

Modern ECC signatures

2011 Bernstein–Duif–Lange–Schwabe–Yang:

Ed25519 signature scheme = EdDSA using conservative Curve25519 elliptic curve.

<https://ed25519.cr.yp.to>

32-byte public keys,

64-byte signatures,

$\approx 2^{125.8}$ security level.

Deployed in SSH, Signal, many more applications:

<https://ianix.com/pub>

[/ed25519-deployment.html](https://ianix.com/pub/ed25519-deployment.html)

Many papers have explored Curve25519/Ed25519 speed.

e.g. 2015 Chou software:
on Intel Sandy Bridge (2011),
57164 cycles for keygen,
63526 cycles for signature,
205741 cycles for verification,
159128 cycles for ECDH.

Compare to, e.g., 2000 Brown–Hankerson–López–Menezes:
on Intel Pentium II (1997),
1920000 cycles for ECDH
using NIST P-256 curve.

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No! Beware change in CPU.

Maybe $A_C > B_C$; $A_D > B_D$;

C does more work per cycle than

D , thanks to CPU manufacturer.

Sometimes people measure cost

in seconds instead of cycles.

Then they benefit

from more work per cycle and

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For each of these operations
on each of these curves,
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Simplest implementations
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Questions in algorithm design
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How to build the fastest software
on, e.g., an ARM Cortex-A8.
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Answers feed back into cryptosystem
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Point
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 $x_1, x_2 \mapsto$

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 questions):
 Pentium II/III
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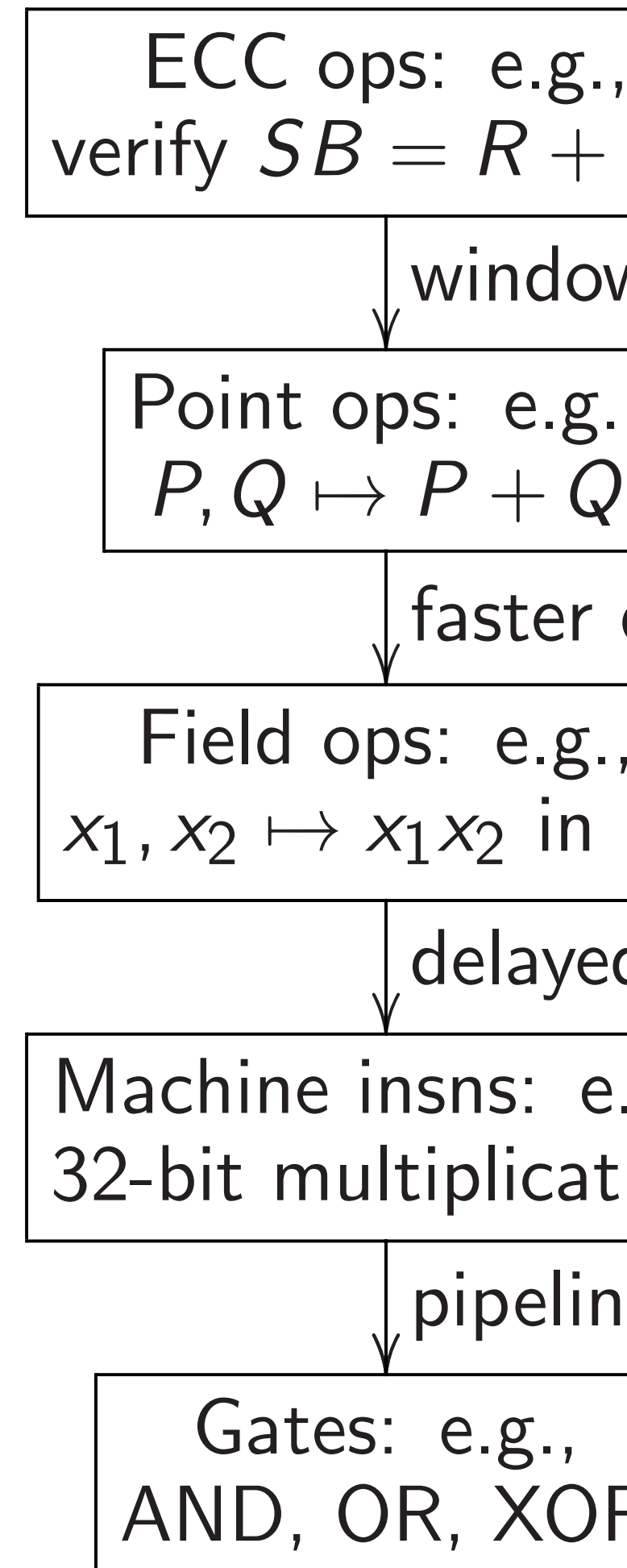
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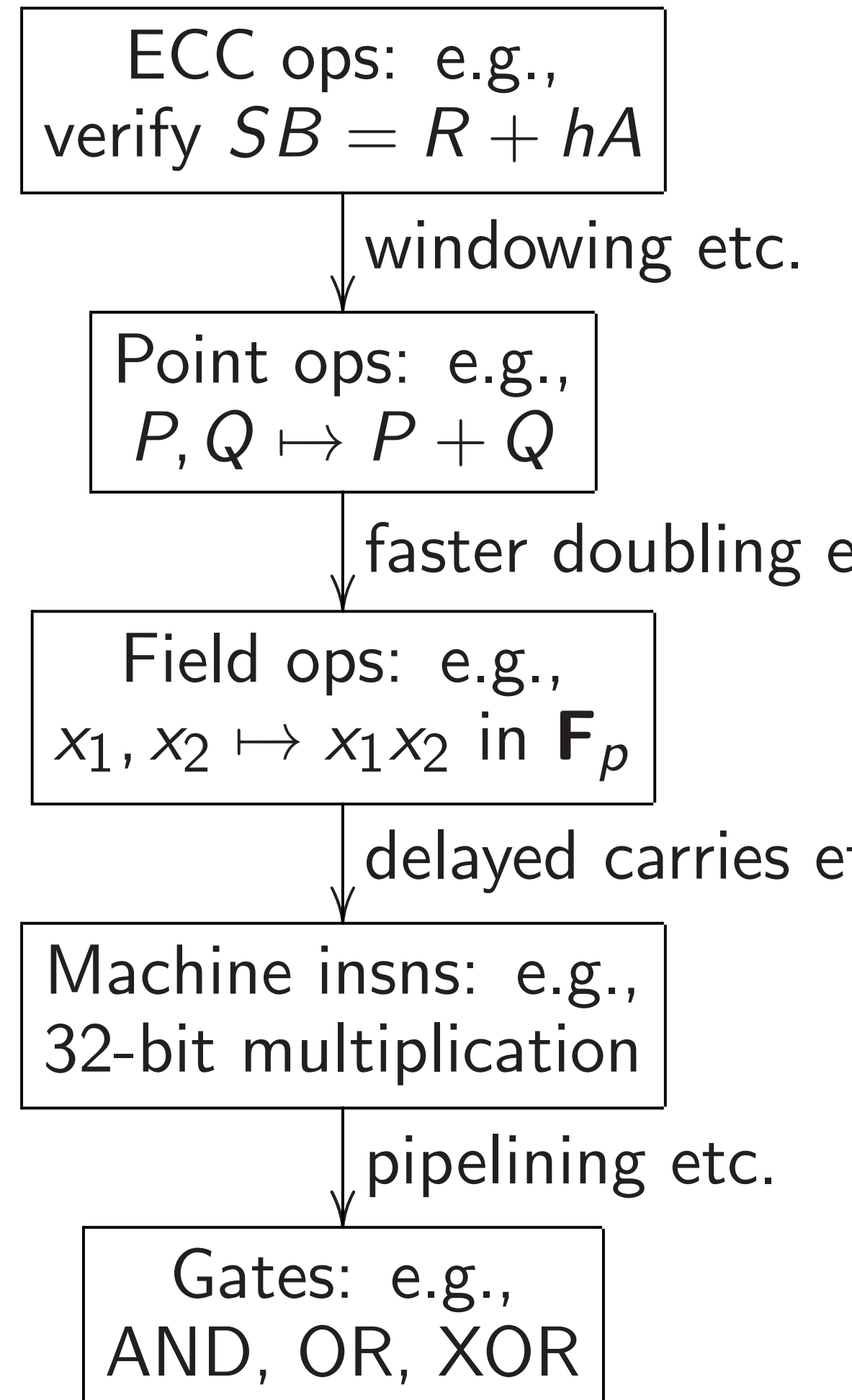
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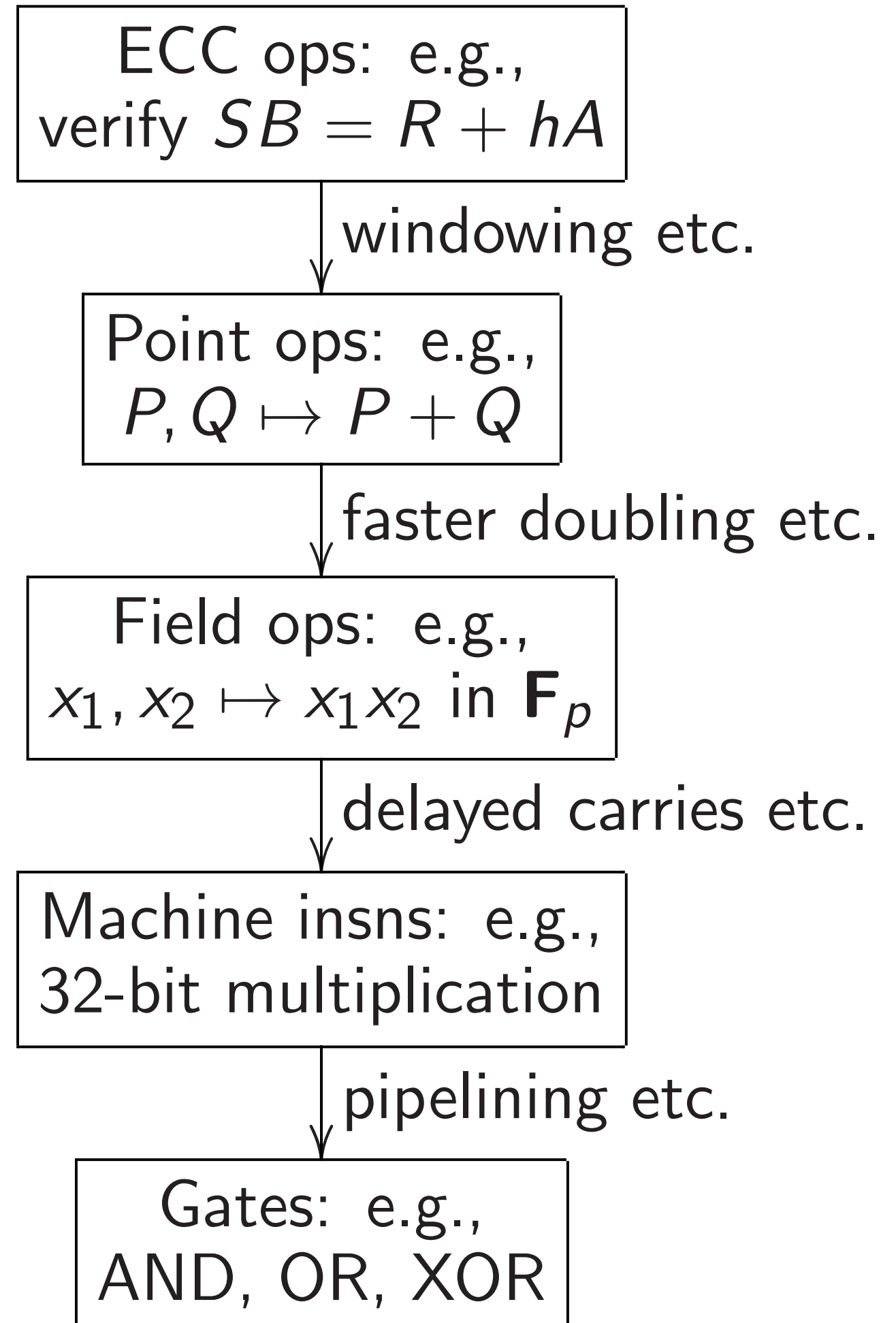
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Several levels to optimize:

