6.2 to get the complete distinct-degree factorization.

ALGORITHM 6.4. Distinct-degree factorization.

Input: A squarefree polynomial $f \in F_q[x]$ of degree n .

Output: The set of all pairs (g, d) such that g is the product of all monic irreducible factors of f of degree d with $g \neq 1$.

1. In the notation (3.1), compute ξ^q using repeated squaring.
2. Let $m = \lfloor n^{1/2} \rfloor$, and $m' = \lceil n/m \rceil$.

3. Apply Algorithm 3.1 with arguments (f, ξ, ξ^q, m) to obtain $\mu_j = \xi^{q^j}$ for $j = 0, \dots, m$.
4. Apply Algorithm 3.1 with arguments $(f, \xi^q, \mu_m, m' - 1)$ to obtain $\nu_i = \xi^{q^{im+1}}$ for $i = 0, \dots, m' - 1$.
5. Compute the polynomial

$$g(y) = \prod_{0 \leq j < m} (\mu_j(y) - \xi) \bmod f(y) \in R[y]$$

using the method described in Lemma 6.1. [Then $g(y) \equiv \prod_{0 \leq j < m} (y^{q^j} - \xi) \bmod f(y)$.]

6. Compute $\tau_i = g(\nu_i) \in R$ for $i = 0, \dots, m' - 1$ using a fast multi-point evaluation algorithm. [Then $\tau_i = \prod_{0 \leq j < m} (\xi^{q^{im+1+j}} - \xi)$.]
7. Put $S = \emptyset$, $f^* = f$, and for $i = 0, \dots, m' - 1$, do the following:
 - (i) Compute $h = \gcd(\tau_i, f^*)$.
 - (ii) Replace f^* with f^*/h .
 - (iii) If $h \neq 1$, compute the distinct-degree factorization of h by invoking Algorithm 6.2 with arguments $(h, im + 1, im + m, \nu_i, \mu_1)$ and add the result to S .
8. Return S .

THEOREM 6.5. Algorithm 6.4 works correctly as specified, and uses

$$O(M(n^{3/2})n^{1/2} \log n + M(n) \log q)$$

operations in F_q and space for $O(n^{3/2})$ elements of F_q . In particular, the running time can be bounded by

$$O((n^2 \log n + n \log q) \cdot \log n \log \log n)$$

operations in F_q .

PROOF. We first prove correctness. The only subtle point is that the assertion made in step 6 holds because we are evaluating the polynomial $g(y)$ at points

ν_i that are zeros of $f(y)$. Thus, $g(\nu_i) = \prod_{0 \leq j < m} (\nu_i^{q^j} - \xi)$. Consider the i th iteration of the loop in step 7. We have

$$\tau_i \equiv \prod_{0 \leq j < m} (x^{q^{im+1+j}} - x) \pmod{f}.$$

It follows from Fact 3.3 that the polynomial h computed in the i th iteration is the product of those irreducible factors of f whose degree lies in the range $im + 1, \dots, im + m$. The invocation of Algorithm 6.2 then correctly computes the distinct-degree decomposition of h .

We now estimate the running time, measured in terms of operations in F_q .

- Step 1 takes time $O(M(n) \log q)$.
- Steps 3 and 4 take time $O(M(n^{3/2})n^{1/2} \log n)$, by Theorem 3.2.
- Step 5 takes time $O(M(n^{3/2}) \log n)$, by Lemma 6.1.
- Step 6 takes time $O(M(n^{3/2})n^{1/2})$, by Lemma 2.2(ii).
- The total time spent in steps 7(a) and 7(b) is $O(n^{1/2}M(n) \log n)$.

Implicit in step 7(c) is the computation at the i th iteration of $\mu_1 \bmod h$ and $\nu_i \bmod h$. The total time spent doing this is $O(n^{1/2}M(n))$. Suppose that Algorithm 6.2 is invoked with polynomials of degree n_i for $0 \leq i < m'$; then $\sum_{0 \leq i < m'} n_i = n$. The i th invocation takes time $O(M(n^{1/2}n_i)n^{1/2} \log n)$. The sum of this quantity over all values of i is $O(M(n^{3/2})n^{1/2} \log n)$, which also bounds the total time for step 7.

Thus the running time is bounded by $O(M(n^{3/2})n^{1/2} \log n + M(n) \log q)$, and the assertion about space is easy to check. \square

As things now stand, distinct-degree factorization is the bottleneck in our general polynomial factoring algorithm. Furthermore, the bottleneck in our distinct-degree factorization algorithm is multi-point polynomial evaluation. We now indicate a direction in which further progress in reducing the time complexity of distinct-degree factorization—and hence the complexity of the general factoring problem—may be possible.