ECC ops: e.g.,
verify $SB = R + hA$

↓ windowing etc.

Point ops: e.g.,
 $P, Q \mapsto P + Q$

↓ faster doubling etc.

Field ops: e.g.,
 $x_1, x_2 \mapsto x_1 x_2$ in \mathbf{F}_p

↓ delayed carries etc.

Machine insns: e.g.,
32-bit multiplication

↓ pipelining etc.

Gates: e.g.,
AND, OR, XOR

Single-sc

Fundam

$n, P \mapsto$

Input n

$\{0, 1, \dots$

Input P

Will build

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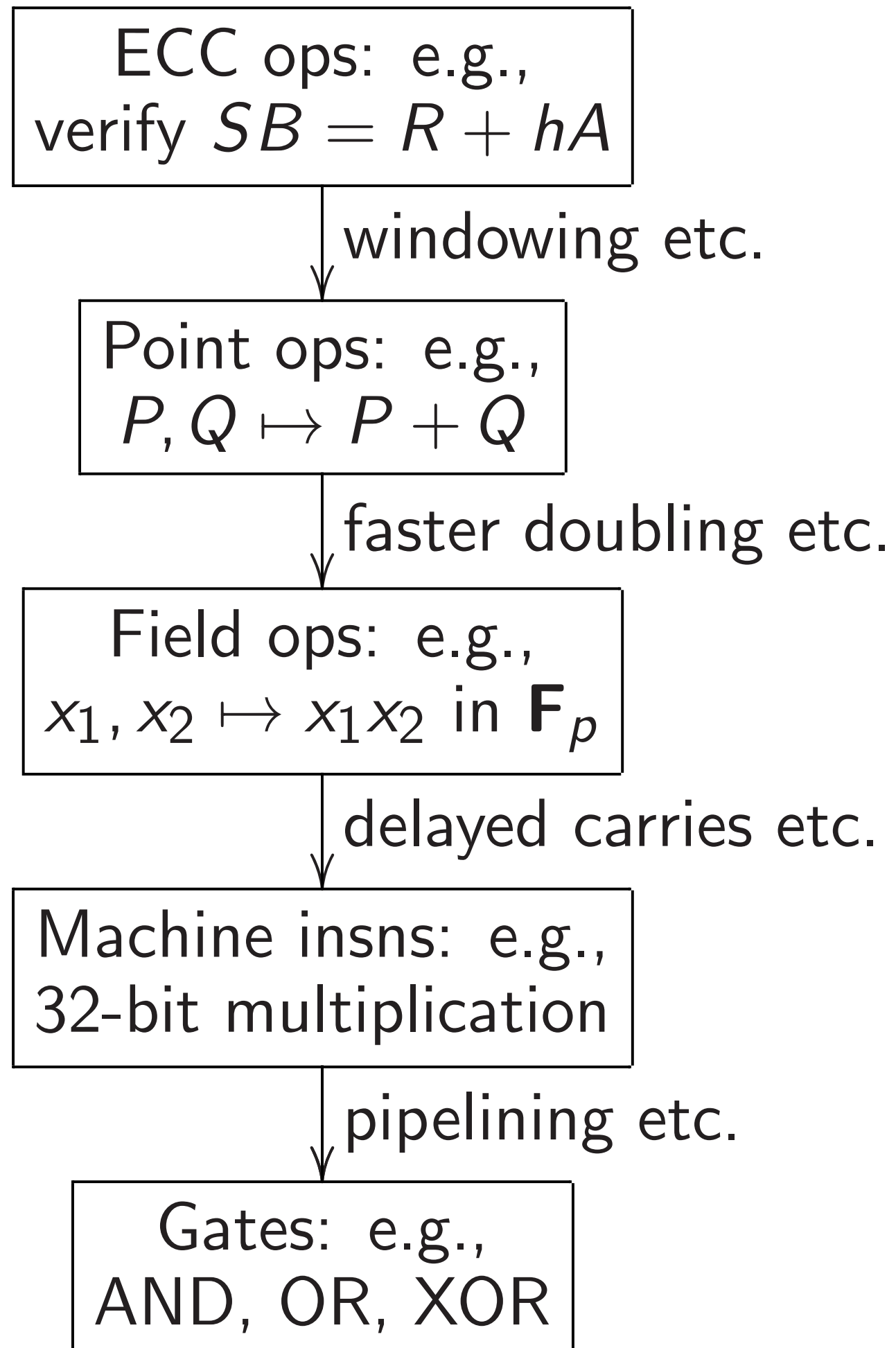
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m, P, n, C

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Several levels to optimize:



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Single-scalar multi

Fundamental ECC
 $n, P \mapsto nP$.