We define the *special multi-point polynomial evaluation problem* to be that of evaluating a “large” polynomial with “small” coefficients at a “small” number of “large” points. To be precise, let $f \in F_q[x]$ have degree n , and let $R = F_q[x]/(f)$. The special multi-point polynomial evaluation problem is that of computing $g(\alpha_1), \dots, g(\alpha_m)$, where $m \leq n^{1/2}$, $\alpha_1, \dots, \alpha_m \in R$, and g is a

polynomial of degree n over R whose coefficients each have canonical representatives of degree at most $n^{1/2}$.

Since the number of elements of F_q required to encode both the input and the output of this problem is only $O(n^{3/2})$, it is not a priori clear that this problem cannot be solved using, say, $O(n^{3/2})$ operations in F_q . However, at present the best algorithms we know of take $O(n^2)$ operations in F_q . All of the multi-point polynomial evaluation problems that arise in Algorithm 6.4 are of this special type. As the next theorem indicates, any improvement in solving this special problem will immediately lead to an improvement in the timing of distinct-degree factorization.

THEOREM 6.6. *Suppose that the special multi-point evaluation problem can be solved using $O(n^\kappa)$ operations in F_q , with $3/2 \leq \kappa \leq 2$. Then Algorithm 6.4 can be implemented with $O(n^\kappa + n \log q)$ operations in F_q .*

PROOF. Excluding steps 3, 4 and 6, and the invocations of Algorithm 6.2, Algorithm 6.4 uses $O(n^{3/2} + n \log q)$ operations in F_q . Using the algorithm for special multi-point evaluation, steps 3, 4, and 6 can be implemented with $O(n^\kappa)$ operations in F_q .

Now suppose we have a polynomial of degree n over R whose coefficients have canonical representatives of degree at most $n^{1/2}$, and we want to evaluate it at t points, where $t > n^{1/2}$. Using the algorithm for special multi-point evaluation, this can be done using $O((t/n^{1/2})n^\kappa)$ operations in F_q by processing the points $n^{1/2}$ at a time.

From the observation in the previous paragraph, if at iteration i of the loop in step 7, Algorithm 6.2 is invoked with a polynomial of degree n_i , then the number of operations in F_q used in this invocation is $O(n^{1/2}n_i^{\kappa-1/2})$. The sum of this quantity over all i is $O(n^\kappa)$, since $\kappa \geq 3/2$. Thus the total time of the algorithm is $O(n^\kappa + n \log q)$ operations in F_q . \square

We conjecture that $\kappa < 2$ can be achieved.

7. Applications to other problems in finite fields

In this section, we use the techniques of the previous sections to obtain new algorithms for several problems in finite fields.

PROBLEM 7.1. *Given a polynomial of degree n over F_q , determine if it is irreducible.*

PROBLEM 7.2. *Given a polynomial of degree n over F_q , determine if all of its irreducible factors are distinct and have the same degree, and if so, determine the degree.*

Problem 7.2 is interesting since our algorithm for the equal-degree factorization problem is more efficient than our algorithm for factoring a general polynomial. Therefore, we want to be able to quickly test if we are in this special case, and if so, determine the degree of the irreducible factors since this is required as input to our equal-degree factoring algorithm. We use a well-known criterion for irreducibility (see Knuth 1981, Exercise 4.6.2-16).

FACT 7.3. *A polynomial $f \in F_q[x]$ of degree n is irreducible if and only if $x^{q^n} - x \equiv 0 \pmod{f}$ and $\gcd(f, x^{q^{n/t}} - x) = 1$ for all primes t dividing n .*

The following is an easy generalization, the proof of which we leave to the reader.

FACT 7.4. *Let f be a polynomial of degree n over F_q . The irreducible factors of f are all distinct and have the same degree if and only if $x^{q^n} - x \equiv 0 \pmod{f}$ and for all prime powers t dividing n , $\gcd(f, x^{q^{n/t}} - x)$ is either 1 or f . Moreover, if these conditions are met, then each irreducible factor of f has degree $d = n / \prod_t t$, where t ranges over the set of maximal prime power divisors of n such that $x^{q^{n/t}} - x \equiv 0 \pmod{f}$.*

We can use Algorithm 5.2 to compute $x^{q^m} \pmod{f}$ for the various values of m in the above two facts. In Fact 7.3, their number is bounded by the number of distinct prime divisors of n , which is $O(\log n / \log \log n)$ (see Hardy & Wright 1984, § 22.10). In Fact 7.4, this number is bounded by the number of prime power divisors of n , which is $O(\log n)$; however, we can achieve the same bound $O(\log n / \log \log n)$ by performing a binary search on the exponents of the prime factors. Recalling that C_T and C_S are time and space bounds for the modular composition problem (see Fact 5.1), we have the following result.

THEOREM 7.5. *Problems 7.1 and 7.2 can be solved by algorithms using*

$$O((C_T(n) + M(n))(\log n)^2 / \log \log n + M(n) \log q)$$

operations in F_q and space for $O(C_S(n))$ elements of F_q .

In particular, we have the following time and space bounds (time is measured in operations in F_q and space is measured in storage for elements of F_q):

- (i) time $O(n^2(\log n)^3 + n \log n \log \log n \log q)$ and space $O(n)$;
 (ii) time $O(n^{1.7} + n \log n \log \log n \log q)$ and space $O(n \log n)$.

Related to the problem of testing irreducibility is that of constructing an irreducible polynomial of given degree n over F_q . A probabilistic algorithm of Rabin (1980) for this problem uses $O(n^3 \log q)$ operations in F_q , and another random method due to Ben-Or (1981) uses $O(n^2 \log q)$ operations, and Shoup (1993) presents an $O(n^2 + n \log q)$ method.

PROBLEM 7.6. Find a normal basis for F_{q^n} over F_q , i.e., an F_q -vector space basis of the form $\alpha, \alpha^q, \dots, \alpha^{q^{n-1}}$ for F_{q^n} .

In von zur Gathen & Giesbrecht (1990), this problem is solved probabilistically with $O(n^2 \log q)$ operations in F_q . The core of that algorithm is to compute $\beta_i = \alpha^{q^i} \in F_{q^n}$ for $1 \leq i < n$ and $\gcd(\sum_{0 \leq i < n} \beta_i y^i, y^n - 1) \in F_{q^n}[y]$, for $O(\log n)$ many random $\alpha \in F_{q^n}$. Using Algorithm 3.1 to calculate these coefficients, we obtain the following improvement.

THEOREM 7.7. Problem 7.6 can be solved by a probabilistic algorithm using an expected number

$$O(M(n^2)(\log n)^2 + M(n) \log q),$$

or

$$O((n^2(\log n)^2 + n \log q) \cdot \log n \log \log n)$$

of operations in F_q and space for $O(n^2)$ elements of F_q .

Similarly, one can determine the additive order of an element of F_{q^n} with $O((n^2 \log n + n \log q) \cdot \log n \log \log n)$ operations in F_q .

Interestingly, our algorithm is not the fastest way to compute what is usually called a trace, namely in a field extension.

PROBLEM 7.8. Compute the trace $T_{F_{q^n}/F_q}(\alpha) = \sum_{0 \leq i < n} \alpha^{q^i} \in F_q$, given $f \in F_q[x]$ monic and irreducible of degree n , and $\alpha \in F_{q^n} = F_q[x]/(f)$.

Write $T = T_{F_{q^n}/F_q}$ and $\xi = (x \bmod f) \in F_{q^n}$, and let $\tilde{f} = \prod_{0 \leq i < n} (1 - \xi^{q^i} y) \in F_q[y]$ be the reverse of f . The expansion of the logarithmic derivative gives the following equation in $F_{q^n}[[y]]$:

$$y \tilde{f}' / \tilde{f} = - \sum_{1 \leq i} \sum_{0 \leq j < n} \xi^{iq^j} y^i = - \sum_{1 \leq i} T(\xi^i) y^i$$

(see Lidl & Niederreiter 1983, §5.2). The first n coefficients of the left-hand side are easy to calculate with a Newton iteration (see Borodin & Munro 1975, §4.4), and give $T(\xi^i)$ for $0 \leq i < n$. Since

$$T\left(\sum_{0 \leq i < n} a_i \xi^i\right) = \sum_{0 \leq i < n} a_i T(\xi^i)$$

for $a_0, \dots, a_{n-1} \in F_q$, T can then be evaluated as an inner product.

THEOREM 7.9. Problem 7.8 can be solved with $O(M(n))$ operations in F_q and space for $O(n)$ elements of F_q .

8. Using classical arithmetic

We have presented a factoring algorithm that uses $O(n^2 + n \log q)$ operations in F_q . To achieve this running time, we have used asymptotically fast algorithms for polynomial arithmetic. One might ask of what value our methods are in the range of values for n in which classical polynomial arithmetic algorithms are faster than the asymptotically fast algorithms. We attempt to answer this question in this section. All running time estimates stated here assume classical arithmetic is used—that is, we assume that $M(n) = n^2$ and $MM(n) = n^3$.

The pioneering methods of Berlekamp (1970) and Cantor & Zassenhaus (1981) give factorization algorithms that use either $O(n^3 \log q)$ operations in F_q and space for $O(n)$ elements of F_q , or $O(n^3 + n^2 \log q)$ operations in F_q and space for $O(n^2)$ elements of F_q .

Using the above values for $M(n)$ and $MM(n)$ in Fact 5.1(iii), one finds that modular composition of polynomials over F_q of degree n can be done using classical arithmetic with $O(n^2(\log n)^3)$ operations in F_q . In fact, this is a slight over-estimate, and the running time of this algorithm is actually $O((n \log n)^2)$ operations in F_q . The space requirement is $O(n \log n)$ elements of F_q .