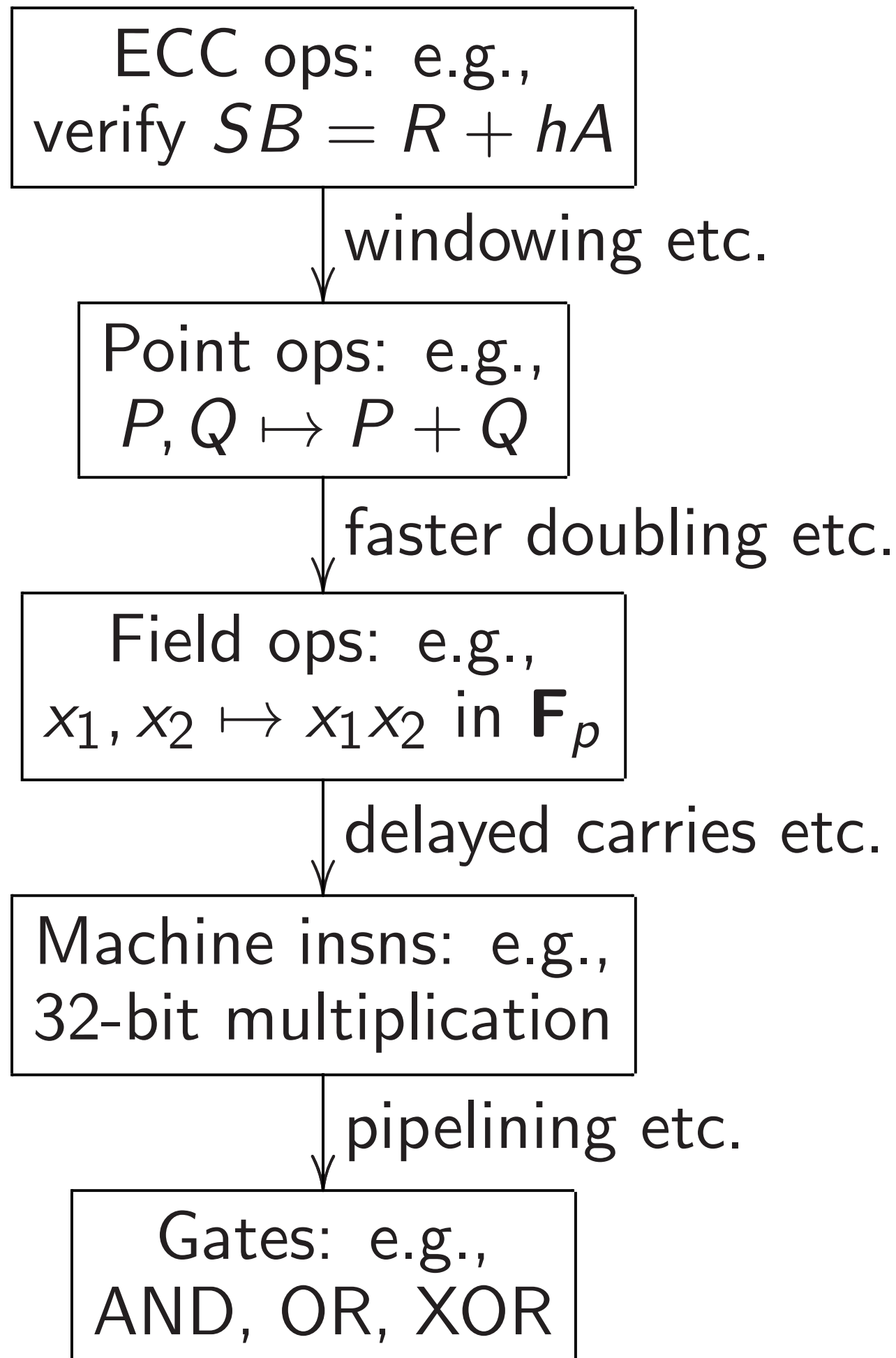
Input n is integer
 $\{0, 1, \dots, 2^{256} - 1$

Input P is point o

Will build $n, P \mapsto$
 using additions $P,$
 and subtractions P

Later will also look
 double-scalar mult
 $m, P, n, Q \mapsto mP -$

Several levels to optimize:



Single-scalar multiplication

Fundamental ECC operation
 $n, P \mapsto nP$.

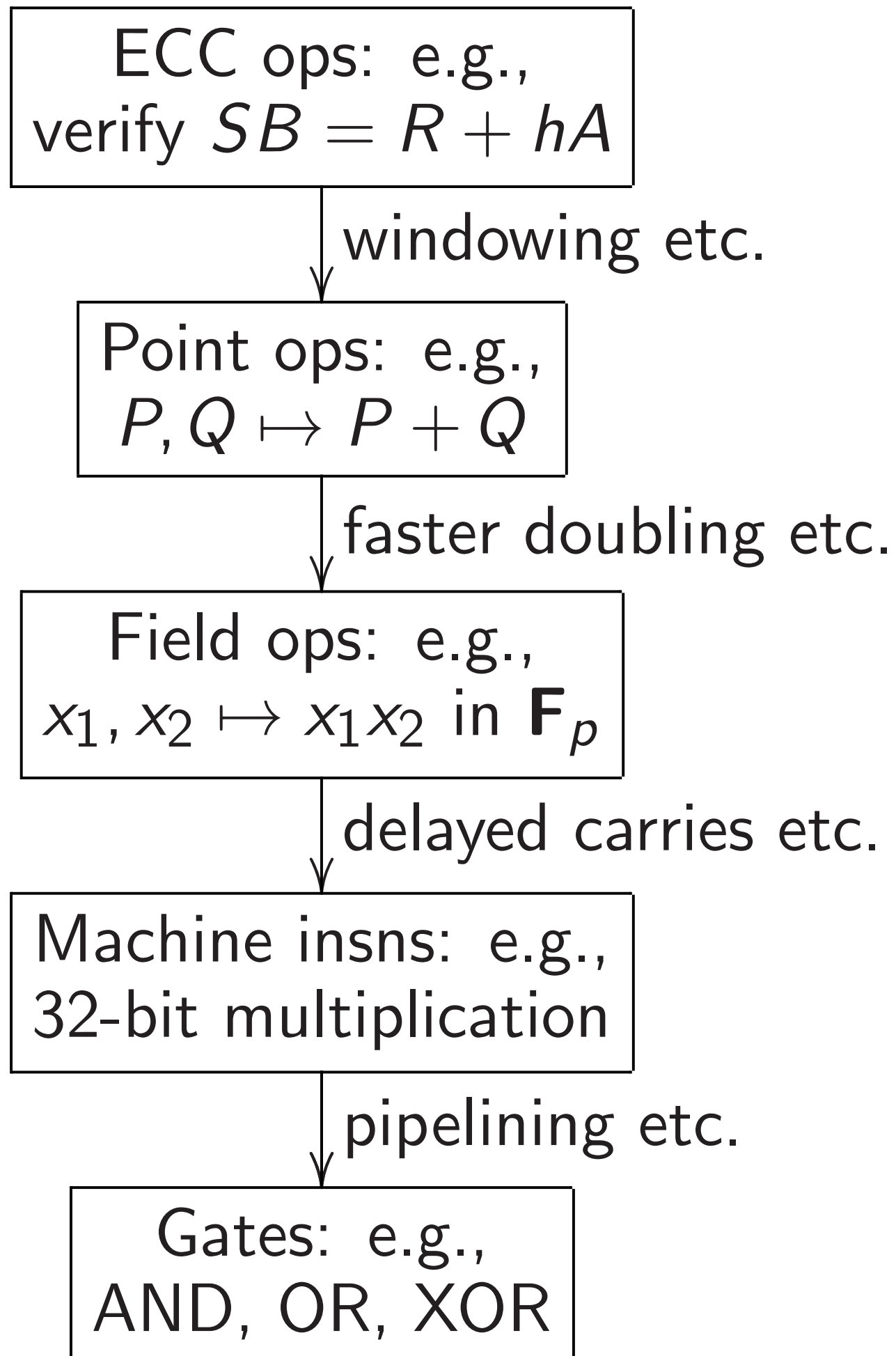
Input n is integer in, e.g.,
 $\{0, 1, \dots, 2^{256} - 1\}$.