As a consequence, by implementing the Frobenius map with modular composition, we can solve the distinct-degree factorization problem by successively computing the iterates of the Frobenius map one at a time using $O(n^3(\log n)^2 + n^2 \log q)$ operations in F_q and space for $O(n \log n)$ elements of F_q . Algorithm 5.2 solves the problem of equal-degree factorization with $O(n^2(\log n)^3 + n^2 \log q)$ operations in F_q and space for $O(n \log n)$ elements of F_q .

It is a bit difficult to interpret the meaning of these results. After all, an algorithm that uses only classical arithmetic might be so complicated in other respects that even for small inputs it is no faster than a different algorithm that uses asymptotically fast arithmetic. However, these results do suggest

that our techniques could perhaps be used to implement an algorithm that is more efficient than other algorithms in its use of time or space for inputs of moderate size. For equal-degree factorization, preliminary experiments seem to indicate considerable improvements.

9. Deterministic equal-degree factorization

All of the algorithms we have discussed for squarefree and distinct-degree factorization are deterministic, and those for equal-degree factorization are probabilistic. In this section, we solve the latter problem with a *deterministic* procedure that combines the methods of the previous sections with those in Shoup (1991).

Suppose that the polynomial $f \in \mathbb{F}_q[x]$ to be factored is the product of r distinct monic irreducible polynomials, each of the same degree d , so that $n = \deg f = dr$. We denote by p the characteristic of \mathbb{F}_q , and write $q = p^k$. We assume that we have an element $\eta \in \mathbb{F}_q$ such that $\mathbb{F}_q = \mathbb{F}_p(\eta)$, i.e., the elements $1, \eta, \dots, \eta^{k-1}$ are linearly independent over \mathbb{F}_p . In the usual representation of \mathbb{F}_q as $\mathbb{F}_p[z]/(g)$, with $g \in \mathbb{F}_p[z]$ irreducible of degree k , we can take $\eta = (z \bmod g)$.

We describe and analyze a deterministic algorithm for this problem. The number of operations in \mathbb{F}_q in one variant of this algorithm is

$$O^-(ndk + np^{1/2} \min\{dk, r\}), \quad (9.2)$$

and in a second variant, it is

$$O^-(n^2 + nk \min\{d, r\} + np^{1/2} \min\{dk, r\}). \quad (9.3)$$

Combining the second variant with one of our distinct-degree factorization algorithms, we obtain a general factorization algorithm that is deterministic and uses

$$O^-(n^2 + n^{3/2}k + n^{3/2}k^{1/2}p^{1/2})$$

operations in \mathbb{F}_q , since $\min\{d, r\} \leq n^{1/2}$ and $\min\{dk, r\} \leq (nk)^{1/2}$. Thus, in the special case where p is bounded, this algorithm uses $O^-(n^2 + n^{3/2}k)$ operations in \mathbb{F}_q .

At the end of this section, we briefly mention another variant that extracts a single irreducible factor of an arbitrary polynomial using

$$O^-(n^2 + nk + np^{1/2})$$

operations in \mathbb{F}_q . Thus, when p is bounded, this is $O^-(n^2 + nk)$, which is as fast as our probabilistic algorithm to within logarithmic factors.

As usual, we write $f = f_1 \cdots f_r$ and $R = \mathbb{F}_q[x]/(f)$. Under the Chinese Remainder isomorphism of R onto $\bigoplus_{1 \leq i \leq r} \mathbb{F}_{q^d}$, we identify an element $\alpha \in R$ with its component representation $(\alpha_1, \dots, \alpha_r)$, where each α_i is in \mathbb{F}_{q^d} , and we call α_i the i th component of α .

The (*absolute*) *Berlekamp subalgebra* A of R consists of those elements in R whose components lie in the prime field \mathbb{F}_p . A set $S \subseteq R$ is called a *separating set* if for all i, j with $1 \leq i < j \leq r$ there exists $\alpha \in S$ whose i th and j th components are not equal.

Computing a separating set in A lies at the heart of all known deterministic factoring algorithms. Consider two distinct factors f_i and f_j . If $S \subseteq A$ is a separating set, then we know that for some $\alpha \in S$ with component representation $(\alpha_1, \dots, \alpha_r)$, we have $\alpha_i \neq \alpha_j$. Therefore, the polynomial $\tilde{\alpha} - \alpha_i \in \mathbb{F}_q[x]$ is divisible by f_i but not f_j , and its gcd with f splits f_i and f_j apart. Now, given α , we do not know the values of its components, but we do know that they lie in \mathbb{F}_p , so by trying all values of a in \mathbb{F}_p , we can find a polynomial $\tilde{\alpha} - a$ that is divisible by one of f_i or f_j but not both.

The procedure just described requires time proportional to p in the worst case. If p is large we can speed this process up as follows. Compute $\beta = (\alpha - z)^{(p-1)/2}$ for $z = 0, 1, 2, \dots$. The i th component of β is $\chi(\alpha_i - z)$, where χ is the quadratic character on \mathbb{F}_p . Thus, if $\chi(\alpha_i - z) \neq \chi(\alpha_j - z)$, then either $\tilde{\beta}$ or $\tilde{\beta} - 1$ is divisible by f_i or f_j but not both. It was shown in Shoup (1990) that for any pair of distinct elements $a, b \in \mathbb{F}_p$, the least nonnegative integer z such that $\chi(a - z) \neq \chi(b - z)$ is bounded by $O(p^{1/2} \log p)$; this bound was subsequently improved in Shparlinski (1992) to $O(p^{1/2})$. Thus, the number of values of z we need to try in order to separate f_i from f_j is $O(p^{1/2})$.

If p is small and the size of S is not too large, we can use these observations to obtain an efficient algorithm to completely factor f . However, when p is small it is the computation of a separating set in A that is actually the bottleneck in deterministic factoring algorithms. Our main technical contribution in this section is a faster method for computing a separating set.

Let $T: R \rightarrow R$ be the trace map that sends $\alpha \in R$ to $\alpha + \alpha^p + \dots + \alpha^{p^{k-1}}$.

THEOREM 9.1. *Let*

$$H = (y - \xi)(y - \xi^q) \cdots (y - \xi^{q^{d-1}}) = h_0 + h_1 y + \dots + h_{d-1} y^{d-1} + y^d \in R[y].$$

Then the set

$$S = \{T(\eta^v h_u) : 0 \leq v < k, 0 \leq u < d\} \subseteq A$$

is a separating set.

PROOF. Let B be the subring of R consisting of those elements whose components lie in F_q . We claim the following two properties:

1. Each h_u lies in B .
2. $\{h_0, \dots, h_{d-1}\}$ is a separating set.

Let $(h_{u,1}, \dots, h_{u,r})$ be the component representation of h_u for $0 \leq u < d$, and let $f_i = x^d + \sum_{0 \leq u < d} a_{u,i} x^u$ with each $a_{u,i}$ in F_q . Let (ξ_1, \dots, ξ_r) be the component representation of ξ . Then we have $f_i = \prod_{0 \leq u < d} (x - \xi_i^u)$, from which it follows that $h_{u,i} = a_{u,i}$ for $0 \leq u < d$ and $1 \leq i \leq r$. The claims now easily follow.

The map T acts component-wise on B as the trace T_{F_q/F_p} from F_q to F_p . So it is clear that $S \subseteq A$. Given i and j with $1 \leq i < j \leq r$, we choose h_u such that $h_{u,i} \neq h_{u,j}$. Then, since T_{F_q/F_p} is an F_p -linear map from F_q onto F_p , and since $1, \eta, \dots, \eta^{k-1}$ form a basis for F_q over F_p , we see that for some $v < k$ we have $T_{F_q/F_p}(\eta^v h_{u,i}) \neq T_{F_q/F_p}(\eta^v h_{u,j})$. For this u and v , the i th and j th components of $T(\eta^v h_u)$ are not equal, and hence S is a separating set. \square

The observation that the coefficients of H form a separating set was made in Shoup (1990). We now address the complexity of computing the elements in this separating set.

THEOREM 9.2. Let $H \in R[y]$ be the polynomial in Theorem 9.1.

- (i) There is a deterministic algorithm that computes the coefficients of H using

$$O(M(dn) \log d + M(n)d \log q)$$

operations in F_q , and space for $O(dn)$ elements of F_q .

- (ii) There is a deterministic algorithm that computes the coefficients of H using

$$O(M(dn)r \log d + M(n) \log q)$$

operations in F_q , and space for $O(dn)$ elements of F_q .

- (iii) There is a deterministic algorithm that takes as input an element $\alpha \in R$ and computes $T(\eta^v \alpha)$ for $0 \leq v < k$ using

$$O(M(n) \log q + nM(k) \log k)$$

operations in F_q , and space for $O(kn)$ elements of F_q .