Input P is point on elliptic curve

Will build $n, P \mapsto nP$
 using additions $P, Q \mapsto P + Q$
 and subtractions $P, Q \mapsto P - Q$

Later will also look at
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double-scalar multiplication

$$m, P, n, Q \mapsto mP + nQ.$$

Left-to-right

```
def scalar
```

```
    if n =
```

```
    if n =
```

```
    R = s
```

```
    R = R
```

```
    if n >
```

```
    return
```

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- $n // 2$ i

- Recurs

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double-scalar multiplication

$$m, P, n, Q \mapsto mP + nQ.$$

Left-to-right binary

```
def scalarmult(n, P):
    if n == 0: return O
    if n == 1: return P
    R = scalarmult(n // 2, P)
    R = R + R
    if n % 2: R = R + P
    return R
```

Two Python notes

- `n//2` in Python
- Recursion depth
See `sys.setrecursionlimit`

Single-scalar multiplication

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$$m, P, n, Q \mapsto mP + nQ.$$

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- $n//2$ in Python means $\lfloor n/2 \rfloor$
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e.g. $20P$
- $2 \left(\frac{n-1}{2} P \right) + P$
e.g. $21P$

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operation:

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nP

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- $2 \left(\frac{n}{2} P \right)$ if $n \in 2\mathbb{Z}$
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Left-to-right binary method

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This recursion computes nP

- $2 \binom{n}{2} P$ if $n \in 2\mathbf{Z}$.
e.g. $20P = 2 \cdot 10P$.
- $2 \binom{n-1}{2} P + P$ if $n \in 1 + 2\mathbf{Z}$.
e.g. $21P = 2 \cdot 10P + P$.

Base cases in recursion:

$0P = 0$. For Edwards: $0 = 0$
 $1P = P$. Could omit this case.

Assuming $n \geq 0$ for simplicity.
 Otherwise use $nP = -(-nP)$.

Left-to-right binary method

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def scalarmult(n,P):
    if n == 0: return 0
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right binary method

```
def armult(n,P):
```

```
    if n == 0: return 0
```

```
    if n == 1: return P
```

```
    R = armult(n//2,P)
```

```
    R = R + R
```

```
    if n % 2: R = R + P
```

```
    return R
```

Python notes:

`n//2` in Python means $\lfloor n/2 \rfloor$.

Recursion depth is limited.

```
sys.setrecursionlimit.
```

8

This recursion computes nP as

- $2 \binom{n}{2} P$ if $n \in 2\mathbf{Z}$.

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9

If $0 \leq n$

this algo

$$\leq 2b - 2$$

$$\leq b - 1$$

$$\leq b - 1$$

Example

$$31P = 2$$

$$31 = (1$$

4 doublings

Average

$$35P = 2$$

$$35 = (10$$

5 doublings

by method

,P):

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This recursion computes nP as

- $2 \left(\frac{n}{2} P \right)$ if $n \in 2\mathbf{Z}$.

e.g. $20P = 2 \cdot 10P$.

- $2 \left(\frac{n-1}{2} P \right) + P$ if $n \in 1 + 2\mathbf{Z}$.

e.g. $21P = 2 \cdot 10P + P$.

Base cases in recursion:

$0P = 0$. For Edwards: $0 = (0, 1)$.

$1P = P$. Could omit this case.

Assuming $n \geq 0$ for simplicity.

Otherwise use $nP = -(-n)P$.

If $0 \leq n < 2^b$ then

this algorithm uses

$\leq 2b - 2$ additions

$\leq b - 1$ doublings

$\leq b - 1$ additions

Example of worst

$31P = 2(2(2(2P +$

$31 = (11111)_2$; $b = 5$

4 doublings; 4 more

Average case is be

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9

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Non-adjacent form

```
def scalarmult(n
    if n == 0: ret
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    if n % 4 == 1:
        R = scalarmu
        R = R + R
        return (R +
    if n % 4 == 3:
        R = scalarmu
        R = R + R
        return (R +
    R = scalarmult
    return R + R
```

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Non-adjacent form (NAF)

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def scalarmult(n,P):
    if n == 0: return 0
    if n == 1: return P
    if n % 4 == 1:
        R = scalarmult((n-1)/2,P)
        R = R + R
        return (R + R) + P
    if n % 4 == 3:
        R = scalarmult((n+1)/2,P)
        R = R + R
        return (R + R) - P
    R = scalarmult(n/2,P)
    return R + R
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```

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```
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```
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    return R + R
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algorithm uses

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```
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```

```
        return (R + R) - P
```

```
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```
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```

Subtract

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$$31P = 2(2(2(2P)))$$

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Worst case: $\approx b$ d
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Subtraction on the curve is as cheap as addition.

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$$31P = 2(2(2(2(2P)))) - P.$$

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On average $\approx b/3$ additions.

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Adjacent form (NAF)

```

def scalarmult(n,P):
    if n == 0: return 0
    if n == 1: return P
    if n % 4 == 1:
        R = scalarmult((n-1)/4,P)
        R = R + R
        return (R + R) + P
    if n % 4 == 3:
        R = scalarmult((n+1)/4,P)
        R = R + R
        return (R + R) - P
    return scalarmult(n/2,P)
    n = n // 2
    R = R + R

```

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Worst case: $\approx b$ doublings
plus $\approx b/2$ additions of $\pm P$.
On average $\approx b/3$ additions.

Width-2

```

def width2(n,P):
    if n == 0: return 0
    if n == 1: return P
    if n % 4 == 1:
        R = width2((n-1)/4,P)
        R = R + R
        R = R + R
        return (R + R) + P
    if n % 4 == 3:
        R = width2((n+1)/4,P)
        R = R + R
        R = R + R
        return (R + R) - P
    return width2(n/2,P)
    n = n // 2
    R = R + R

```

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,P):

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urn P

lt((n-1)/4,P)

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Worst case: $\approx b$ doublings

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On average $\approx b/3$ additions.

Width-2 signed sli

def window2(n,P,

if n == 0: ret

if n == 1: ret

if n == 3: ret

if n % 8 == 1:

R = window2(

R = R + R

R = R + R

return (R +

if n % 8 == 3:

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separated by ≥ 2 doublings.

Worst case: $\approx b$ doublings
plus $\approx b/2$ additions of $\pm P$.
On average $\approx b/3$ additions.

Width-2 signed sliding window