PROOF. We can compute $\xi, \xi^q, \dots, \xi^{q^{d-1}}$, where $\xi = (x \bmod f) \in R$, by repeated squaring using $O(d \log q)$ operations in R , and hence $O(M(n)d \log q)$ operations in F_q . Alternatively, we can just compute ξ^q using repeated squaring, and then apply Algorithm 3.1. We can form the product $\prod_{0 \leq u < d} (y - \xi^{q^u}) \in R[y]$ using a divide-and-conquer algorithm with $O(M(dn) \log d)$ operations in F_q . This proves (i) and (ii). For (iii), we compute the quantities

$$\eta_w = \eta^{p^w} \text{ and } \alpha_w = \alpha^{p^w} \quad \text{for } 0 \leq w < k$$

by repeated squaring using $O(M(n) \log q)$ operations in F_q . Then for $0 \leq v < k$, we have

$$T(\eta^v \alpha) = \sum_{0 \leq w < k} \eta_w^v \alpha_w.$$

The next section provides an algorithm that given the η_w 's and the α_w 's computes all of these *weighted power sums* with $O(M(k) \log k)$ operations in R , and hence $O(nM(k) \log k)$ operations in F_q , since the multiplications involve at most one element outside of F_q (Theorem 10.4). That algorithm also uses space for $O(k)$ elements of R . \square

Part (iii) of this theorem says that for a given coefficient h_u of H , we can compute $T(\eta^v h_u)$ for $0 \leq v < k$ using $O(n \log q)$ operations in F_q . To compute the entire separating set S in Theorem 9.1, we could repeat this for each of the d coefficients of H . This would require $O(nd \log q)$ operations in F_q , which for fixed p is optimal up to logarithmic factors, but is more than we can allow if we want to achieve the running time bound (9.3). However, as we shall see below, it is possible to avoid computing S in its entirety, and it is sufficient to consider $\min\{d, r\}$ coefficients of H . Thus, we can compute the required portion of S using only $O(\min\{d, r\}n \log q)$ operations in F_q (plus $O(dn)$ to determine that portion).

Before we describe our deterministic equal-degree factorization algorithm, we need a few more observations. A set $U \subseteq F_q[x]$ of monic nonconstant polynomials is called a *refinement* of f if $f = \prod_{g \in U} g$. We use the following two operations.

1. *Refine*(U, α), where U is a refinement of f and $\alpha \in R$. This replaces U by the refinement U' obtained in the following way: for each $g \in U$, if $g_1 = \gcd(g, \alpha)$ is a trivial divisor of g , put g in U' ; otherwise, put g_1 and g/g_1 in U' .
2. *Useful*(U, α), where U is a refinement of f and $\alpha \in R$. This evaluates to *true* if $(\alpha \bmod g) \notin F_q$ for some $g \in U$; otherwise, it evaluates to *false*.

The following lemma addresses the cost of these operations.

LEMMA 9.3. Operation $\text{Refine}(U, \alpha)$ can be performed using $O(M(n) \log n)$ operations in F_q and space for $O(n)$ elements of F_q . The predicate $\text{Useful}(U, \alpha)$ can be evaluated using $O(M(n) \log r)$ operations in F_q and space for $O(n)$ elements of F_q .

PROOF. To implement either of these operations, we begin by computing $\alpha \bmod g$ for all $g \in U$. Since $\#U \leq r$, this can be done with $O(M(n) \log r)$ operations in F_q and space for $O(n)$ elements in F_q in a manner similar to the proof of Lemma 2.1. This gives us the bound for computing $\text{Useful}(U, \alpha)$. The algorithm for $\text{Refine}(U, \alpha)$ then proceeds by computing $\gcd(\alpha \bmod g, g)$ for all $g \in U$. This takes $O(M(n) \log n)$ operations in F_q and space for $O(n)$ elements in F_q . \square

ALGORITHM 9.4. Deterministic equal-degree factorization.

Input: A squarefree monic polynomial f over F_q of degree n , a positive integer d , and an element $\eta \in F_q$. It is assumed that f is the product of r irreducible factors, each of degree d (so that $n = rd$), and that $F_q = F_{p^k} = F_p(\eta)$.

Output: The set of monic irreducible factors of f .

1. In the notation (3.1), compute the coefficients of

$$H = (y - \xi)(y - \xi^q) \cdots (y - \xi^{q^{d-1}}) \\ = h_0 + h_1 y + \cdots + h_{d-1} y^{d-1} + y^d \in R[y].$$

2. Set $U = \{f\}$.
3. For $u = 0, \dots, d-1$ do step 4.
4. If $\text{Useful}(U, h_u)$, then do steps 5 and 6.
5. Compute $(\alpha_0, \alpha_1, \dots, \alpha_{k-1}) = (T(h_u), T(\eta h_u), \dots, T(\eta^{k-1} h_u))$.
6. For $v = 0, \dots, k-1$ do steps 7 and 8.
7. Set $z = 0$.
8. While $\text{Useful}(U, \alpha_v)$ do steps 9 through 11.
9. $\text{Refine}(U, \alpha_v - z)$.
10. If $p \neq 2$, then $\text{Refine}(U, (\alpha_v - z)^{(p-1)/2} - 1)$.
11. Replace z by $z + 1$.
12. Return U .

THEOREM 9.5. Algorithm 9.4 works correctly as specified. If step 1 is implemented using Theorem 9.2(i), then it uses

$$O(M(dn) \log d + M(n)d \log(rq) \\ + \min\{d, r\}[nM(k) \log k + M(n)k \log r] + \min\{dk, r\}p^{1/2}M(n) \log(np))$$

or, more briefly,

$$O(ndk + np^{1/2} \min\{dk, r\})$$

operations in F_q . If step 1 is implemented using Theorem 9.2(ii), then it uses

$$O(\min\{d, r\}[nM(k) \log k + M(n)k \log(pr)] \\ + M(dn)r \log d + \min\{dk, r\}p^{1/2}M(n) \log(np))$$

or, more briefly,

$$O(n^2 + nk \min\{d, r\} + np^{1/2} \min\{dk, r\})$$

operations in F_q . In either case, it uses space for $O((d+k)n)$ elements of F_q .

PROOF. It follows immediately from Theorem 9.1 that the algorithm would work correctly if the two Useful tests were left out. However, their insertion only avoids unnecessary work; this proves correctness.

To estimate the running time, notice that if $\text{Useful}(U, h_u)$ at step 4 is true, then execution of steps 5 and 6 produces a proper refinement of U ; hence, steps 5 and 6 are executed at most $\min\{d, r\}$ times. Similarly, the inner loop at steps 9 through 11 is executed at most $O(p^{1/2} \min\{dk, r\})$ times.

The statements about the running time follow from these observations, along with the estimates from Theorem 9.2. The space bound is clear. \square

REMARK. Notice that Algorithm 9.4, using Theorem 9.2(ii) to implement step 1, discovers a nontrivial factor after executing just $O(n^2 + nk + np^{1/2})$ operations in F_q . This implies that there is a deterministic algorithm that takes as input an arbitrary polynomial $f \in F_q[x]$ of degree n , and produces as output a single irreducible factor of f using $O(n^2 + nk + np^{1/2})$ operations in F_q .

10. Generalized power sums

The algorithm in the proof of Theorem 9.2(iii) uses as a subroutine an algorithm for the following general problem: given $\alpha_0, \dots, \alpha_{k-1}$ and $\beta_0, \dots, \beta_{k-1}$ in a ring R , compute the *weighted power sums*

$$\gamma_\ell = \sum_{0 \leq i < k} \alpha_i^\ell \beta_i \quad \text{for } 0 \leq \ell < k.$$

In this section, R is an arbitrary commutative ring with unity. This problem is equivalent to computing the matrix-vector product $V^T b$, where $V \in R^{k \times k}$ is the Vandermonde matrix with $V_{ij} = \alpha_i^{j-1}$ and $b = (\beta_0, \dots, \beta_{k-1})^T \in R^{k \times 1}$. We shall present an algorithm for this problem that uses $O(M(k) \log k)$ operations in R , and space for $O(k)$ elements in R .

An algorithm for this problem is described in Canny *et al.* (1989). It uses $O(M(k) \log k)$ operations in R , and it is possible—using Lemma 2.1—to implement this algorithm so that it uses space for $O(k)$ elements in R . This algorithm works by first computing $c = V^{-1}b$, and then applying the Hankel matrix $V^T V$ to c . In their application, R is a field and the α_i 's are distinct. The use of this algorithm over an arbitrary ring R is hindered by the fact that it performs several divisions by elements in R , and it is not clear that these can easily be avoided.

There are general methods by which one can transform an algebraic circuit for computing the matrix-vector product Mb into an algebraic circuit of about the same size for computing $M^T b$ (Baur & Strassen 1983, Kaminski *et al.* 1988). Since computing Vb is known to take $O(M(k) \log k)$ operations in R , the same bound applies to computing $V^T b$. The algorithms that result from these general transformations require space proportional to their running times.

Either of these methods would have sufficed to obtain the running time bound in Theorem 9.2(iii); however, our algorithm below is still perhaps of interest in itself, since it avoids divisions, requires space for only $O(k)$ elements of R , and has a fairly natural description. We start with an algorithm for computing the first nontrivial elements of a linearly recurrent sequence.

LEMMA 10.1. Suppose we are given elements c_1, \dots, c_k and $\gamma_0, \dots, \gamma_{k-1}$ in R . For $\ell \geq k$ let γ_ℓ be defined by the recurrence

$$\gamma_\ell + c_1 \gamma_{\ell-1} + \dots + c_k \gamma_{\ell-k} = 0. \quad (10.4)$$

Then we can compute $\gamma_k, \dots, \gamma_{2k-1}$ using $O(M(k))$ operations in R and space for $O(k)$ elements in R .