```
def window2(n,P,P3):
    if n == 0: return 0
    if n == 1: return P
    if n == 3: return P3
    if n % 8 == 1:
        R = window2((n-1)/8,P,P3)
        R = R + R
        R = R + R
        return (R + R) + P
    if n % 8 == 3:
        R = window2((n-3)/8,P,P3)
        R = R + R
        R = R + R
        return (R + R) + P3
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        R = R + R
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    if n % 8 == 3:
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        R = R + R
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Width-2 signed sliding windows

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        R = R + R
        R = R + R
        return (R + R) + P3

```

```

if n % 8 == 5:
    R = window2(
    R = R + R
    R = R + R
    return (R +
if n % 8 == 7:
    R = window2(
    R = R + R
    R = R + R
    return (R +
R = window2(n/
return R + R

def scalarmult(n
    return window2

```


Width-2 signed sliding windows

```

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        return (R + R) + P
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        R = R + R
        R = R + R
        return (R + R) + P3

```

```

if n % 8 == 5:
    R = window2((n+3)/8,P,P3)
    R = R + R
    R = R + R
    return (R + R) - P3
if n % 8 == 7:
    R = window2((n+1)/8,P,P3)
    R = R + R
    R = R + R
    return (R + R) - P
R = window2(n/2,P,P3)
return R + R

def scalarmult(n,P):
    return window2(n,P,P+P+P+P+P+P+P+P)

```

Width-2 signed sliding windows

```

def window2(n,P,P3):
    if n == 0: return 0
    if n == 1: return P
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    if n % 8 == 1:
        R = window2((n-1)/8,P,P3)
        R = R + R
        R = R + R
        return (R + R) + P
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        R = window2((n-3)/8,P,P3)
        R = R + R
        R = R + R
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```

```

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        return (R + R) - P3
    if n % 8 == 7:
        R = window2((n+1)/8,P,P3)
        R = R + R
        R = R + R
        return (R + R) - P
    R = window2(n/2,P,P3)
    return R + R

def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

signed sliding windows

```

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    n % 8 == 1: return P
    n % 8 == 3: return P3
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        R = R + R
        R = R + R
        return (R + R) + P
    n % 8 == 3:
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        R = R + R
        R = R + R
        return (R + R) + P3

```

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    R = R + R
    R = R + R
    return (R + R) - P
R = window2(n/2,P,P3)
return R + R

def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

Worst case
 $\approx b/3$ add
 On average

13

ding windows

P3):

urn 0

urn P

urn P3

 $(n-1)/8, P, P3)$ $R) + P$ $(n-3)/8, P, P3)$ $R) + P3$

```

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    R = R + R
    return (R + R) - P
R = window2(n/2, P, P3)
return R + R

```

```

def scalarmult(n, P):
    return window2(n, P, P+P+P)

```

14

Worst case: $\approx b$ d
 $\approx b/3$ additions of
On average $\approx b/4$

```

if n % 8 == 5:
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    R = R + R
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return R + R

def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

Worst case: $\approx b$ doublings per
 $\approx b/3$ additions of $\pm P$ or \pm
On average $\approx b/4$ additions.

```

if n % 8 == 5:
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def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

Worst case: $\approx b$ doublings plus $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
 On average $\approx b/4$ additions.

```

if n % 8 == 5:
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```

Worst case: $\approx b$ doublings plus $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
On average $\approx b/4$ additions.

Width-3 signed sliding windows:
Precompute $P, 3P, 5P, 7P$.
On average $\approx b/5$ additions.

```

if n % 8 == 5:
    R = window2((n+3)/8,P,P3)
    R = R + R
    R = R + R
    return (R + R) - P3
if n % 8 == 7:
    R = window2((n+1)/8,P,P3)
    R = R + R
    R = R + R
    return (R + R) - P
R = window2(n/2,P,P3)
return R + R

```

```

def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

Worst case: $\approx b$ doublings plus $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
On average $\approx b/4$ additions.

Width-3 signed sliding windows:
Precompute $P, 3P, 5P, 7P$.
On average $\approx b/5$ additions.

Width 4: Precompute
 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
On average $\approx b/6$ additions.


```

if n % 8 == 5:
    R = window2((n+3)/8,P,P3)
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    R = R + R
    return (R + R) - P3
if n % 8 == 7:
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    R = R + R
    R = R + R
    return (R + R) - P
R = window2(n/2,P,P3)
return R + R

def scalarmult(n,P):
    return window2(n,P,P+P+P)

```

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 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
On average $\approx b/6$ additions.

Cost of precomputation
eventually outweighs savings.
Optimal: $\approx b$ doublings plus
roughly $b/\lg b$ additions.

`% 8 == 5:`

```
    window2((n+3)/8,P,P3)
```

```
    R + R
```

```
    R + R
```

```
    return (R + R) - P3
```

`% 8 == 7:`

```
    window2((n+1)/8,P,P3)
```

```
    R + R
```

```
    R + R
```

```
    return (R + R) - P
```

```
    window2(n/2,P,P3)
```

```
    n R + R
```

```
    larmult(n,P):
```

```
    n window2(n,P,P+P+P)
```

Worst case: $\approx b$ doublings plus $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
On average $\approx b/4$ additions.

Width-3 signed sliding windows:
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On average $\approx b/6$ additions.

Cost of precomputation eventually outweighs savings.
Optimal: $\approx b$ doublings plus roughly $b/\lg b$ additions.

Double-s

Want to
 m, P, n, C

e.g. verify
by comp
computi
checking

Obvious
Comput

e.g. $b =$
 ≈ 256 do
 ≈ 256 do
 ≈ 50 ad
 ≈ 50 ad

$(n+3)/8, P, P3)$

$R) - P3$

$(n+1)/8, P, P3)$

$R) - P$

$2, P, P3)$

$, P) :$

$(n, P, P+P+P)$

Worst case: $\approx b$ doublings plus
 $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
 On average $\approx b/4$ additions.

Width-3 signed sliding windows:
 Precompute $P, 3P, 5P, 7P$.
 On average $\approx b/5$ additions.

Width 4: Precompute
 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
 On average $\approx b/6$ additions.

Cost of precomputation
 eventually outweighs savings.
 Optimal: $\approx b$ doublings plus
 roughly $b/\lg b$ additions.

Double-scalar mult

Want to quickly compute
 $m, P, n, Q \mapsto mP - nQ$

e.g. verify signature
 by computing $h =$
 computing $SB - k$
 checking whether

Obvious approach:
 Compute mP ; compute

e.g. $b = 256$:
 ≈ 256 doublings for
 ≈ 256 doublings for
 ≈ 50 additions for
 ≈ 50 additions for

Worst case: $\approx b$ doublings plus
 $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
 On average $\approx b/4$ additions.

Width-3 signed sliding windows:
 Precompute $P, 3P, 5P, 7P$.
 On average $\approx b/5$ additions.

Width 4: Precompute
 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
 On average $\approx b/6$ additions.

Cost of precomputation
 eventually outweighs savings.
 Optimal: $\approx b$ doublings plus
 roughly $b/\lg b$ additions.

Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB -$

Obvious approach:

Compute mP ; compute nQ ;

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

Worst case: $\approx b$ doublings plus
 $\approx b/3$ additions of $\pm P$ or $\pm 3P$.
 On average $\approx b/4$ additions.

Width-3 signed sliding windows:
 Precompute $P, 3P, 5P, 7P$.
 On average $\approx b/5$ additions.

Width 4: Precompute
 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
 On average $\approx b/6$ additions.

Cost of precomputation
 eventually outweighs savings.
 Optimal: $\approx b$ doublings plus
 roughly $b/\lg b$ additions.

Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB - hA$.

Obvious approach:

Compute mP ; compute nQ ; add.

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

case: $\approx b$ doublings plus
 additions of $\pm P$ or $\pm 3P$.
 average $\approx b/4$ additions.

signed sliding windows:
 compute $P, 3P, 5P, 7P$.
 average $\approx b/5$ additions.

: Precompute
 $P, 3P, 5P, 7P, 9P, 11P, 13P, 15P$.
 average $\approx b/6$ additions.

precomputation
 only outweighs savings.
 : $\approx b$ doublings plus
 $b/\lg b$ additions.

Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB - hA$.

Obvious approach:

Compute mP ; compute nQ ; add.

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

Joint do

Do much
 $2X + 2Y$

def sca

if m =

retu

if n =

retu

R = s

R = R

if m

if n

return

doublings plus
 $\pm P$ or $\pm 3P$.
 additions.
 sliding windows:
 $3P, 5P, 7P$.
 additions.
 compute
 $1P, 13P, 15P$.
 additions.
 cation
 ghs savings.
 doublings plus
 additions.

Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB - hA$.

Obvious approach:

Compute mP ; compute nQ ; add.

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

Joint doublings

Do much better by
 $2X + 2Y$ into $2(X + Y)$

```
def scalarmult2(m, n, P, Q):
    if m == 0:
        return scalar(P, 0)
    if n == 0:
        return scalar(Q, 0)
    R = scalarmult(P, m)
    R = R + scalar(Q, n)
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```

Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB - hA$.