PROOF. Define the polynomials

$$G = \sum_{0 \leq j \leq k} c_j x^j, \quad U = \sum_{0 \leq j < k} \gamma_j x^j, \quad V = \sum_{0 \leq j < k} \gamma_{k+j} x^j,$$

where c_0 is taken to be 1. Now consider the products

$$GU = \sum_{0 \leq j < 2k} u_j x^j, \quad GV = \sum_{0 \leq j < 2k} v_j x^j.$$

Clearly (10.4), for $k \leq \ell < 2k$, is equivalent to the condition that

$$G \cdot (U + x^k V) \text{ rem } x^{2k}$$

have degree less than k . Let $GU = F_0 + x^k F_1$, where F_0 and F_1 are polynomials of degree at most $k-1$. Then we can rewrite this as

$$GV \equiv -F_1 \text{ mod } x^k.$$

Let $H \in R[x]$ have degree less than k and satisfy $HG \equiv 1 \text{ mod } x^k$; this exists since $c_0 = 1$. Then we have

$$V \equiv -HF_1 \text{ mod } x^k.$$

Thus we can compute H by a Newton iteration, find F_1 from GU , and obtain the desired outputs from $V = -HF_1 \text{ rem } x^k$. All this can be done within the stated time and space bounds. \square

Our algorithm for computing weighted power sums is based on the following lemma.

LEMMA 10.2. Let $\alpha_0, \dots, \alpha_{k-1}$ and $\beta_0, \dots, \beta_{k-1}$ be in R . Define $\gamma_\ell \in R$ for $\ell \geq 0$, $G \in R[x]$, and $c_0, \dots, c_k \in R$ by

$$\gamma_\ell = \sum_{0 \leq i < k} \alpha_i^\ell \beta_i, \quad G = \prod_{0 \leq j < k} (1 - \alpha_j x) = \sum_{0 \leq j \leq k} c_j x^j \in R[x].$$

Then

$$\gamma_\ell + c_1 \gamma_{\ell-1} + \dots + c_k \gamma_{\ell-k} = 0 \quad \text{for all } \ell \geq k.$$

PROOF. Let

$$\tilde{G} = \sum_{0 \leq j \leq k} c_j x^{k-j} = \prod_{0 \leq j < k} (x - \alpha_j).$$

Then for $\ell \geq k$, we have

$$\begin{aligned} \sum_{0 \leq j \leq k} c_j \gamma_{\ell-j} &= \sum_{0 \leq j \leq k} c_j \sum_{0 \leq i < k} \alpha_i^{\ell-j} \beta_i \\ &= \sum_{0 \leq i < k} \alpha_i^{\ell-k} \beta_i \sum_{0 \leq j \leq k} c_j \alpha_i^{k-j} = \sum_{0 \leq i < k} \alpha_i^{\ell-k} \beta_i \tilde{G}(\alpha_i) = 0. \quad \square \end{aligned}$$

ALGORITHM 10.3. Weighted power sums.

Input: Elements $\alpha_0, \dots, \alpha_{k-1}$ and $\beta_0, \dots, \beta_{k-1}$ in R , where k is a power of 2.

Output: $\gamma_0, \dots, \gamma_{k-1}$ and c_0, \dots, c_k , as defined in Lemma 10.2.

1. If $k = 1$, then return β_0 and $(1 - \alpha_0 x)$.
2. Otherwise, divide the problem into two pieces of size $k/2$, and recursively compute the following quantities:
 - (i) $\gamma'_\ell = \sum_{0 \leq i < k/2} \alpha_i^\ell \beta_i$ for $0 \leq \ell < k/2$,
 - (ii) the coefficients of $G' = (1 - \alpha_0 x) \cdots (1 - \alpha_{k/2-1} x)$
 - (iii) $\gamma''_\ell = \sum_{k/2 \leq i < k} \alpha_i^\ell \beta_i$ for $0 \leq \ell < k/2$,
 - (iv) the coefficients of $G'' = (1 - \alpha_{k/2} x) \cdots (1 - \alpha_{k-1} x)$
3. Extend the sequences $(\gamma'_0, \dots, \gamma'_{k/2-1})$ and $(\gamma''_0, \dots, \gamma''_{k/2-1})$ to the corresponding sequences of length k using the coefficients of G' and G'' , and the algorithm of Lemma 10.1.
4. For $0 \leq \ell < k$, set $\gamma_\ell = \gamma'_\ell + \gamma''_\ell$, and compute the coefficients of $G = G'G''$.
5. Return $\gamma_0, \dots, \gamma_{k-1}$ and the coefficients of G .

THEOREM 10.4. Algorithm 10.3 works correctly as specified, and uses $O(M(\log k))$ operations in R and space for $O(k)$ elements in R .

The proof is straightforward. In this algorithm, we assumed for simplicity that k is a power of 2. Of course this restriction can easily be removed, either by modifying the algorithm slightly, or by padding the input.

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