Obvious approach:

Compute mP ; compute nQ ; add.

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

Joint doublings

Do much better by merging
 $2X + 2Y$ into $2(X + Y)$.

```
def scalarmult2(m,P,n,Q):
    if m == 0:
        return scalarmult(n,Q)
    if n == 0:
        return scalarmult(m,P)
    R = scalarmult2(m//2,P,
    R = R + R
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```


Double-scalar multiplication

Want to quickly compute
 $m, P, n, Q \mapsto mP + nQ$.

e.g. verify signature (R, S)
 by computing $h = H(R, M)$,
 computing $SB - hA$,
 checking whether $R = SB - hA$.

Obvious approach:

Compute mP ; compute nQ ; add.

e.g. $b = 256$:

≈ 256 doublings for mP ,

≈ 256 doublings for nQ ,

≈ 50 additions for mP ,

≈ 50 additions for nQ .

Joint doublings

Do much better by merging
 $2X + 2Y$ into $2(X + Y)$.

```
def scalarmult2(m,P,n,Q):
    if m == 0:
        return scalarmult(n,Q)
    if n == 0:
        return scalarmult(m,P)
    R = scalarmult2(m//2,P,n//2,Q)
    R = R + R
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```

Scalar multiplication

quickly compute

$$Q \mapsto mP + nQ.$$

Verify signature (R, S)

Computing $h = H(R, M)$,

finding $SB - hA$,

checking whether $R = SB - hA$.

Naïve approach:

compute mP ; compute nQ ; add.

256:

256 doublings for mP ,

256 doublings for nQ ,

256 additions for mP ,

256 additions for nQ .

Joint doublings

Do much better by merging

$$2X + 2Y \text{ into } 2(X + Y).$$

```
def scalarmult2(m, P, n, Q):
```

```
    if m == 0:
```

```
        return scalarmult(n, Q)
```

```
    if n == 0:
```

```
        return scalarmult(m, P)
```

```
    R = scalarmult2(m//2, P, n//2, Q)
```

```
    R = R + R
```

```
    if m % 2: R = R + P
```

```
    if n % 2: R = R + Q
```

```
    return R
```

For exam

$$35P = 2$$

$$31Q = 2$$

into $35P$

$$2(2(2(2($$

$$+P$$

$\approx b$ doubl

$\approx b/2$ ad

$\approx b/2$ ad

Combine

≈ 256 do

≈ 50 ad

≈ 50 ad

multiplication

compute

 $+ nQ$.re (R, S) $H(R, M)$, hA , $R = SB - hA$.compute nQ ; add.or mP ,or nQ , mP , nQ .Joint doublings

Do much better by merging
 $2X + 2Y$ into $2(X + Y)$.

```
def scalarmult2(m,P,n,Q):
    if m == 0:
        return scalarmult(n,Q)
    if n == 0:
        return scalarmult(m,P)
    R = scalarmult2(m//2,P,n//2,Q)
    R = R + R
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```

For example: merge

 $35P = 2(2(2(2(2P$ $31Q = 2(2(2(2Q+$ into $35P + 31Q =$ $2(2(2(2(2P+Q)+$ $+P+Q$. $\approx b$ doublings (me $\approx b/2$ additions of $\approx b/2$ additions of

Combine idea with

 ≈ 256 doublings fo ≈ 50 additions usin ≈ 50 additions usin

Joint doublings

Do much better by merging
 $2X + 2Y$ into $2(X + Y)$.

```
def scalarmult2(m,P,n,Q):
    if m == 0:
        return scalarmult(n,Q)
    if n == 0:
        return scalarmult(m,P)
    R = scalarmult2(m//2,P,n//2,Q)
    R = R + R
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```

- hA .

add.

For example: merge

$$35P = 2(2(2(2(2P)))) + P$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+Q)+P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows.

≈ 256 doublings for $b = 256$

≈ 50 additions using P ,

≈ 50 additions using Q .

Joint doublings

Do much better by merging
 $2X + 2Y$ into $2(X + Y)$.

```
def scalarmult2(m,P,n,Q):
    if m == 0:
        return scalarmult(n,Q)
    if n == 0:
        return scalarmult(m,P)
    R = scalarmult2(m//2,P,n//2,Q)
    R = R + R
    if m % 2: R = R + P
    if n % 2: R = R + Q
    return R
```

For example: merge

$$35P = 2(2(2(2(2P))) + P) + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) + P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

doublings

is much better by merging
 X and Y into $2(X + Y)$.

```
scalarmult2(m,P,n,Q):
```

```
    == 0:
```

```
    return scalarmult(n,Q)
```

```
    == 0:
```

```
    return scalarmult(m,P)
```

```
    return scalarmult2(m//2,P,n//2,Q)
```

```
    R = R + R
```

```
    % 2: R = R + P
```

```
    % 2: R = R + Q
```

```
    return R
```

For example: merge

$$35P = 2(2(2(2(2P)))) + P + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) \\ + P + Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

Batch verification

Verifying a batch of signatures
 need to compute

$$S_1 B = A$$

$$S_2 B = A$$

$$S_3 B = A$$

etc.

Obvious

Check each

by merging
($X + Y$).

$\text{mult}(m, P, n, Q)$:

$\text{mult}(n, Q)$

$\text{mult}(m, P)$

$\text{mult}(m//2, P, n//2, Q)$

$R + P$

$R + Q$

For example: merge

$$35P = 2(2(2(2(2P))) + P) + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) + P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

Batch verification

Verifying many signatures
need to be confident

$$S_1 B = R_1 + h_1 A_1$$

$$S_2 B = R_2 + h_2 A_2$$

$$S_3 B = R_3 + h_3 A_3$$

etc.

Obvious approach:

Check each equation

For example: merge

$$35P = 2(2(2(2(2P))) + P) + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) + P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately

For example: merge

$$35P = 2(2(2(2(2P)))) + P + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) + P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

For example: merge

$$35P = 2(2(2(2(2P)))) + P + P,$$

$$31Q = 2(2(2(2Q+Q)+Q)+Q)+Q$$

into $35P + 31Q =$

$$2(2(2(2(2P+Q)+Q)+Q)+P+Q) + P+Q.$$

$\approx b$ doublings (merged!),

$\approx b/2$ additions of P ,

$\approx b/2$ additions of Q .

Combine idea with windows: e.g.,

≈ 256 doublings for $b = 256$,

≈ 50 additions using P ,

≈ 50 additions using Q .

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

Much faster approach:

Check random linear combination
of the equations.

Example: merge

$$2(2(2(2P))) + P + P,$$

$$2(2(2(2Q+Q)+Q)+Q)+Q$$

$$P + 31Q =$$

$$2(2(2(2P+Q)+Q)+Q)+P+Q)$$

$$P+Q.$$

Sublings (merged!),

Conditions of P ,

Conditions of Q .

The idea with windows: e.g.,

Sublings for $b = 256$,

Conditions using P ,

Conditions using Q .

Batch verification

Verifying many signatures:

need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

Much faster approach:

Check random linear combination
of the equations.

Pick ind

128-bit

Check w

$$(z_1 S_1 +$$

$$z_1 R_1 +$$

$$z_2 R_2 +$$

$$z_3 R_3 +$$

(If \neq : S

Document

Easy to

forgeries

of foolin

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

Much faster approach:

Check random linear combination
of the equations.

Pick independent
128-bit z_1, z_2, z_3, \dots

Check whether

$$(z_1 S_1 + z_2 S_2 + z_3 S_3) B =$$

$$z_1 R_1 + (z_1 h_1) A_1 +$$

$$z_2 R_2 + (z_2 h_2) A_2 +$$

$$z_3 R_3 + (z_3 h_3) A_3 + \dots$$

(If \neq : See 2012 B

Doumen–Lange–O

Easy to prove:

forgeries have prob

of fooling this che

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

Much faster approach:

Check random linear combination
of the equations.

Pick independent uniform random
128-bit z_1, z_2, z_3, \dots

Check whether

$$(z_1 S_1 + z_2 S_2 + z_3 S_3 + \dots) B$$

$$z_1 R_1 + (z_1 h_1) A_1 +$$

$$z_2 R_2 + (z_2 h_2) A_2 +$$

$$z_3 R_3 + (z_3 h_3) A_3 + \dots$$

(If \neq : See 2012 Bernstein–
Doumen–Lange–Oosterwijk.)

Easy to prove:

forgeries have probability $\leq 2^{-128}$
of fooling this check.

Batch verification

Verifying many signatures:
need to be confident that

$$S_1 B = R_1 + h_1 A_1,$$

$$S_2 B = R_2 + h_2 A_2,$$

$$S_3 B = R_3 + h_3 A_3,$$

etc.

Obvious approach:

Check each equation separately.

Much faster approach:

Check random linear combination
of the equations.

Pick independent uniform random
128-bit z_1, z_2, z_3, \dots

Check whether

$$\begin{aligned} (z_1 S_1 + z_2 S_2 + z_3 S_3 + \dots) B = \\ z_1 R_1 + (z_1 h_1) A_1 + \\ z_2 R_2 + (z_2 h_2) A_2 + \\ z_3 R_3 + (z_3 h_3) A_3 + \dots \end{aligned}$$

(If \neq : See 2012 Bernstein–
Doumen–Lange–Oosterwijk.)

Easy to prove:

forges have probability $\leq 2^{-128}$
of fooling this check.

Verification

g many signatures:

be confident that

$$R_1 + h_1 A_1,$$

$$R_2 + h_2 A_2,$$

$$R_3 + h_3 A_3,$$

approach:

each equation separately.

ster approach:

andom linear combination

equations.

Pick independent uniform random
128-bit z_1, z_2, z_3, \dots

Check whether

$$(z_1 S_1 + z_2 S_2 + z_3 S_3 + \dots)B =$$

$$z_1 R_1 + (z_1 h_1)A_1 +$$

$$z_2 R_2 + (z_2 h_2)A_2 +$$

$$z_3 R_3 + (z_3 h_3)A_3 + \dots$$

(If \neq : See 2012 Bernstein–
Doumen–Lange–Oosterwijk.)

Easy to prove:

forgeries have probability $\leq 2^{-128}$

of fooling this check.

Multi-sc

Review o

1939 Br

$$\approx (1 + 1)$$

additions.

$$P \mapsto nP$$

1964 Str

$$\approx (1 + A)$$

additions.

$$P_1, \dots, A$$

if n_1, \dots

Pick independent uniform random
128-bit z_1, z_2, z_3, \dots

Check whether

$$(z_1 S_1 + z_2 S_2 + z_3 S_3 + \dots)B = \\ z_1 R_1 + (z_1 h_1)A_1 + \\ z_2 R_2 + (z_2 h_2)A_2 + \\ z_3 R_3 + (z_3 h_3)A_3 + \dots$$

(If \neq : See 2012 Bernstein–
Doumen–Lange–Oosterwijk.)

Easy to prove:

forges have probability $\leq 2^{-128}$
of fooling this check.

Multi-scalar multi

Review of asympto

1939 Brauer (wind

$$\approx (1 + 1/\lg b)b$$

additions to comp

$$P \mapsto nP \text{ if } n < 2^b$$

1964 Straus (joint

$$\approx (1 + k/\lg b)b$$

additions to comp

$$P_1, \dots, P_k \mapsto n_1 P_1$$

$$\text{if } n_1, \dots, n_k < 2^b.$$

Pick independent uniform random
128-bit z_1, z_2, z_3, \dots

Check whether

$$(z_1 S_1 + z_2 S_2 + z_3 S_3 + \dots)B = \\ z_1 R_1 + (z_1 h_1)A_1 + \\ z_2 R_2 + (z_2 h_2)A_2 + \\ z_3 R_3 + (z_3 h_3)A_3 + \dots$$

(If \neq : See 2012 Bernstein–
Doumen–Lange–Oosterwijk.)

Easy to prove:

forgeries have probability $\leq 2^{-128}$
of fooling this check.

Multi-scalar multiplication

Review of asymptotic speeds

1939 Brauer (windows):

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$$P_1, \dots, P_k \mapsto n_1 P_1 + \dots + n_k P_k \\ \text{if } n_1, \dots, n_k < 2^b.$$

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if $n_1, \dots, n_k < 2^b$.

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1976 Yao:

$$\approx (1 + k/\lg b)b$$

additions to compute

$$P \mapsto n_1 P, \dots, n_k P$$

if $n_1, \dots, n_k < 2^b$.

1976 Pippenger:

Similar asymptotic

but replace $\lg b$ with

Faster than Straus

if k is large.

(Knuth says "general")

as if speed were the

Multi-scalar multiplication

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1976 Yao:

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additions to compute

$$P \mapsto n_1 P, \dots, n_k P$$

$$\text{if } n_1, \dots, n_k < 2^b.$$

1976 Pippenger:

Similar asymptotics,

but replace $\lg b$ with $\lg(kb)$.

Faster than Straus and Yao

if k is large.

(Knuth says “generalization”
as if speed were the same.)

Multi-scalar multiplication

Review of asymptotic speeds:

1939 Brauer (windows):

$$\approx (1 + 1/\lg b)b$$

additions to compute

$$P \mapsto nP \text{ if } n < 2^b.$$

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Polynomial multiplication

of asymptotic speeds:

Knuth (windows):

$$(1 + k/\lg b)b$$

operations to compute

$$P \text{ if } n < 2^b.$$

Straus (joint doublings):

$$(k/\lg b)b$$

operations to compute

$$P_k \mapsto n_1 P_1 + \cdots + n_k P_k$$

$$, n_k < 2^b.$$

1976 Yao:

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More general

algorithm

ℓ sums of

$$\approx \left(\min_{i=1, \dots, \ell} \right)$$

if all coefficients

Within 1

Application

Asymptotic speeds:

(doublings):

compute

(doublings):

compute

$$P_1 + \dots + n_k P_k$$

1976 Yao:

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additions to compute

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More generally, Pippenger's

algorithm computes

ℓ sums of multiple

$$\approx \left(\min\{k, \ell\} + \frac{1}{\lg b} \right)$$

if all coefficients are

Within $1 + \epsilon$ of optimal

1976 Yao:

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additions to compute

$$P \mapsto n_1 P, \dots, n_k P$$

if $n_1, \dots, n_k < 2^b$.

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Faster than Straus and Yao
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More generally, Pippenger’s
algorithm computes

ℓ sums of multiples of k in P

$$\approx \left(\min\{k, \ell\} + \frac{k\ell}{\lg(k\ell b)} \right) b$$

if all coefficients are below 2^b

Within $1 + \epsilon$ of optimal.

1976 Yao:

$$\approx (1 + k/\lg b)b$$

additions to compute

$$P \mapsto n_1 P, \dots, n_k P$$

if $n_1, \dots, n_k < 2^b$.

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ℓ sums of multiples of k inputs.

$$\approx \left(\min\{k, \ell\} + \frac{k\ell}{\lg(k\ell b)} \right) b \text{ adds}$$

if all coefficients are below 2^b .

Within $1 + \epsilon$ of optimal.

1976 Yao:

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Within $1 + \epsilon$ of optimal.

Various special cases of
Pippenger’s algorithm were
reinvented and patented by
1993 Brickell–Gordon–McCurley–
Wilson, 1995 Lim–Lee, etc.
Is that the end of the story?

o:

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$, n_k < 2^b$.

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No! 198

If $n_1 \geq$

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More generally, Pippenger's algorithm computes ℓ sums of multiples of k inputs.

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If $n_1 \geq n_2 \geq \dots$ th

$n_1 P_1 + n_2 P_2 + n_3 P_3 + \dots$

$(n_1 - qn_2)P_1 + n_2 P_2 + \dots$

$n_3 P_3 + \dots$ where

Remarkably simple

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More generally, Pippenger's algorithm computes ℓ sums of multiples of k inputs.

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 $(n_1 - qn_2)P_1 + n_2(qP_1 + P_2 +$
 $n_3 P_3 + \dots)$ where $q = \lfloor n_1/n_2 \rfloor$

Remarkably simple;
competitive with Pippenger
for random choices of n_i 's;
much better memory usage.

More generally, Pippenger's algorithm computes ℓ sums of multiples of k inputs.

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Example

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–Lee, etc.

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If $n_1 \geq n_2 \geq \dots$ then

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Example of Bos–C

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$

$300P$, $146P$, $77P$,

No! 1989 Bos–Coster:

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9 Bos–Coster:

$n_2 \geq \dots$ then

$$n_2 P_2 + n_3 P_3 + \dots =$$

$$n_2) P_1 + n_2(q P_1 + P_2) +$$

$$\dots \text{ where } q = \lfloor n_1/n_2 \rfloor.$$

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Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

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Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce

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$$01001100$$

$$01001000$$

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$$00000000$$

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$$P_3 + \dots =$$

$$(qP_1 + P_2) +$$

$$q = \lfloor n_1/n_2 \rfloor.$$

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Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row

$$000100000 = 32$$

$$000010000 = 16$$

$$010011010 = 154$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$,
 $154P$, $146P$, $77P$,
Plus one extra add
add $146P$ into $154P$
obtaining $300P$.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$010011010 = 154 \leftarrow$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $154P$, $146P$, $77P$, $2P$, $1P$.

Plus one extra addition:

add $146P$ into $154P$,

obtaining $300P$.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$010011010 = 154 \leftarrow$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $154P$, $146P$, $77P$, $2P$, $1P$.

Plus one extra addition:

add $146P$ into $154P$,

obtaining $300P$.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$000001000 = 8 \leftarrow$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 2 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$000001000 = 8$$

$$001000101 = 69 \leftarrow$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 3 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$000001000 = 8$$

$$001000101 = 69$$

$$000001000 = 8 \leftarrow$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 4 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$000001000 = 8$$

$$000100101 = 37 \leftarrow$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 5 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000100000 = 32$$

$$000010000 = 16$$

$$000001000 = 8$$

$$000000101 = 5 \leftarrow$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 6 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000010000 = 16 \leftarrow$$

$$000010000 = 16$$

$$000001000 = 8$$

$$000000101 = 5$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 7 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000010000 = 16$$

$$000001000 = 8$$

$$000000101 = 5$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 7 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000001000 = 8 \leftarrow$$

$$000001000 = 8$$

$$000000101 = 5$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 8 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

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Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

$$000001000 = 8$$

$$000000101 = 5$$

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$$000000010 = 2$$

$$000000001 = 1$$

plus 8 additions.

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Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

$$000000101 = 5$$

$$000001000 = 8$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 8 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000101 = 5$$

$$000000011 = 3 \leftarrow$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 9 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000010 = 2 \leftarrow$$

$$000000011 = 3$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 10 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000010 = 2$$

$$000000001 = 1 \leftarrow$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 11 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

$$000000001 = 1$$

$$000000010 = 2$$

$$000000001 = 1$$

plus 11 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000001 = 1$$

$$000000001 = 1 \leftarrow$$

$$000000001 = 1$$

plus 12 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

$$000000001 = 1$$

$$000000001 = 1$$

plus 12 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

$$000000001 = 1$$

plus 12 additions.

Example of Bos–Coster:

$$000100000 = 32$$

$$000010000 = 16$$

$$100101100 = 300$$

$$010010010 = 146$$

$$001001101 = 77$$

$$000000010 = 2$$

$$000000001 = 1$$

Goal: Compute $32P$, $16P$,
 $300P$, $146P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

plus 12 additions.

Final addition chain: 1, 2, 3, 5, 8,
 16, 32, 37, 69, 77, 146, 154, 300.

Short, no temporary storage,
 low two-operand complexity.

of Bos–Coster:

$$00 = 32$$

$$00 = 16$$

$$00 = 300$$

$$010 = 146$$

$$01 = 77$$

$$010 = 2$$

$$001 = 1$$

compute $32P$, $16P$,
 $46P$, $77P$, $2P$, $1P$.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

plus 12 additions.

Final addition chain: 1, 2, 3, 5, 8,
 16, 32, 37, 69, 77, 146, 154, 300.

Short, no temporary storage,
 low two-operand complexity.

Revised

$$32P_1 + 2$$

$$77P_5 + 2$$

First con

and ther

$$32P_1 + 2$$

$$77P_5 + 2$$

Same sc

Ed25519

verify ba

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verifying

Coster:

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

plus 12 additions.

Final addition chain: 1, 2, 3, 5, 8,
16, 32, 37, 69, 77, 146, 154, 300.

Short, no temporary storage,
low two-operand complexity.

$2P$, $16P$,
 $2P$, $1P$.

Revised goal: Com

$$32P_1 + 16P_2 + 30$$

$$77P_5 + 2P_6 + 1P_7$$

First compute P'_4

and then recursive

$$32P_1 + 16P_2 + 15$$

$$77P_5 + 2P_6 + 1P_7$$

Same scalars show

Ed25519 batch ve

verify batch of 64

about twice as fas

verifying each sepa

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

plus 12 additions.

Final addition chain: 1, 2, 3, 5, 8,
16, 32, 37, 69, 77, 146, 154, 300.

Short, no temporary storage,
low two-operand complexity.

Revised goal: Compute
 $32P_1 + 16P_2 + 300P_3 + 146P_4 +$
 $77P_5 + 2P_6 + 1P_7$.

First compute $P'_4 = P_4 + P_3$
and then recursively compute
 $32P_1 + 16P_2 + 154P_3 + 146P_4 +$
 $77P_5 + 2P_6 + 1P_7$.

Same scalars show up as before.

Ed25519 batch verification:
verify batch of 64 signatures
about twice as fast as
verifying each separately.

Reduce largest row:

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0$$

$$000000000 = 0 \leftarrow$$

plus 12 additions.

Final addition chain: 1, 2, 3, 5, 8,
16, 32, 37, 69, 77, 146, 154, 300.

Short, no temporary storage,
low two-operand complexity.

Revised goal: Compute

$$32P_1 + 16P_2 + 300P_3 + 146P_4 + 77P_5 + 2P_6 + 1P_7.$$

First compute $P'_4 = P_4 + P_3$

and then recursively compute

$$32P_1 + 16P_2 + 154P_3 + 146P'_4 + 77P_5 + 2P_6 + 1P_7.$$

Same scalars show up as before.

Ed25519 batch verification:

verify batch of 64 signatures
about twice as fast as
verifying each separately.